

# MGS 2009: FUN Lecture 1

## *Lazy Functional Programming*

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# What Is a Functional Language? (1)

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  - Implicit state.
  - Computation essentially a sequence of side-effecting actions.

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- **Imperative Languages:**
  - Implicit state.
  - Computation essentially a sequence of side-effecting actions.
- **Declarative Languages** (Lloyd 1994):
  - **No** implicit state.
  - A program can be regarded as a theory.
  - Computation can be seen as deduction from this theory.
  - Examples: Logic and Functional Languages.

# What Is a Functional Language? (2)

Another perspective:

- *Algorithm = Logic + Control*

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Another perspective:

- ***Algorithm = Logic + Control***
- Declarative programming emphasises the logic (“what”) rather than the control (“how”).
- Examples:
  - Resolution (logic programming)
  - Lazy evaluation (found in some functional and logic languages)

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# What Is a Functional Language? (3)

Declarative languages for practical use tend to be only *weakly declarative*; i.e., not totally free of control aspects. For example:

- Equations in functional languages are directed.
- Order of patterns often matters for pattern matching.
- Constructs for taking control over the order of evaluation.

# What Is a Functional Language? (4)

Exactly what constitute a functional language is somewhat contentious.

Pragmatically, a functional language is one that encourages a mostly declarative, **functional style** of programming.

Typical features/characteristics:

- Functions are first-class entities.
- Computation expressed through function application.
- Recursive (and co-recursive) definitions.

# What Is a Functional Language? (5)

This “definition” covers both:

- **Pure** functional languages: no side effects
  - (Weakly) declarative: equational reasoning valid (referentially transparent).
  - Example: Haskell
- **Mostly** functional languages: some side effects, e.g. for I/O.
  - Equational reasoning with care.
  - Examples: ML, OCaml, Scheme, Erlang

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- In this and the following lectures we will explore ***Purely Functional Programming*** through the use of Haskell.
- Theme of today: ***Relinquishing control: exploiting lazy evaluation***

Will assume some familiarity with functional programming in a language like Haskell or ML. Will explain Haskell syntax and other points as needed: ***Just ask!***

# Evaluation Orders (1)

Consider:

```
sqr x = x * x
```

```
dbl x = x + x
```

```
main = sqr (dbl (2 + 3))
```

Many possible reduction orders. Innermost, leftmost **redex** first is called **Applicative Order Reduction** (AOR):

```
main ⇒ sqr (dbl (2 + 3)) ⇒ sqr (dbl 5)  
⇒ sqr (5 + 5) ⇒ sqr 10 ⇒ 10 * 10 ⇒ 100
```

This is just **Call-By-Value**.



# Evaluation Orders (2)

Outermost, leftmost redex first is called **Normal Order Reduction** (NOR):

```
main ⇒ sqr (dbl (2 + 3))  
⇒ dbl (2 + 3) * dbl (2 + 3)  
⇒ ((2 + 3) + (2 + 3)) * dbl (2 + 3)  
⇒ (5 + (2 + 3)) * dbl (2 + 3)  
⇒ (5 + 5) * dbl (2 + 3) ⇒ 10 * dbl (2 + 3)  
⇒ ... ⇒ 10 * 10 ⇒ 100
```

(Applications of arithmetic operations only considered redexes once arguments are numbers.)

Demand-driven evaluation or **Call-By-Need**

# Why Normal Order Reduction? (1)

NOR seems rather inefficient. Any use?

- Best possible termination properties. Two important theorems from the  $\lambda$ -calculus:
  - Church-Rosser Theorem I:  
No term has more than one normal form.
  - Church-Rosser Theorem II:  
If a term has a normal form, then NOR will find it.

# Why Normal Order Reduction? (2)

- More expressive power; e.g.:
  - “Infinite” data structures
  - Circular programming
- More declarative code as control aspects (order of evaluation) left implicit.

# Strict vs. Non-strict Semantics (1)

- $\perp$ , or “bottom”, the *undefined value*, representing *errors* and *non-termination*.
- A function  $f$  is *strict* iff:

$$f \perp = \perp$$

For example,  $+$  is strict in both its arguments:

$$(0/0) + 1 = \perp + 1 = \perp$$

$$1 + (0/0) = 1 + \perp = \perp$$

# Strict vs. Non-strict Semantics (2)

Consider:

$$foo\ x = 1$$

What is the value of  $foo\ (0/0)$ ?

- AOR:  $foo\ (\underline{0/0}) \Rightarrow \perp$   
Conceptually,  $foo\ \perp = \perp$ ; i.e.,  $foo$  is strict.
- NOR:  $\underline{foo\ (0/0)} \Rightarrow 1$   
Conceptually,  $foo\ \perp = 1$ ; i.e.,  $foo$  is non-strict.

Thus, NOR results in non-strict semantics.

Note: NOR gave well-defined result, AOR did not.

# Lazy Evaluation (1)

Lazy evaluation is an **technique for implementing NOR** more efficiently:

- An expression is evaluated **only if needed**.
- **Sharing** employed to ensure any one expression evaluated at most once.

# Lazy Evaluation (2)

```
sqr (dbl (2 + 3))
```

# Lazy Evaluation (2)

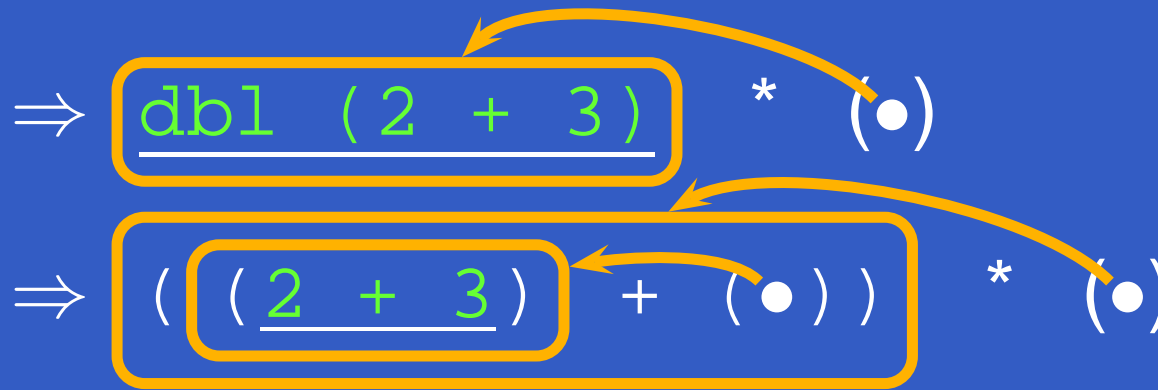
sqr (dbl (2 + 3))

⇒ dbl (2 + 3) \* (•)



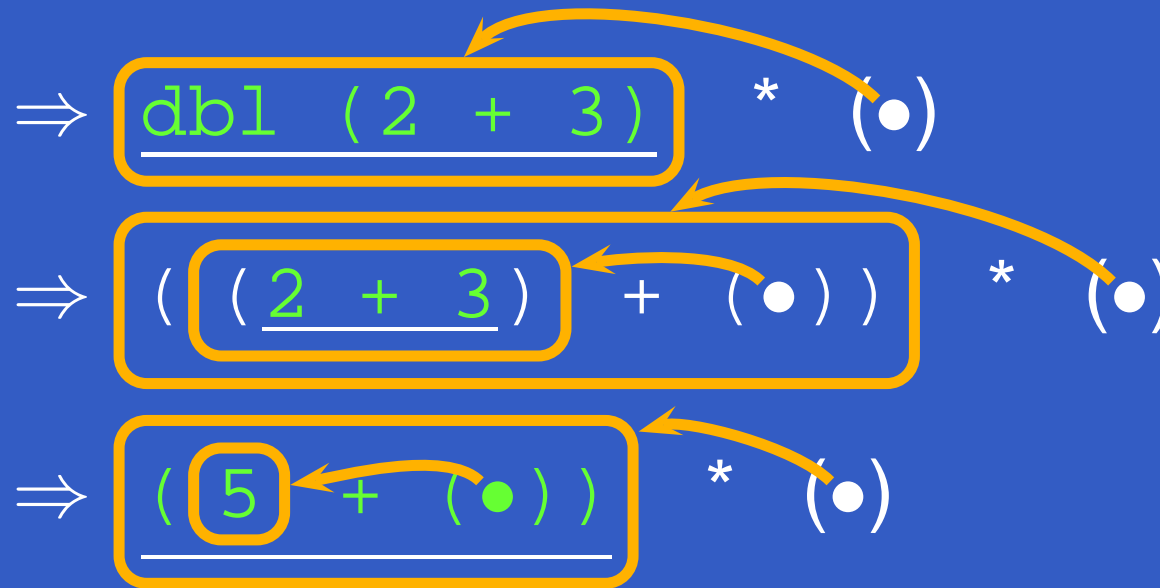
# Lazy Evaluation (2)

sqr (dbl (2 + 3))



