# LiU-FP2016: Lecture 5 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


## 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.


## Imperative vs. Declarative (2)

Another perspective:

- Algorithm = Logic + Control


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- Declarative programming emphasises the logic ("what") rather than the control ("how").


## 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)


## No Control?

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

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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.)


## 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


## Evaluation Orders (1)

Consider:

$$
\begin{aligned}
& \operatorname{sqr} x=x * x \\
& \text { dbl } x=x+x \\
& \text { main }=\operatorname{sqr}(d b l(2+3))
\end{aligned}
$$

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: $2+3$
dbl ( $2+3$ ) sqr (dbl $(2+3))$

## Evaluation Orders (2)

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

$$
\begin{aligned}
& \operatorname{sqr} x=x * x \\
& \text { dbl } x=x+x \\
& \operatorname{main}=\operatorname{sqr}(d b l(2+3))
\end{aligned}
$$

Starting from main:

$$
\begin{aligned}
& \underline{\operatorname{main}} \Rightarrow \operatorname{sqr}(\mathrm{dbl}(\underline{2+3})) \Rightarrow \operatorname{sqr}(\underline{\mathrm{dbl} 5}) \\
& \Rightarrow \operatorname{sqr}(\underline{5+5}) \Rightarrow \underline{\operatorname{sqr} 10} \Rightarrow 10 \star 10 \Rightarrow 100
\end{aligned}
$$

This is just Call-By-Value.

## Evaluation Orders (3)

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

$$
\begin{aligned}
& \underline{m a i n} \Rightarrow \underline{\operatorname{sqr}(\mathrm{dbl}(2+3))} \\
& \Rightarrow \mathrm{dbl}(2+3) * \mathrm{dbl}(2+3) \\
& \Rightarrow((\underline{(2+3)+(2+3)) \star \mathrm{dbl}(2+3)} \\
& \Rightarrow(5+(\underline{2+3})) \star \mathrm{dbl}(2+3) \\
& \Rightarrow(5+5) \star \mathrm{dbl}(2+3) \Rightarrow 10 \star \mathrm{dbl}(2+3) \\
& \Rightarrow \cdots \Rightarrow 10 * 10 \Rightarrow 100
\end{aligned}
$$

(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.

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.

## 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.


## Exercise 1

Consider:

$$
\begin{aligned}
& f x=1 \\
& g x=g x \\
& \operatorname{main}=f\left(\begin{array}{ll}
g & 0
\end{array}\right)
\end{aligned}
$$

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.)

## 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:

$$
\begin{aligned}
& (0 / 0)+1=\perp+1=\perp \\
& 1+(0 / 0)=1+\perp=\perp
\end{aligned}
$$

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

Again, consider:

$$
\begin{aligned}
& \mathrm{f} x=1 \\
& \mathrm{~g} x=\mathrm{x} x
\end{aligned}
$$

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

- AOR: $f(\underline{0 / 0}) \Rightarrow \perp ; f(\underline{g} 0) \Rightarrow \perp$

Conceptually, $f \perp=\perp$; i.e., $f$ is strict.

- NOR: $\underline{\mathrm{f}(0 / 0)} \Rightarrow 1 ; \underline{\mathrm{f}(\mathrm{g} \mathrm{0)})} \Rightarrow 1$ Conceptually, foo $\perp=1$; i.e., foo is non-strict.
Thus, NOR results in non-strict semantics.


## Lazy Evaluation (1)

Lazy evaluation is a technique for implementing NOR more efficiently:

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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.


