

## LiU-FP2010 Part II: Lecture 2

### Lazy Functional Programming

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### Imperative vs. Declarative (1)

- **Imperative Languages:**
  - Implicit state.
  - Computation essentially a sequence of side-effecting actions.
  - Examples: Procedural and OO languages
- **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.

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### Imperative vs. Declarative (2)

Another perspective:

- **Algorithm = Logic + Control**
- Declarative programming emphasises the logic (“what”) rather than the control (“how”).
- Strategy needed for providing the “how”:
  - Resolution (logic programming languages)
  - Lazy evaluation (some functional and logic programming languages)
  - (Lazy) narrowing: (functional logic programming languages)

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### No Control?

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. (E.g. `cut` in Prolog, `seq` in Haskell.)

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### Relinquishing Control

Theme of this lecture: **relinquishing control by exploiting lazy evaluation.**

- Evaluation orders
- Strict vs. Non-strict semantics
- Lazy evaluation
- Applications of lazy evaluation:
  - Programming with infinite structures
  - Circular programming
  - Dynamic programming
  - Attribute grammars

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### Evaluation Orders (1)

Consider:

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

Roughly, any expression that can be evaluated or **reduced** by using the equations as rewrite rules is called a **reducible expression** or **redex**.

Assuming arithmetic, the redexes of the body of `main` are:  $2 + 3$   
 $\text{dbl } (2 + 3)$   
 $\text{sqr } (\text{dbl } (2 + 3))$

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### Evaluation Orders (2)

Thus, in general, many possible reduction orders. Innermost, leftmost redex first is called **Applicative Order Reduction** (AOR). Recall:

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

Starting from `main`:

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

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

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### Evaluation Orders (3)

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**

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### Why Normal Order Reduction? (1)

NOR seems rather inefficient. Any use?

- Best possible termination properties.
- A pure functional languages is just the  $\lambda$ -calculus in disguise. Two central theorems:
  - 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.

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## 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.

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## Exercise 1

Consider:

```
f x = 1
g x = g x
main = f (g 0)
```

Attempt to evaluate `main` using both AOR and NOR. Which order is the more efficient in this case? (Count the number of reduction steps to normal form.)

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## 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:

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

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

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## Strict vs. Non-strict Semantics (2)

Again, consider:

```
f x = 1
g x = g x
```

What is the value of  $f (0/0)$ ? Or of  $f (g 0)$ ?

- AOR:  $f (0/0) \Rightarrow \perp$ ;  $f (g 0) \Rightarrow \perp$   
Conceptually,  $f \perp = \perp$ ; i.e.,  $f$  is strict.
- NOR:  $f (0/0) \Rightarrow 1$ ;  $f (g 0) \Rightarrow 1$   
Conceptually,  $f \circ \perp = 1$ ; i.e.,  $f$  is non-strict.

Thus, NOR results in non-strict semantics.

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## Lazy Evaluation (1)

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

- A redex is evaluated **only if needed**.
- **Sharing** employed to avoid duplicating redexes.
- Once evaluated, a redex is **updated** with the result to avoid evaluating it more than once.

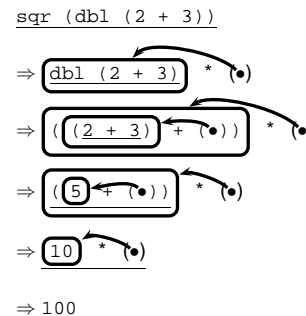
As a result, under lazy evaluation, any one redex is evaluated at most once.

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## Lazy Evaluation (2)

Recall:

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



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## Exercise 2

Evaluate `main` using AOR, NOR, and lazy evaluation:

```
f x y z = x * z
g x      = f (x * x) (x * 2) x
main     = g (1 + 2)
```

(Only consider an applications of an arithmetic operator a redex once the arguments are numbers.)

How many reduction steps in each case?

**Answer:** 7, 8, 6 respectively

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## Infinite Data Structures (1)

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

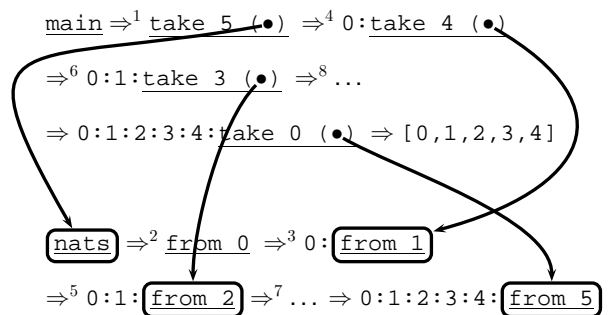
from n = n : from (n+1)

nats = from 0

main = take 5 nats
```

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## Infinite Data Structures (2)



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## Circular Data Structures (2)

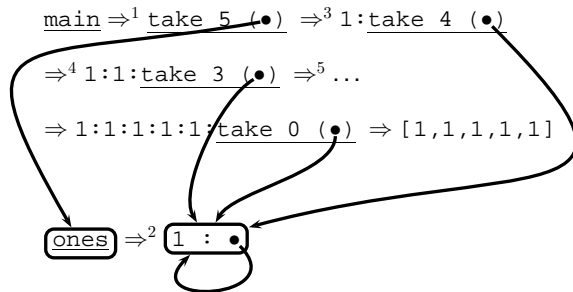
```
take 0 xs = []
take n [] = []
take n (x:xs) = x : take (n-1) xs

ones = 1 : ones

main = take 5 ones
```

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## Circular Data Structures (2)



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## Exercise 3

Given the following tree type

```
data Tree = Empty
          | Node Tree Int Tree
```

define:

- An infinite tree where every node is labelled by 1.
- An infinite tree where every node is labelled by its depth from the root node.

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## Exercise 3: Solution