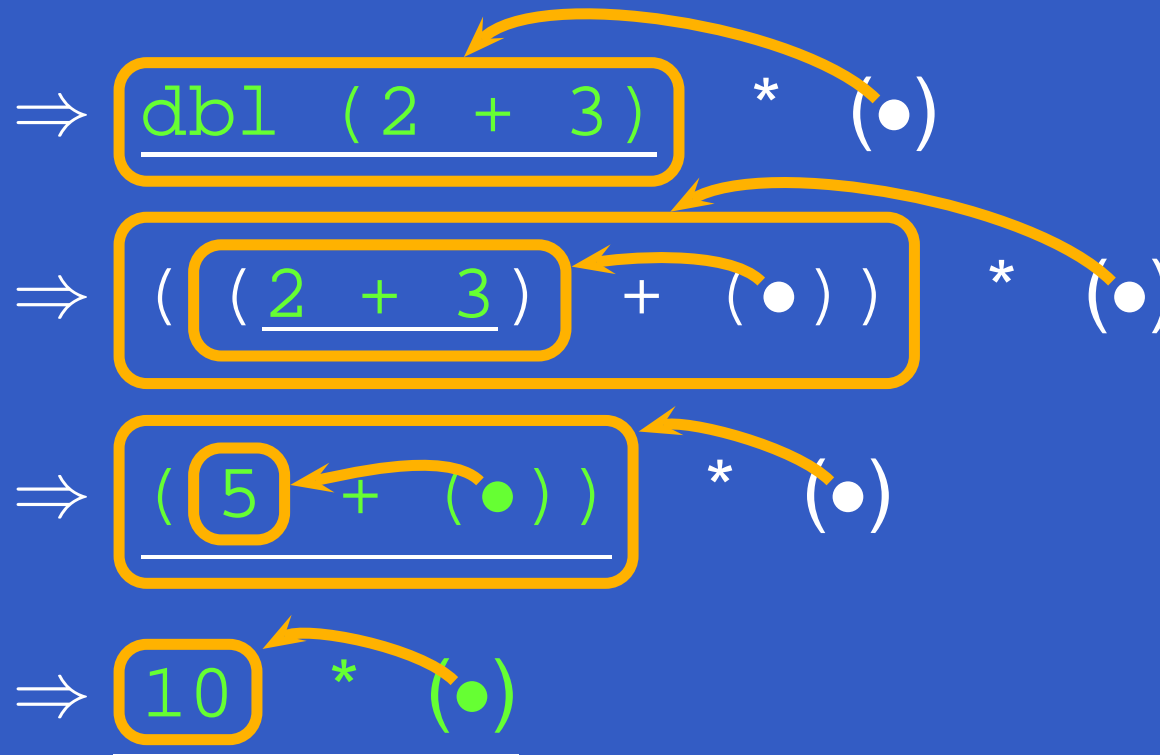
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# Lazy Evaluation (2)

sqr (dbl (2 + 3))



# Infinite Data Structures (1)

```
take 0 xs      = []
```

```
take n []      = []
```

```
take n (x:xs) = x : take (n-1) xs
```

```
from n = n : from (n+1)
```

```
nats = from 0
```

```
main = take 5 nats
```

# Infinite Data Structures (2)

main

nats

# Infinite Data Structures (2)

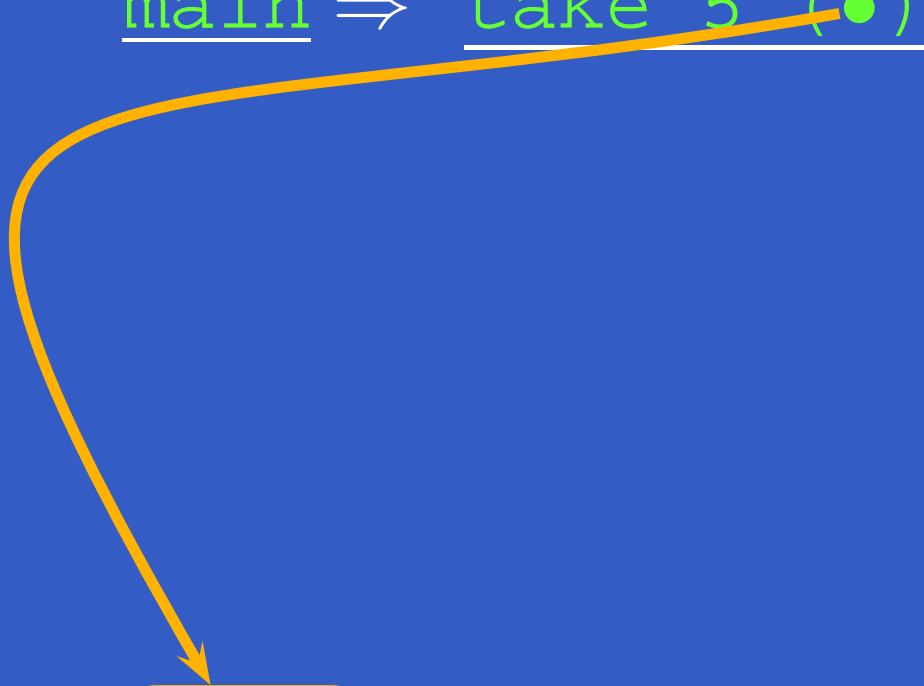
main  $\Rightarrow^1$  take 5 (●)

nats



# Infinite Data Structures (2)

main  $\Rightarrow^1$  take 5 (●)

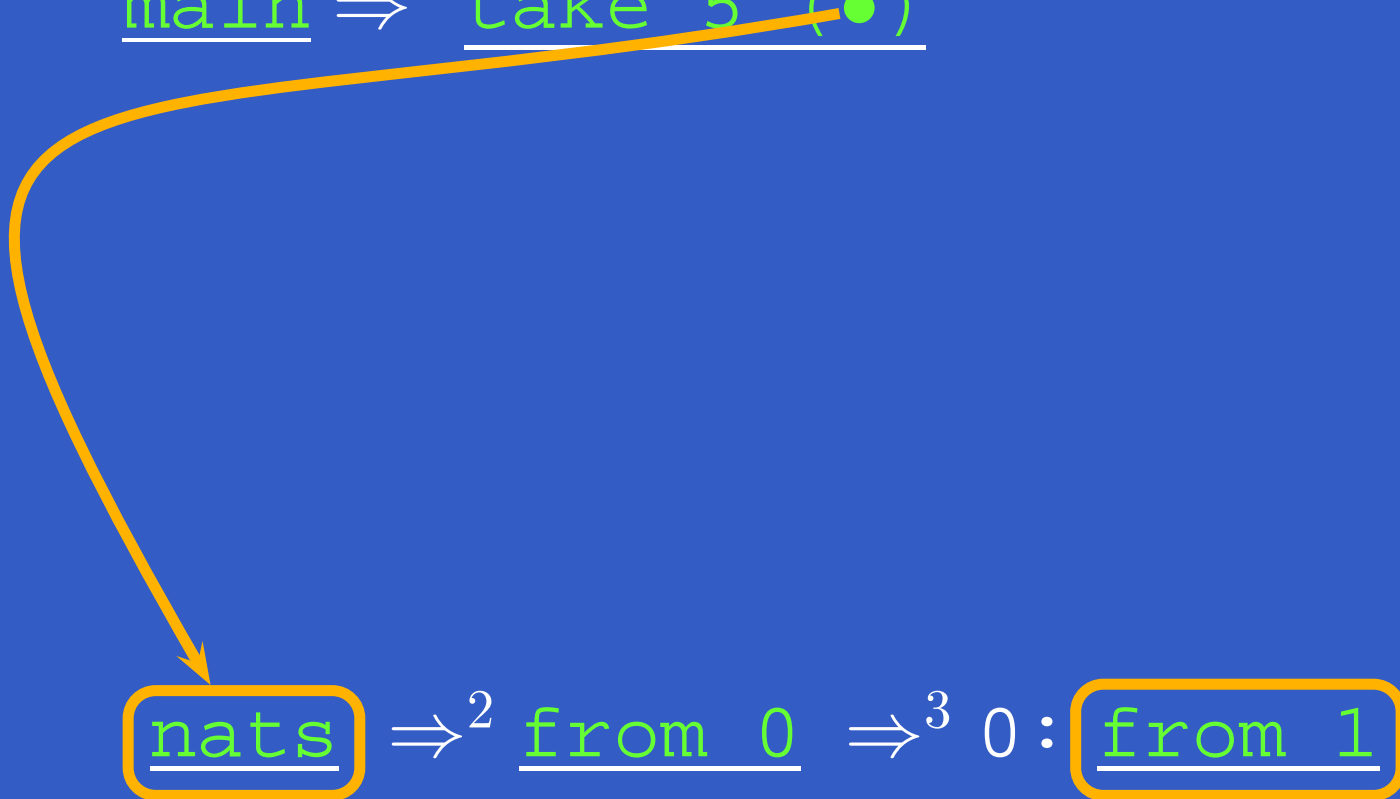


nats  $\Rightarrow^2$  from 0



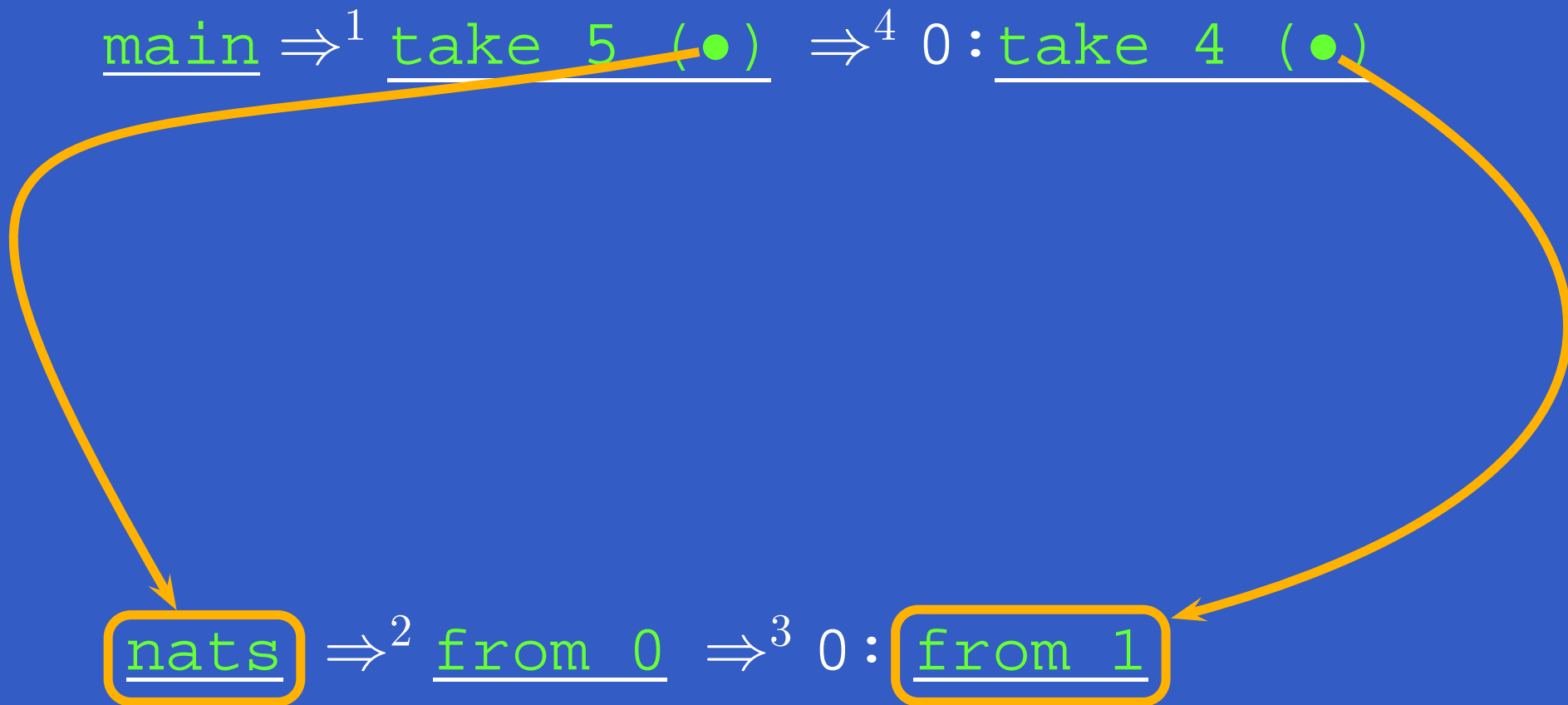
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main  $\Rightarrow^1$  take 5 (●)