## Lazy Evaluation (2)

Recall:
$\underline{\operatorname{sqr}(d . b l(2+3))}$
sir $x=x * x$
$\operatorname{dbl} \mathrm{x}=\mathrm{x}+\mathrm{x}$
main $=$
sqr (dbl $(2+3))$

## Lazy Evaluation (2)

Recall:
sqr $x=x * x$ $\operatorname{dbl} \mathrm{x}=\mathrm{x}+\mathrm{x}$ main =
sqr (d.bl $(2+3))$

## Lazy Evaluation (2)

Recall:
sqr $\mathrm{X}=\mathrm{x} * \mathrm{x}$
$\operatorname{dbl} x=x+x$
main $=$
$\operatorname{sqr}(\operatorname{dbl}(2+3)) \Rightarrow((\underline{2+3})+(0)) *(0)$

## Lazy Evaluation (2)

Recall:
sqr $\mathrm{X}=\mathrm{x} * \mathrm{x}$
$\operatorname{dbl} \mathrm{x}=\mathrm{x}+\mathrm{x}$
main $=$
$\operatorname{sqr}(\operatorname{d.ol}(2+3)) \Rightarrow(\underline{(2+3})+(0))$
$\Rightarrow\left(5+7^{\circ}\right)^{-1}$
sqr (dbl $(2+3))$
$\Rightarrow \mathrm{dbl}(2+3)$ * (o)
$\Rightarrow((2+3)+(0)) *$ (0)

## Lazy Evaluation (2)

Recall:
sqr $\mathrm{x}=\mathrm{x} * \mathrm{x}$
$\operatorname{dbl} \mathrm{x}=\mathrm{x}+\mathrm{x}$
main =
$\operatorname{sqr}(\mathrm{dbl}(2+3)) \Rightarrow((\underline{2+3})+(0))$
$\Rightarrow(5)+(0)) *(0)$
$\Rightarrow 10{ }^{*}(0)$
$\underline{\text { sqr }(d b l(2+3))}$
$\Rightarrow \mathrm{dbl}(2+3)$ * (०)
$\Rightarrow((\underline{2+3})+(0)) *$ (०)

## Lazy Evaluation (2)

Recall:
sqr $x=x * x$
$\operatorname{dbl} x=x+x$
main $=$
$\operatorname{sqr}(\mathrm{dbl}(2+3)) \Rightarrow((\underline{2+3})+(0))$

$\Rightarrow 10{ }^{*}(0)$
$\Rightarrow 100$

## Exercise 2

Evaluate main using AOR, NOR, and lazy evaluation:

$$
\begin{aligned}
f x y z & =x * z \\
g x & =f(x * x)(x * 2) x \\
\operatorname{main} & =g(1+2)
\end{aligned}
$$

(Only consider an applications of an arithmetic operator a redex once the arguments are numbers.)
How many reduction steps in each case?

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Evaluate main using AOR, NOR, and lazy evaluation:

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& g x=f(x * x)(x * 2) x \\
& \text { main }=g(1+2)
\end{aligned}
$$

(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

## Infinite Data Structures (1)

take 0 xs $=[]$
take n [] $=[]$
take $n(x: x s)=x$ : take $(n-1)$ xs
from $n=n$ : from ( $\mathrm{n}+1$ )
hats $=$ from 0
main = take 5 mats

## Infinite Data Structures (2)

main
nats

## Infinite Data Structures (2)

## $\underline{\text { main }} \Rightarrow^{1}$ take 5 (0)

 nats
## Infinite Data Structures (2)

## $\underline{\text { main }} \Rightarrow^{1}$ take 5 (0)

$$
\text { nats } \Rightarrow^{2} \text { from } 0
$$

Infinite Data Structures (2)


Infinite Data Structures (2)


Infinite Data Structures (2)


## Infinite Data Structures (2)

$$
\text { main } \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{4} 0 \text { :take } 4(0)
$$

$$
\Rightarrow^{6} 0: 1: \text { take } 3
$$

## Infinite Data Structures (2)

$$
\text { main } \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{4} 0 \text { :take } 4(0)
$$

$$
\Rightarrow^{6} 0: 1: \text { take } 3
$$

## Infinite Data Structures (2)

$$
\text { main } \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{4} 0 \text { :take } 4(0)
$$

$$
\Rightarrow^{6} 0: 1: \underline{\text { take } 3}(0) \Rightarrow^{8} \ldots
$$