```
treeOnes = Node treeOnes 1 treeOnes

treeFrom n = Node (treeFrom (n + 1))
                  n
                  (treeFrom (n + 1))

treeDepths = treeFrom 0
```

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## Circular Programming (1)

A non-empty tree type:

```
data Tree = Leaf Int | Node Tree Tree
```

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!**

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## Circular Programming (2)

Write a function that replaces all leaf integers by a given integer, and returns the new tree along with 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
```

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## Circular Programming (3)

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

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

Thus:

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

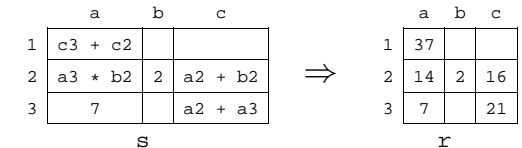
where

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

Intuitively, this works because `fmr` can compute its result without needing to know the **value** of `m`.

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## A Simple Spreadsheet Evaluator



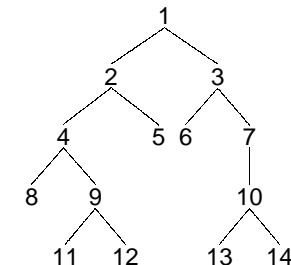
```
r = array (bounds s)
          [ ((i,j), eval r (s!(i,j)))
            | (i,j) <- indices s ]
```

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

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## Breadth-first Numbering (1)

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



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## 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:

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

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## Breadth-first Numbering (3)

The following system of equations defines breadth-first numbering:

$$\begin{aligned} \text{label } t\ 0\ 0 &= 1 & (1) \\ \text{label } t\ (i+1)\ 0 &= \text{label } t\ i\ 0 + \text{width } t\ i & (2) \\ \text{label } t\ i\ (j+1) &= \text{label } t\ i\ j + 1 & (3) \end{aligned}$$

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

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## Breadth-first Numbering (4)

The code that follows 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.
- Idea: the tree numbering function for a subtree takes a stream of labels for the **first node** at each level, and returns a stream of labels for the **node after the last node** at each level.

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## Breadth-first Numbering (5)

- 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.

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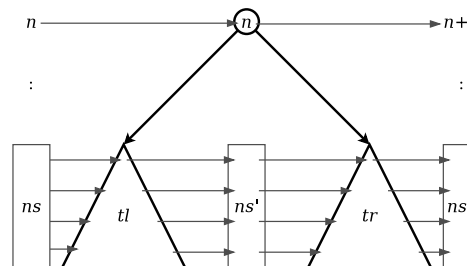
## Breadth-first Numbering (6)

```
bfm :: Tree a -> Tree Integer  Eqns (1) & (2)
bfm t = t'
  where (ns, t') = bfmAux (1 : ns) t

bfmAux :: [Integer] -> Tree a
        -> ([Integer], Tree Integer)  Eqn (3)
bfmAux ns Empty = (ns, Empty)
bfmAux (n : ns) (Node tl _ tr) = ((n + 1) : ns''),
  where (ns', tl') = bfmAux ns tl
        (ns'', tr') = bfmAux ns' tr
```

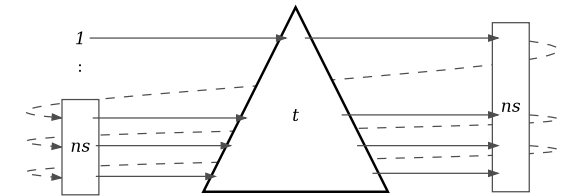
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## Breadth-first Numbering (7)



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## Breadth-first Numbering (8)



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## 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.

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## The Triangulation Problem (1)

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

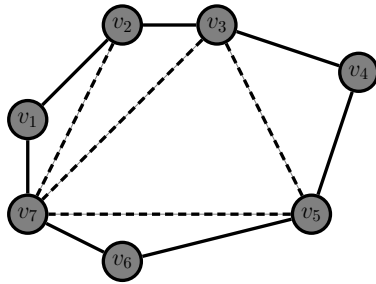
- no two chords cross each other
- the sum of their length is minimal.

We will only consider computing the minimal length.

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

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## The Triangulation Problem (2)



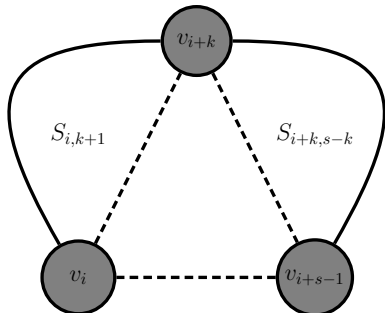
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## The Triangulation Problem (3)

- Let  $S_{is}$  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  $S_{is}$  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  $3^{s-4}$  (non-trivial) calls.
- But for  $n \geq 4$  vertices there are only  $n(n-3)$  non-trivial subproblems!

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## The Triangulation Problem (4)



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## The Triangulation Problem (5)

- 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  $s \geq 4$ :

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

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

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## The Triangulation Problem (6)

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
```

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## Attribute Grammars (1)

Lazy evaluation is also very useful for evaluation of **Attribute Grammars**:

- The attribution function is defined recursively over the tree:
  - takes inherited attributes as extra arguments;
  - returns a tuple of all synthesised attributes.
- As long as there exists **some** possible attribution order, lazy evaluation will take care of the attribute evaluation.

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## Attribute Grammars (2)

- The earlier examples on Circular Programming and Breadth-first Numbering can be seen as instances of this idea.

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## 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.
- Thomas Johnsson. Attribute Grammars as a Functional Programming Paradigm. In *Functional Programming Languages and Computer Architecture, FPCA'87*, 1987

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## 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.

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