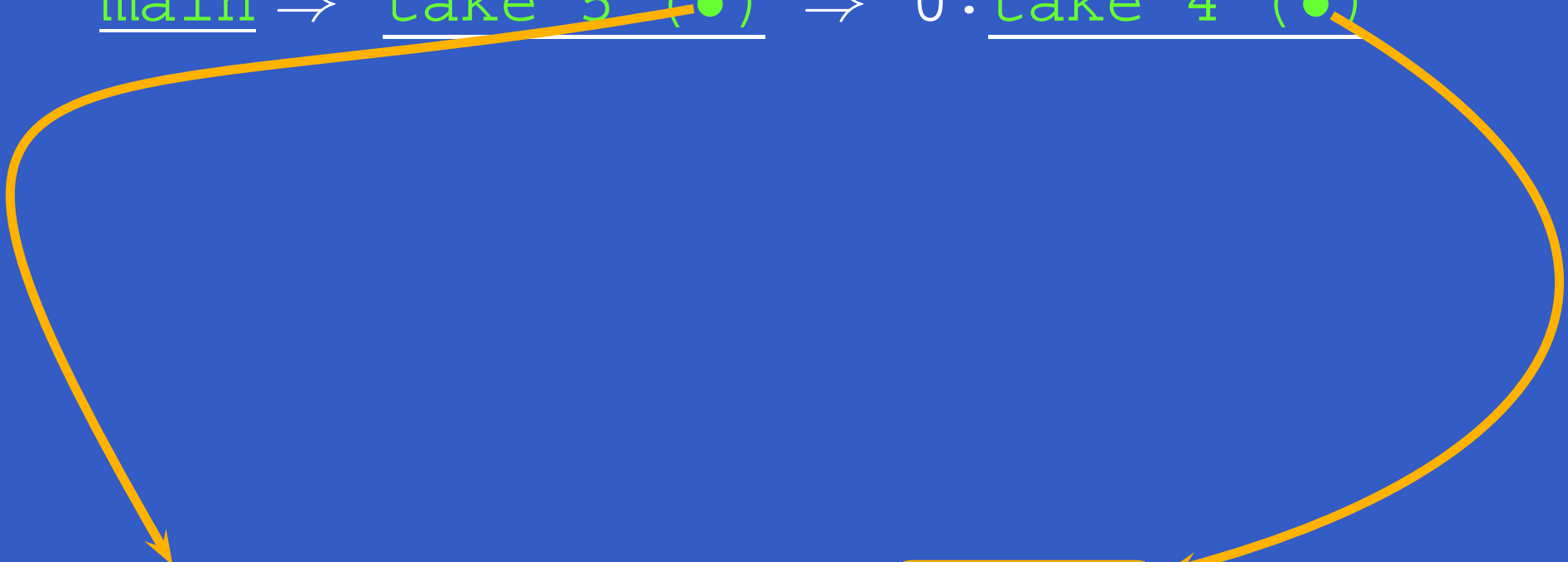
# Infinite Data Structures (2)

main  $\Rightarrow^1$  take 5 (●)  $\Rightarrow^4$  0 : take 4 (●)



# Infinite Data Structures (2)

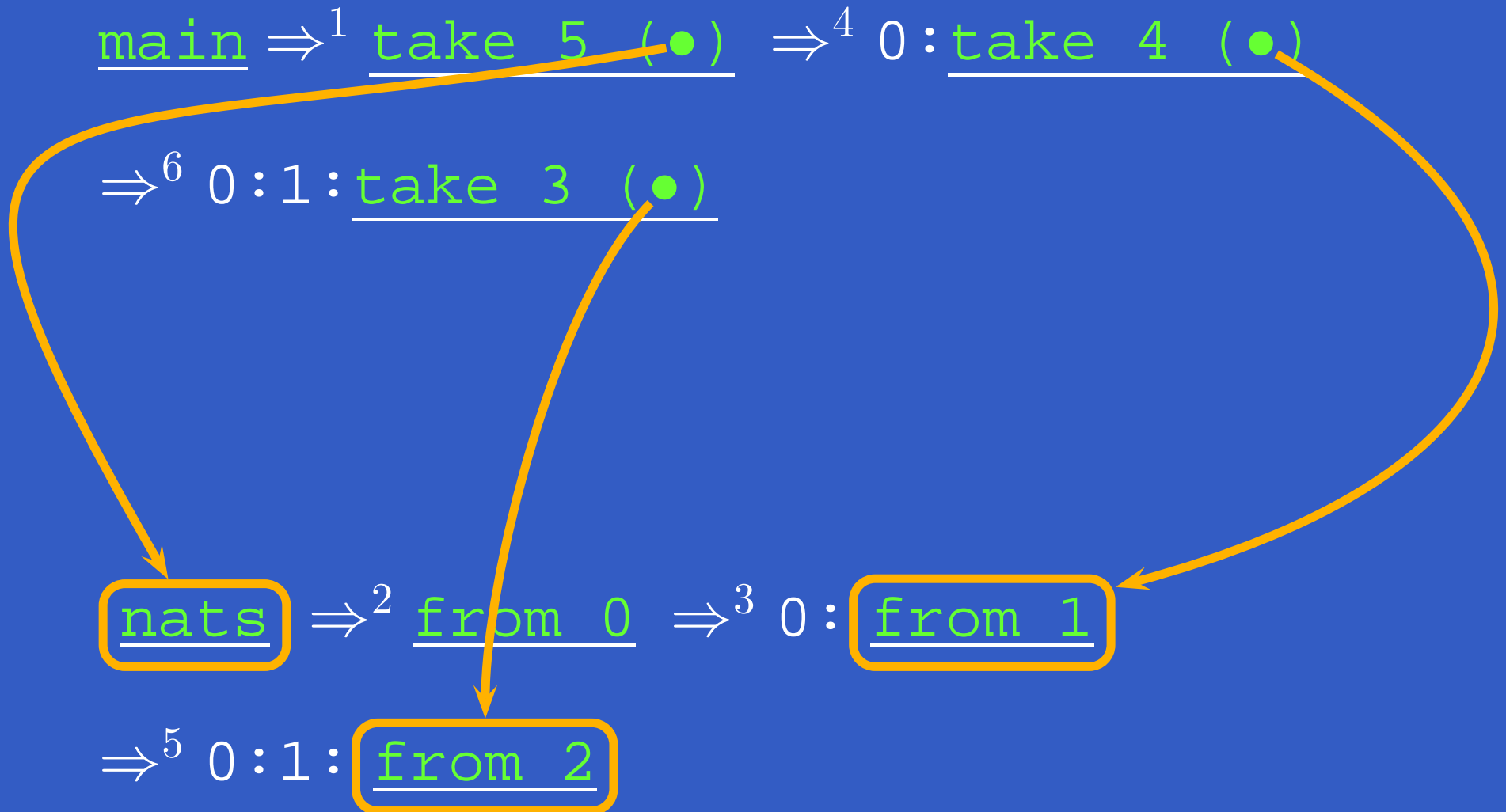
main  $\Rightarrow^1$  take 5 (●)  $\Rightarrow^4$  0 : take 4 (●)



nats  $\Rightarrow^2$  from 0  $\Rightarrow^3$  0 : from 1

$\Rightarrow^5$  0 : 1 : from 2

# Infinite Data Structures (2)



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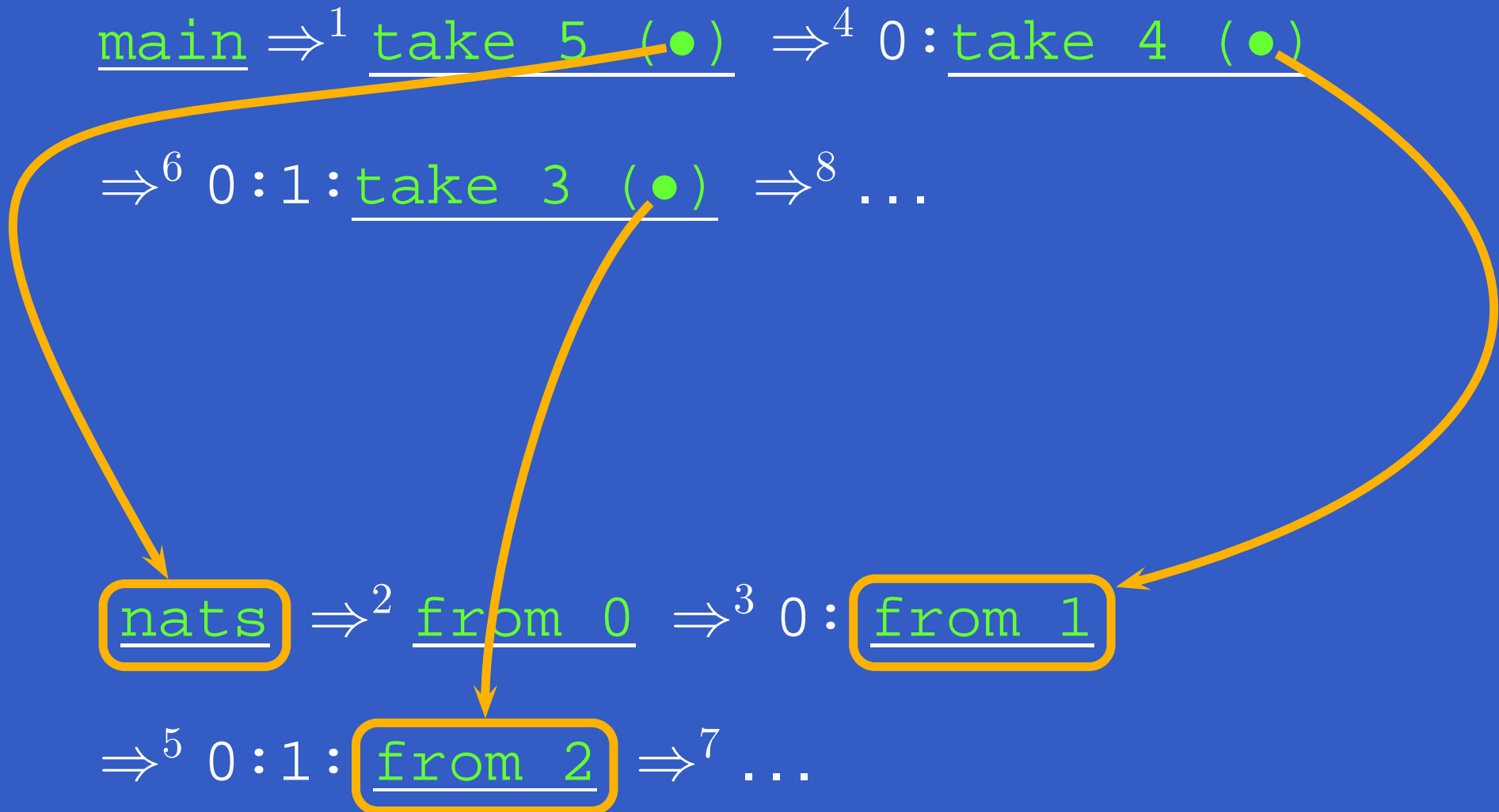
main  $\Rightarrow^1$  take 5 (●)  $\Rightarrow^4$  0 : take 4 (●)

$\Rightarrow^6$  0 : 1 : take 3 (●)

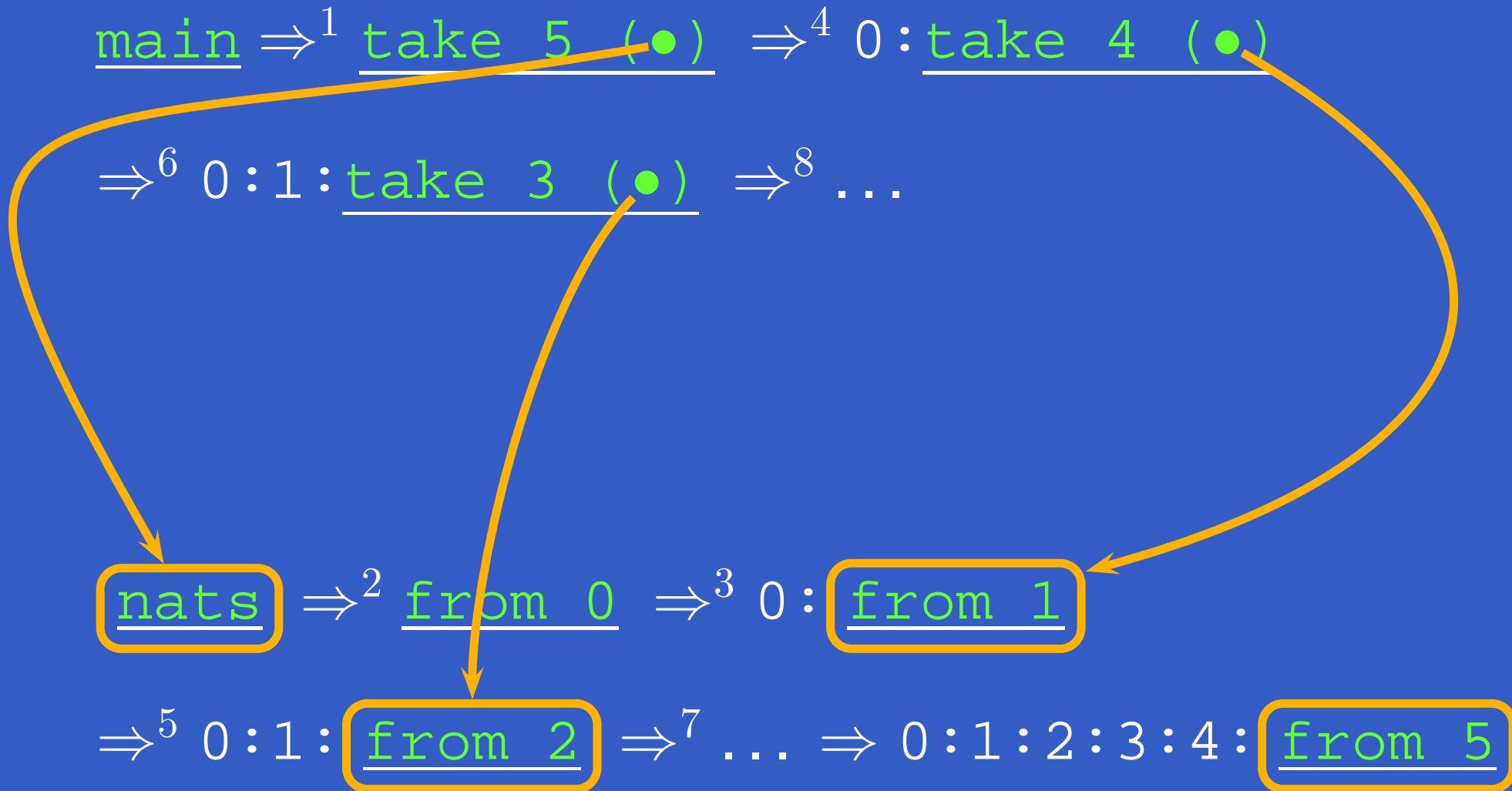
nats  $\Rightarrow^2$  from 0  $\Rightarrow^3$  0 : from 1

$\Rightarrow^5$  0 : 1 : from 2  $\Rightarrow^7$  ...

# Infinite Data Structures (2)



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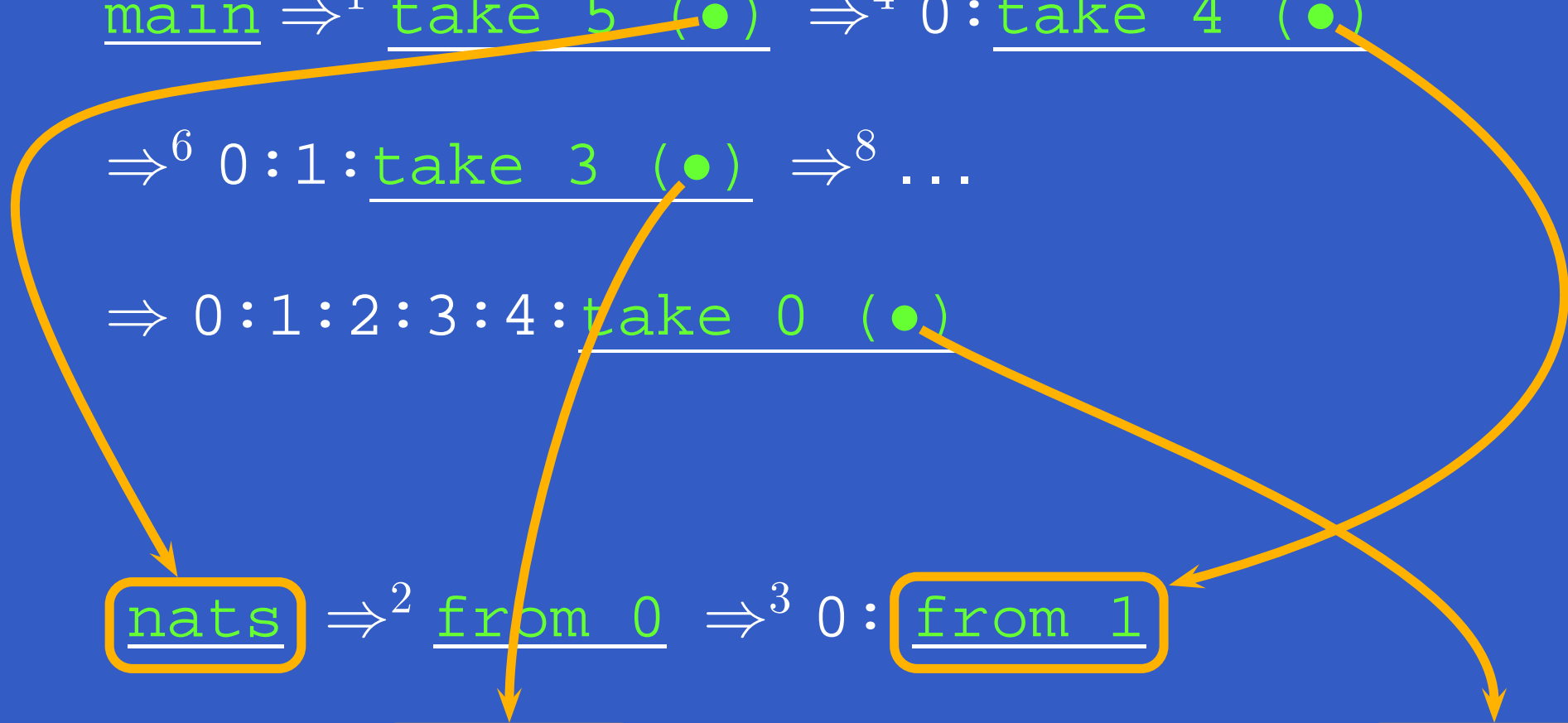
main  $\Rightarrow^1$  take 5 (●)  $\Rightarrow^4$  0: take 4 (●)

$\Rightarrow^6$  0:1: take 3 (●)  $\Rightarrow^8$  ...

$\Rightarrow$  0:1:2:3:4: take 0 (●)

nats  $\Rightarrow^2$  from 0  $\Rightarrow^3$  0: from 1

$\Rightarrow^5$  0:1: from 2  $\Rightarrow^7$  ...  $\Rightarrow$  0:1:2:3:4: from 5





# Infinite Data Structures (2)

main  $\Rightarrow^1$  take 5 (●)  $\Rightarrow^4$  0: take 4 (●)

$\Rightarrow^6$  0:1: take 3 (●)  $\Rightarrow^8$  ...

$\Rightarrow$  0:1:2:3:4: take 0 (●)  $\Rightarrow$  [0,1,2,3,4]

nats  $\Rightarrow^2$  from 0  $\Rightarrow^3$  0: from 1

$\Rightarrow^5$  0:1: from 2  $\Rightarrow^7$  ...  $\Rightarrow$  0:1:2:3:4: from 5

# Circular Data Structures (2)

```
take 0 xs      = []
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```
take n (x:xs) = x : take (n-1) xs
```

```
ones = 1 : ones
```

```
main = take 5 ones
```

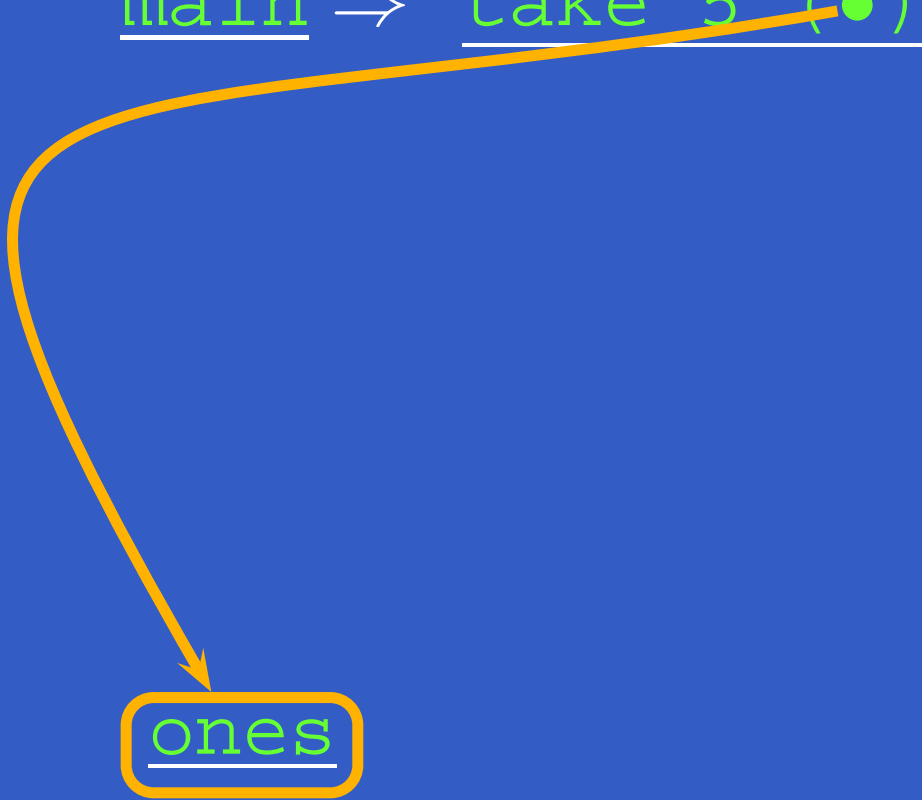
# Circular Data Structures (2)

main

ones

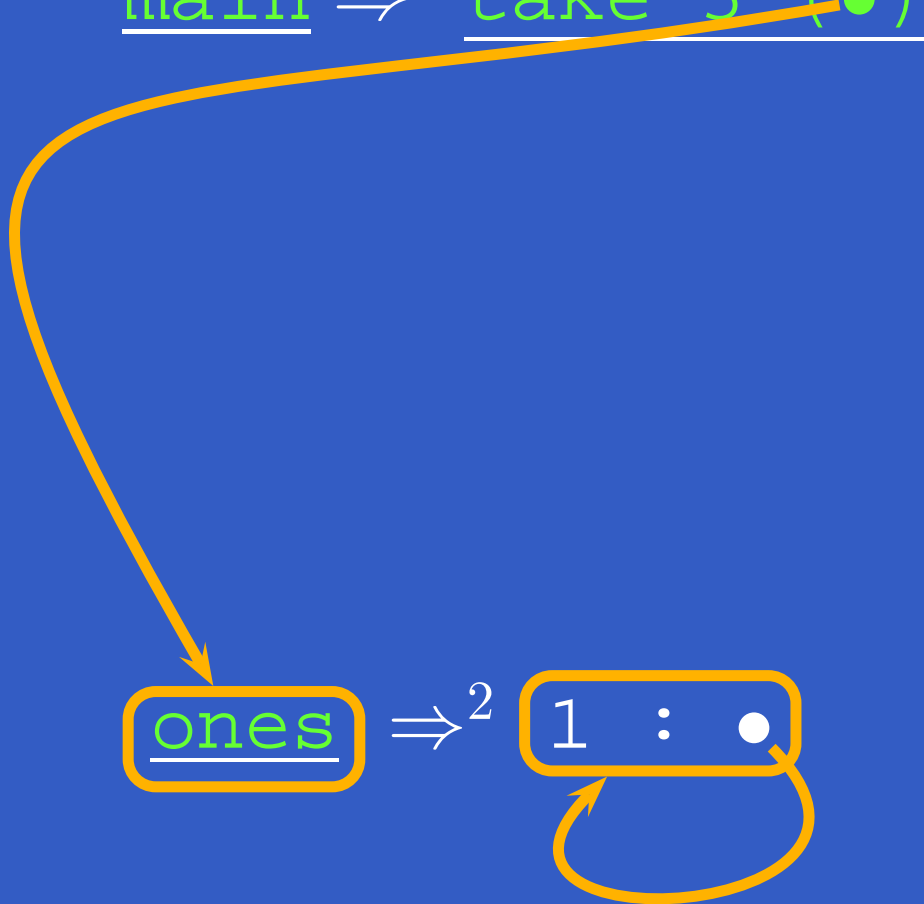
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main  $\Rightarrow^1$  take 5 (●)



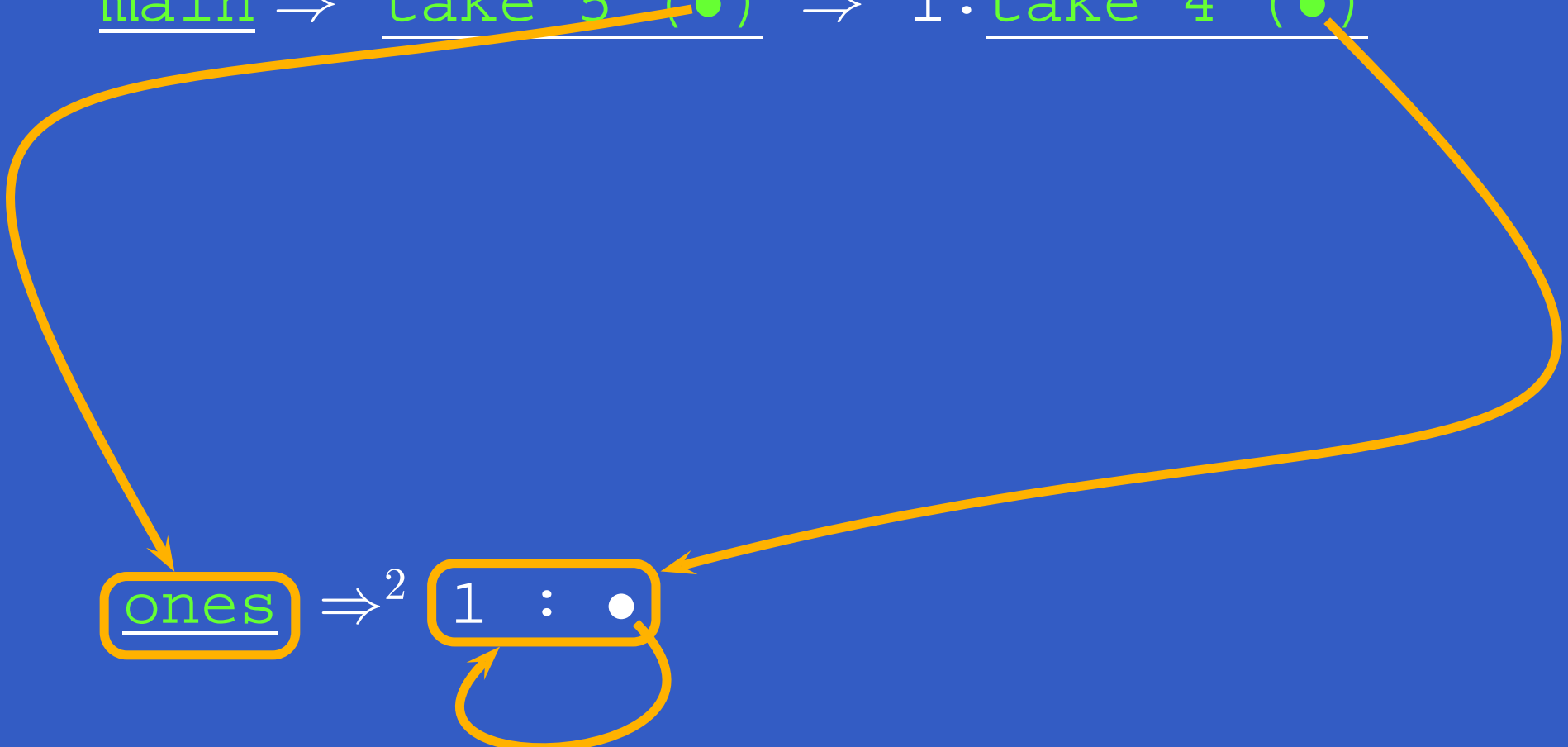
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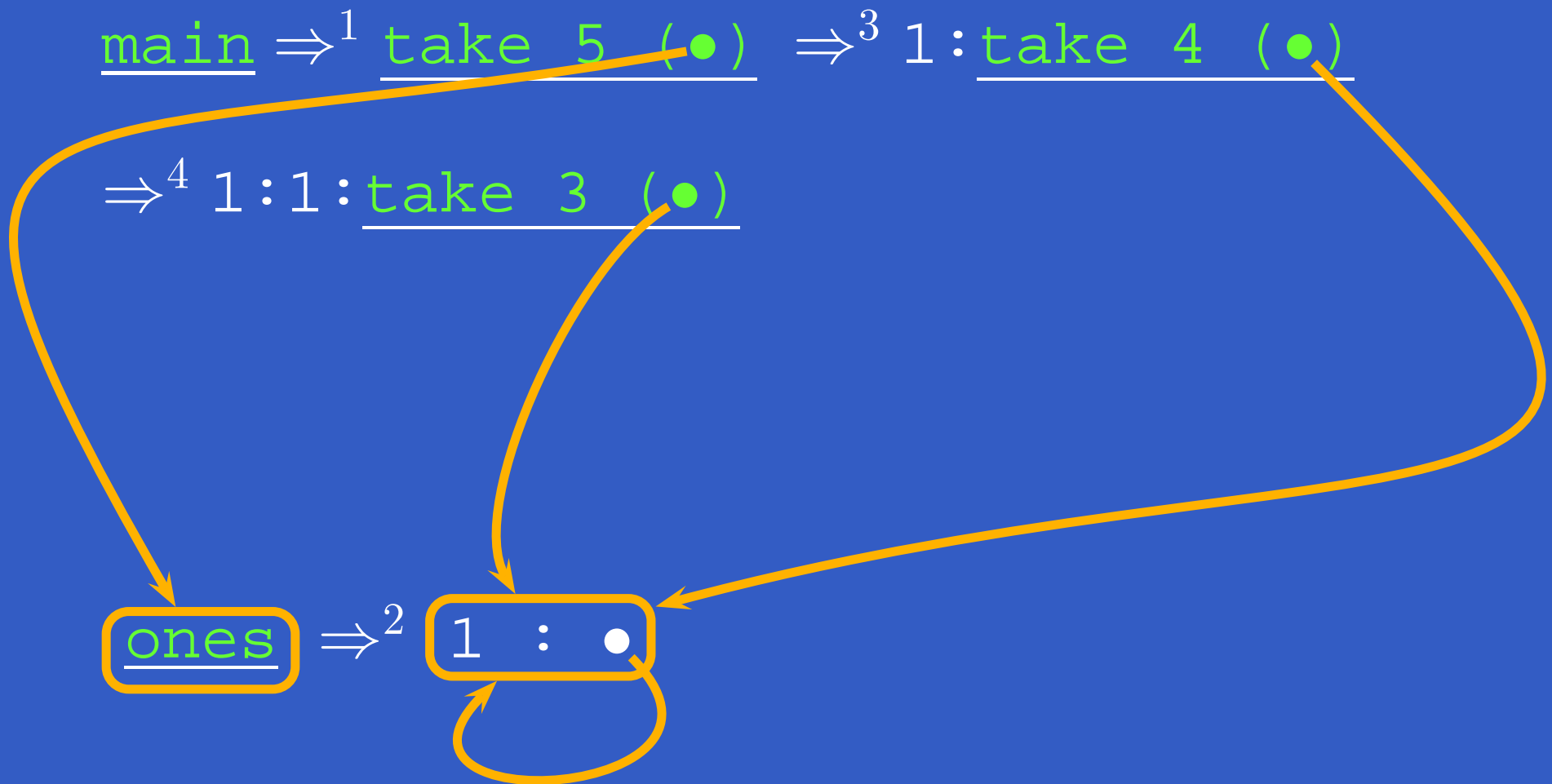


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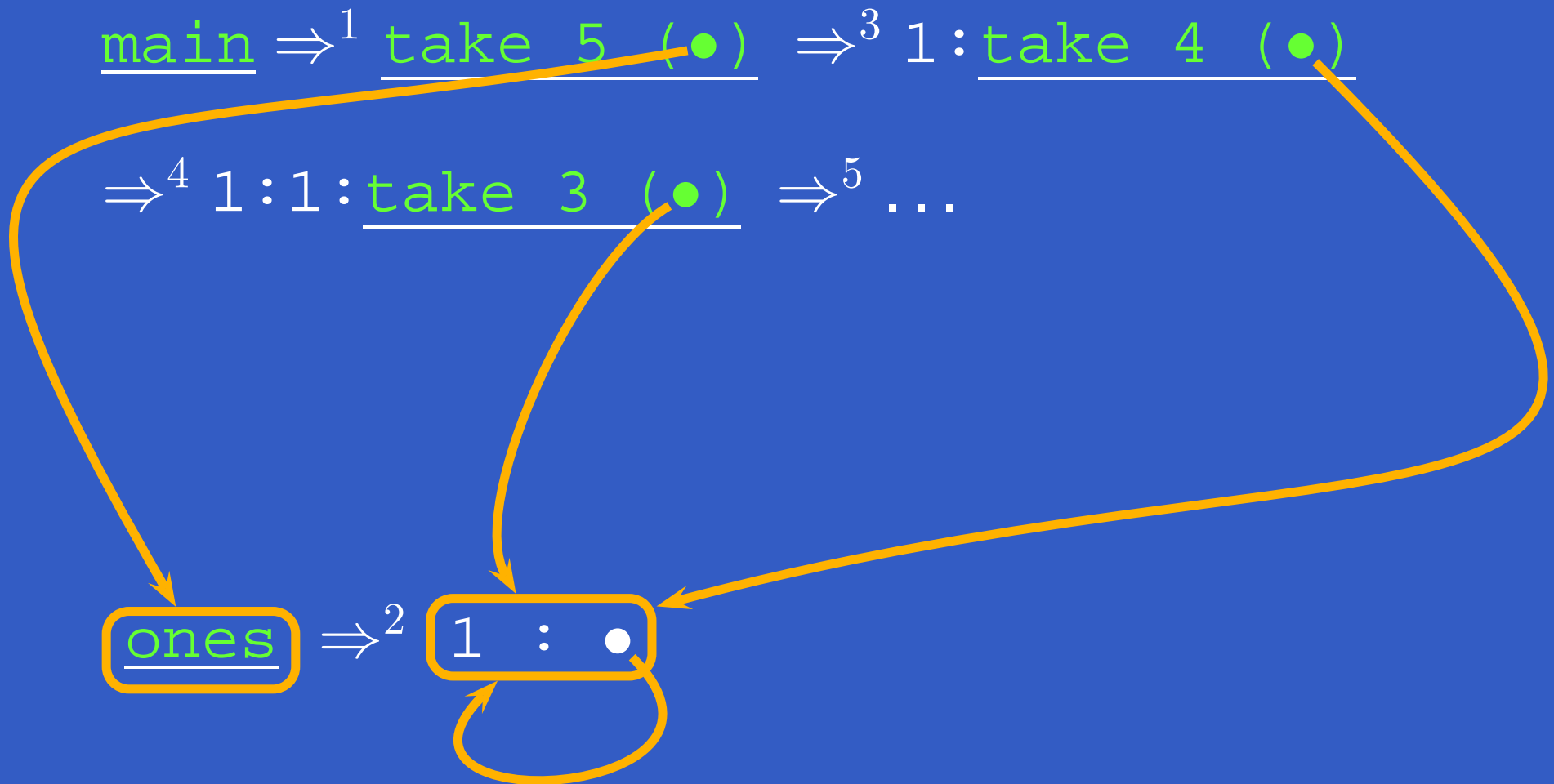
main  $\Rightarrow^1$  take 5 (●)  $\Rightarrow^3$  1 : take 4 (●)



# Circular Data Structures (2)

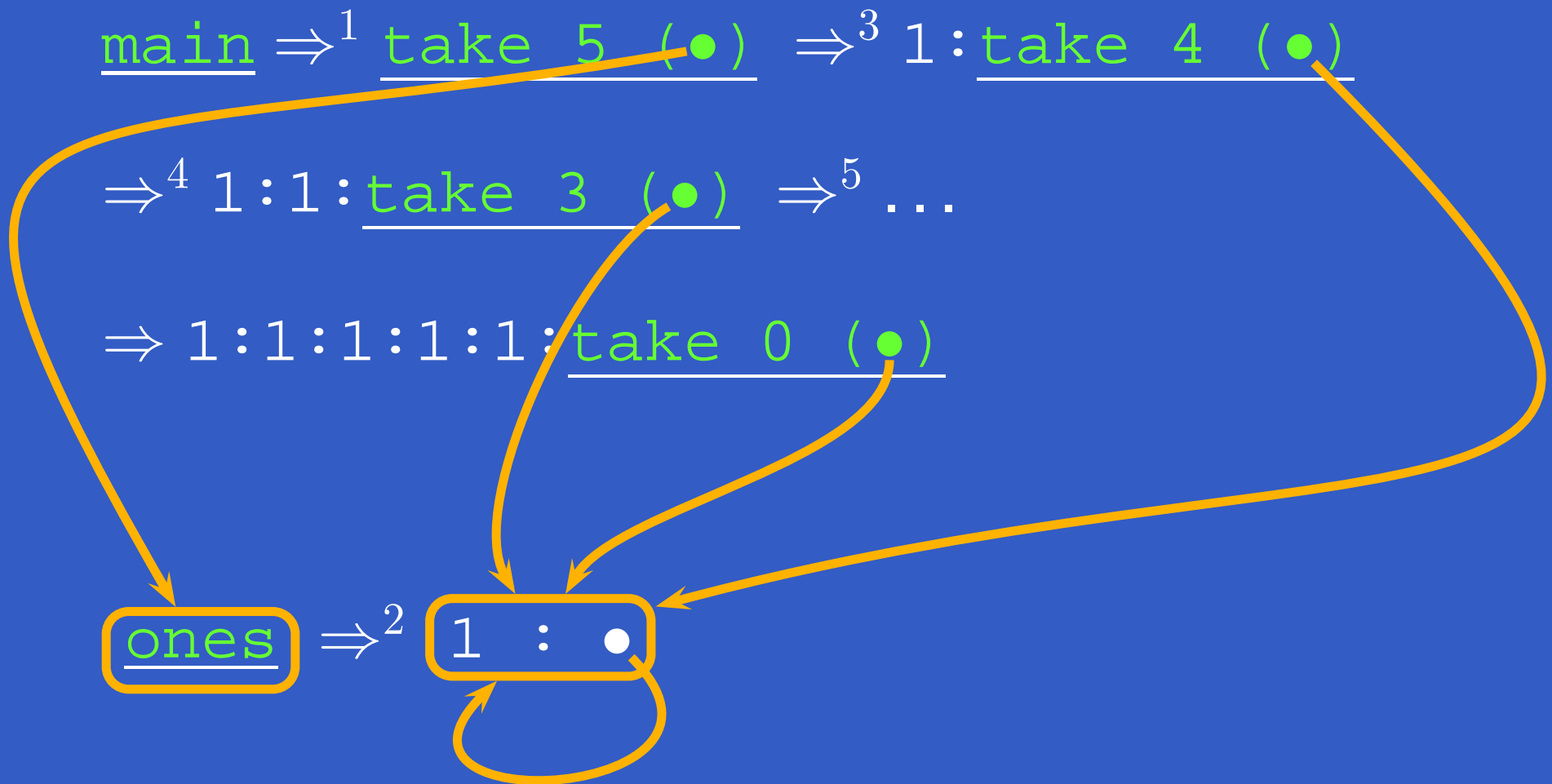


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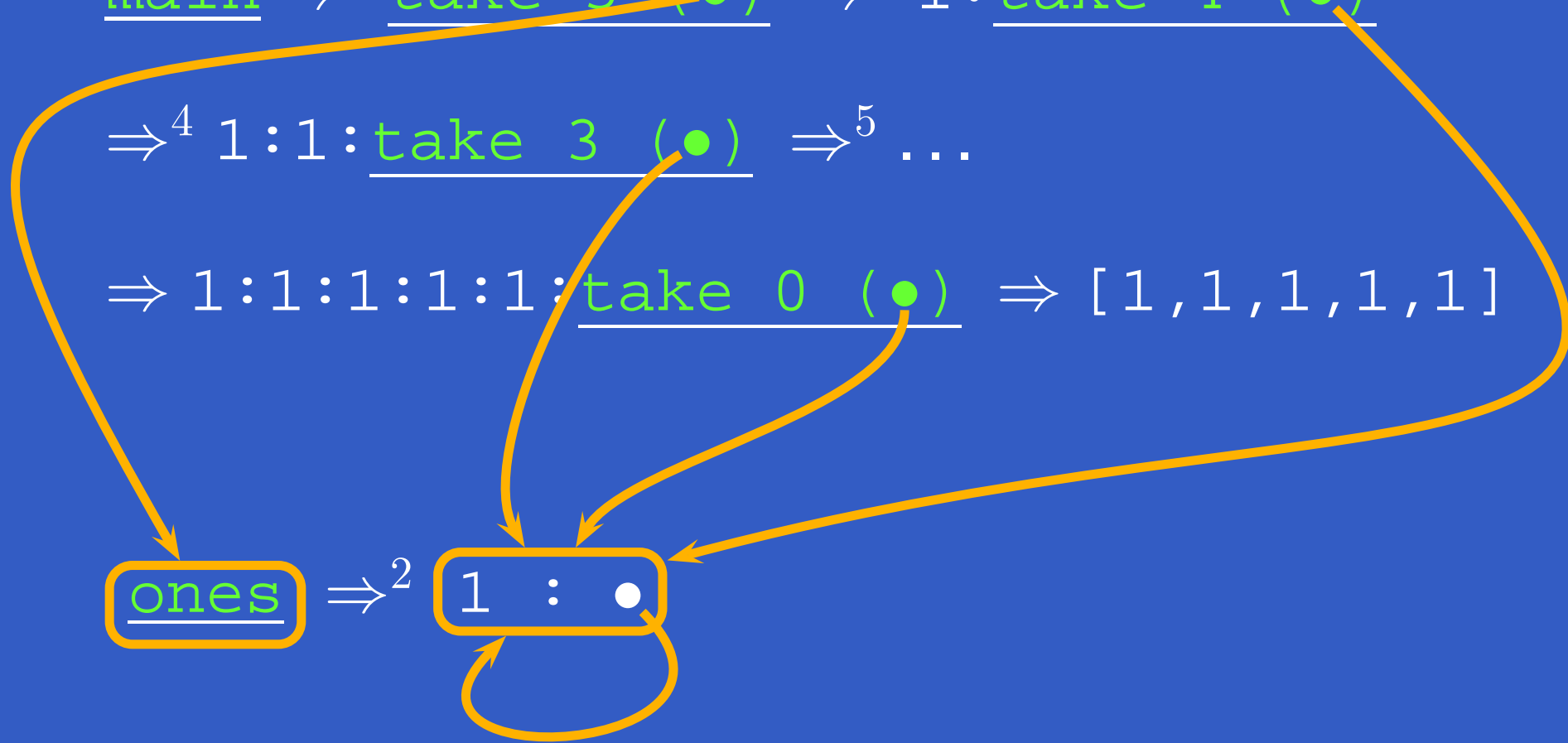
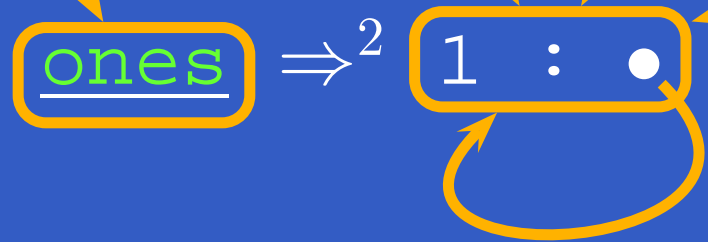


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main  $\Rightarrow^1$  take 5 (●)  $\Rightarrow^3$  1 : take 4 (●)  
 $\Rightarrow^4$  1 : 1 : take 3 (●)  $\Rightarrow^5$  ...  
 $\Rightarrow$  1 : 1 : 1 : 1 : 1 : take 0 (●)  $\Rightarrow$  [1, 1, 1, 1, 1]



# Circular Programming (1)

A non-empty tree type:

```
data Tree = Leaf Int | Node Tree Tree
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Suppose we would like to write a function that replaces each leaf integer in a given tree with the ***smallest*** integer in that tree.

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# Circular Programming (1)

A non-empty tree type:

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data Tree = Leaf Int | Node Tree Tree
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Suppose we would like to write a function that replaces each leaf integer in a given tree with the ***smallest*** integer in that tree.

How many passes over the tree are needed?

***One!***

# Circular Programming (2)

Write a function that replaces all leaf integers by a given integer, and returns the smallest integer of the given tree:

```
fmr :: Int -> Tree -> (Tree, Int)
```

```
fmr m (Leaf i) = (Leaf m, i)
```

```
fmr m (Node tl tr) =  
  (Node tl' tr', min ml mr)
```

where

```
(tl', ml) = fmr m tl
```

```
(tr', mr) = fmr m tr
```

# Circular Programming (3)

For a given tree  $t$ , the desired tree is obtained as the **solution** to the equation:

$$(t', m) = \text{fmr } m \ t$$

Thus:

$$\text{findMinReplace } t = t'$$

where

$$(t', m) = \text{fmr } m \ t$$

Intuitively, this works because  $\text{fmr}$  can compute its result without needing to know the **value** of  $m$ .



# A Simple Spreadsheet Evaluator

	a	b	c
1	c3 + c2		
2	a3 * b2	2	a2 + b2
3	7		a2 + a3

s



	a	b	c
1	37		
2	14	2	16
3	7		21

r

```
r = array (bounds s)
```

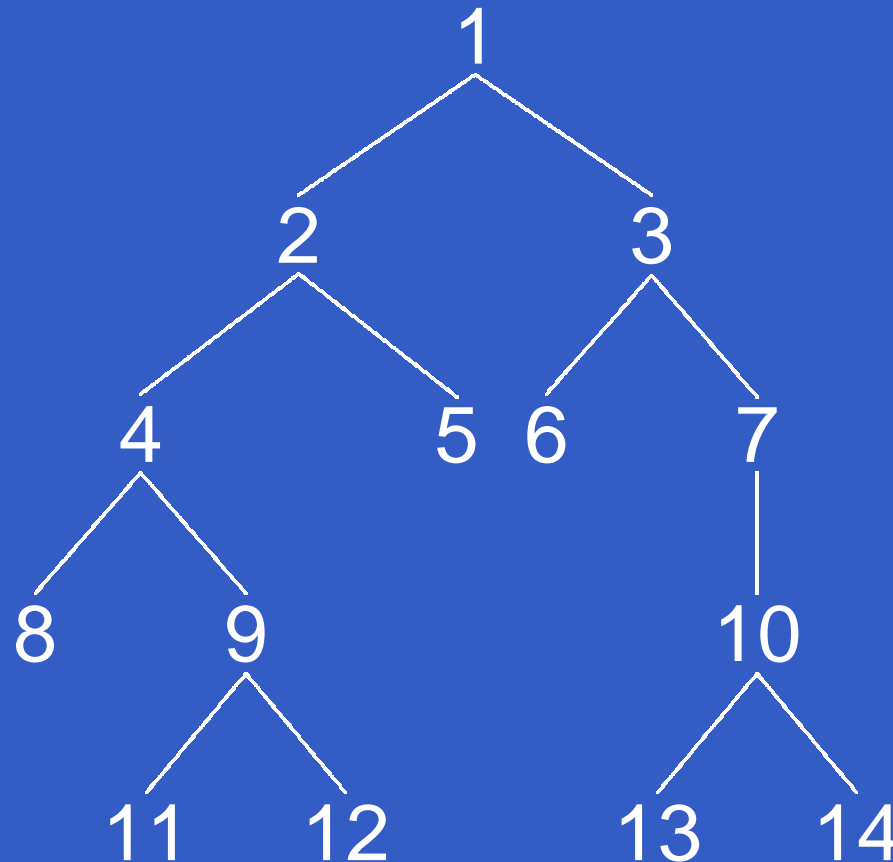
```
  [ ((i,j), eval r (s!(i,j)))
```

```
    | (i,j) <- indices s ]
```

The evaluated sheet is simply the solution to the stated equation. No need to worry about evaluation order. Any caveats?

# Breadth-first Numbering (1)

Consider the problem of numbering a possibly infinitely deep tree in breadth-first order:



# Breadth-first Numbering (2)

The following algorithm is due to G. Jones and J. Gibbons (1992), but the presentation differs.

Consider the following tree type:

```
data Tree a = Empty
            | Node (Tree a) a (Tree a)
```

Define:

$\text{width } t \ i$  The width of a tree  $t$  at level  $i$  (0 origin).

$\text{label } t \ i \ j$  The  $j$ th label at level  $i$  of a tree  $t$  (0 origin).

# Breadth-first Numbering (3)

The following system of equations defines breadth-first numbering:

$$\text{label } t \ 0 \ 0 = 1 \quad (1)$$

$$\text{label } t \ (i + 1) \ 0 = \text{label } t \ i \ 0 + \text{width } t \ i \quad (2)$$

$$\text{label } t \ i \ (j + 1) = \text{label } t \ i \ j + 1 \quad (3)$$

Note that  $\text{label } t \ i \ 0$  is defined for **all** levels  $i$  (as long as the widths of all tree levels are finite).

# Breadth-first Numbering (4)

The code on the next slide sets up the defining system of equations.

- Streams (infinite lists) of labels are used as a mediating data structure to allow equations to be set up between adjacent nodes within levels and between the last node at one level and the first node at the next.
- As there manifestly are no cyclic dependences among the equations, we can entrust the details of solving them to the lazy evaluation machinery, in the safe knowledge that a solution will be found.

# Breadth-first Numbering (5)

```
bfm :: Tree a -> Tree Integer
```

```
bfm t = t'
```

```
where
```

```
(ns, t') = bfmAux (1 : ns) t
```

```
bfmAux :: [Integer] -> Tree a
```

```
-> ([Integer], Tree Integer)
```

```
bfmAux ns Empty = (ns, Empty)
```

```
bfmAux (n : ns) (Node tl _ tr) = ((n + 1) : ns'',  
Node tl' n tr')
```

```
where
```

```
(ns', tl') = bfmAux ns tl
```

```
(ns'', tr') = bfmAux ns' tr
```

# Dynamic Programming

## *Dynamic Programming:*

- Create a **table** of all subproblems that ever will have to be solved.
- Fill in table without regard to whether the solution to that particular subproblem will be needed.
- Combine solutions to form overall solution.

**Lazy Evaluation** is a perfect match as saves us from having to worry about finding a suitable evaluation order.

# The Triangulation Problem (1)

Select a set of **chords** that divides a convex polygon into triangles such that:

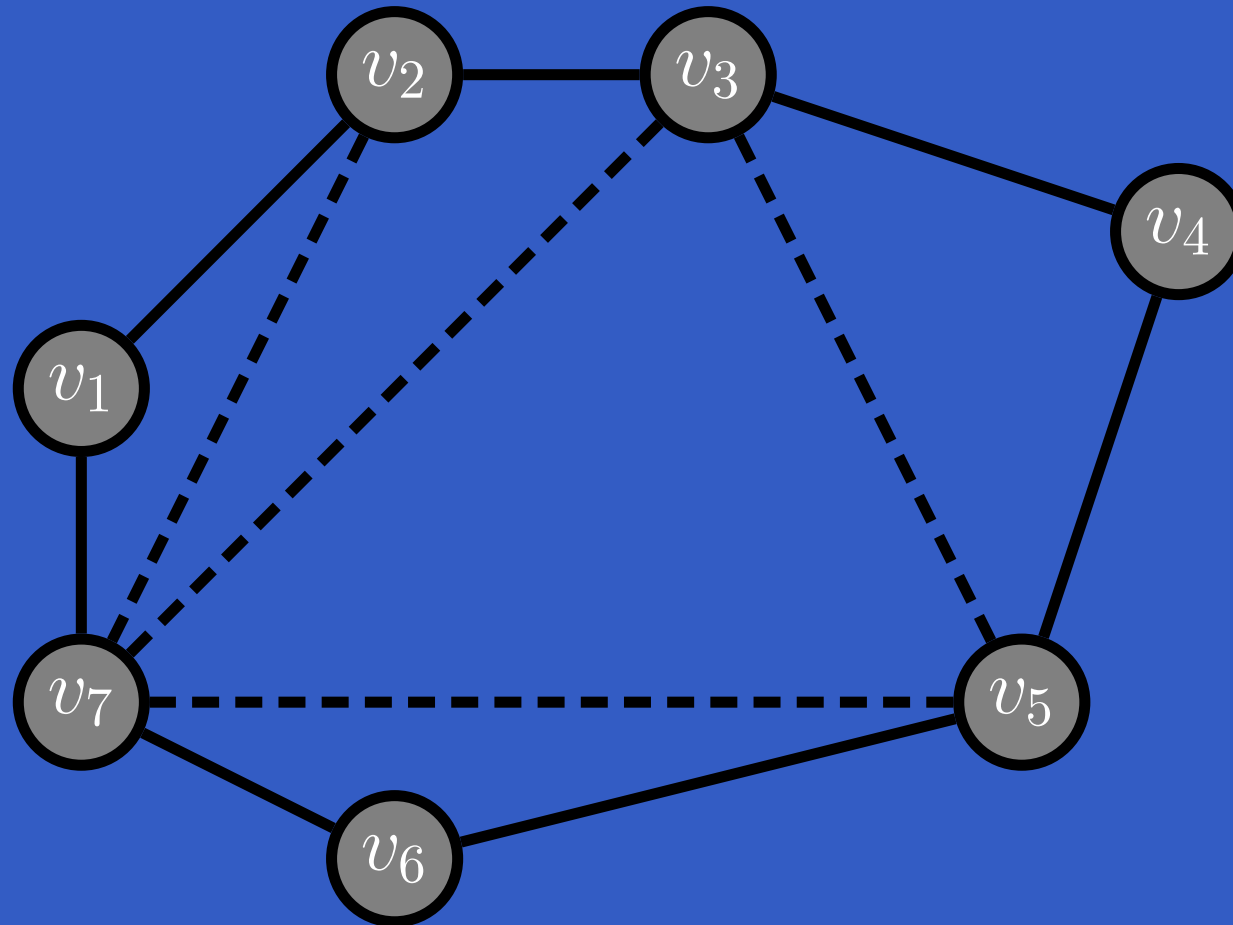
- no two chords cross each other
- the sum of their length is minimal.

We will only consider computing the minimal length.

See Aho, Hopcroft, Ullman (1983) for details.



# The Triangulation Problem (2)



# The Triangulation Problem (3)

- Let  $S_{i,s}$  denote the subproblem of size  $s$  starting at vertex  $v_i$  of finding the minimum triangulation of the polygon  $v_i, v_{i+1}, \dots, v_{i+s-1}$  (counting modulo the number of vertices).
- Subproblems of size less than 4 are trivial.
- Solving  $S_{i,s}$  is done by solving  $S_{i,k+1}$  and  $S_{i+k,s-k}$  for all  $k, 1 \leq k \leq s - 2$
- The obvious recursive formulation results in  $3^{s-4}$  recursive calls.
- But for  $n$  vertices there are only  $n(n - 4)$  non-trivial subproblems!

# The Triangulation Problem (4)

- Let  $C_{is}$  denote the minimal triangulation cost of  $S_{is}$ .
- Let  $D(v_p, v_q)$  denote the length of a chord between  $v_p$  and  $v_q$  (length is 0 for non-chords; i.e. adjacent  $v_p$  and  $v_q$ ).

- For  $s \geq 4$ :

$$C_{is} = \min_{k \in [1, s-2]} \left\{ \begin{array}{l} C_{i, k+1} + C_{i+k, s-k} \\ + D(v_i, v_{i+k}) + D(v_{i+k}, v_{i+s-1}) \end{array} \right\}$$

- For  $s < 4$ ,  $S_{is} = 0$ .

# The Triangulation Problem (5)

These equations can be transliterated straight into Haskell:

```
triCost :: Polygon -> Double
triCost p = cost!(0,n) where
    cost = array ((0,0), (n-1,n))
            ([ ((i,s),
                minimum [ cost!(i, k+1)
                          + cost!((i+k) `mod` n, s-k)
                          + dist p i ((i+k) `mod` n)
                          + dist p ((i+k) `mod` n)
                          ((i+s-1) `mod` n)
                          | k <- [1..s-2] ]))
            | i <- [0..n-1], s <- [4..n] ] ++
            [ ((i,s), 0.0)
            | i <- [0..n-1], s <- [0..3] ] )
n = snd (bounds b) + 1
```

# Reading

- John W. Lloyd. Practical advantages of declarative programming. In *Joint Conference on Declarative Programming, GULP-PRODE'94*, 1994.
- John Hughes. Why Functional Programming Matters. *The Computer Journal*, 32(2):98–197, April 1989.

# Reading

- Geraint Jones and Jeremy Gibbons.  
*Linear-time breadth-first tree algorithms: An exercise in the arithmetic of folds and zips.*  
Technical Report TR-31-92, Oxford University Computing Laboratory, 1992.
- Alfred Aho, John Hopcroft, Jeffrey Ullman.  
*Data Structures and Algorithms.*  
Addison-Wesley, 1983.