## Infinite Data Structures (2)

$$
\underline{\text { main }} \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{4} 0 \text { :take } 4(0)
$$

$$
\Rightarrow^{6} 0: 1: \underline{\text { take } 3}(0) \Rightarrow^{8} \ldots
$$

$$
\text { nets } \Rightarrow^{2} \text { from } 0 \Rightarrow^{3} 0: \text { from } 1
$$

$$
\Rightarrow^{5} 0: 1: \text { from 2 } \Rightarrow^{7} \ldots \Rightarrow 0: 1: 2: 3: 4: \text { from 5 }
$$

## Infinite Data Structures (2)

$$
\begin{aligned}
& \text { main } \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{4} 0 \text { :take } 4(0) \\
& \begin{array}{l}
\Rightarrow^{6} 0: 1: \text { take } 3(0) \Rightarrow^{8} \cdots \\
\Rightarrow 0: 1: 2: 3: 4: \text { ike } 0 \text { (o) }
\end{array} \\
& \begin{array}{l}
\text { nuts } \Rightarrow^{2} \text { from } 0 \Rightarrow{ }^{3} 0: \text { from 1 } \\
\Rightarrow^{5} 0: 1: \text { from } 2
\end{array} \Rightarrow^{7} \ldots \Rightarrow 0: 1: 2: 3: 4: \text { from } 5
\end{aligned}
$$

## Infinite Data Structures (2)

$$
\begin{aligned}
& \text { main } \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{4} 0: \text { take 4 (o) } \\
& \Rightarrow^{6} 0: 1: \text { take } 3(0) \Rightarrow^{8} \cdots \\
& \Rightarrow 0: 1: 2: 3: 4: \text { ane } 0 \quad(0) \Rightarrow[0,1,2,3,4] \\
& \Rightarrow^{5} 0: 1: \text { from } 2
\end{aligned} \Rightarrow^{7} \ldots \Rightarrow 0: 1: 2: 3: 4: \text { from 5 } 5
$$

## Circular Data Structures (2)

take 0 xs $=[]$
take n [] $=[]$
take $n(x: x s)=x$ : take $(n-1)$ xs
ones = 1 : ones
main $=$ take 5 ones

## Circular Data Structures (2)

## main

## ones

## Circular Data Structures (2)

## $\underline{\text { main }} \Rightarrow^{1}$ take $5(0)$

ones

## Circular Data Structures (2)

## main $\Rightarrow^{1}$ take $5(0)$

## Circular Data Structures (2)

$$
\underline{\text { main }} \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{3} 1: \text { take } 4(0)
$$

## Circular Data Structures (2)

$$
\underline{\text { main }} \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{3} 1: \text { take } 4(0)
$$

$$
\Rightarrow^{4} 1
$$

## Circular Data Structures (2)

$$
\underline{\text { main }} \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{3} 1: \text { take } 4(0)
$$

$$
\begin{aligned}
& \Rightarrow^{4} 1: 1: \text { take } 3(0) \Rightarrow^{5} \ldots \\
& \text { ones } \Rightarrow^{2}: 0 .
\end{aligned}
$$

## Circular Data Structures (2)

$$
\underline{\text { main }} \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{3} 1: \text { take } 4(0)
$$

$$
\begin{aligned}
& \Rightarrow^{4} 1: 1: \text { take } 3(0) \Rightarrow^{5} \ldots \\
& \Rightarrow 1: 1: 1: 1: 1 \cdot \text { take } 0(0)
\end{aligned}
$$

## Circular Data Structures (2)

$$
\begin{aligned}
& \frac{\operatorname{main}}{} \Rightarrow^{1} \text { take } 5(0) \Rightarrow^{3} 1: \text { take } 4(0) \\
& \Rightarrow^{4} 1: 1: \underline{\text { take } 3}(0) \Rightarrow^{5} \ldots \\
& \Rightarrow 1: 1: 1: 1: 1 \cdot \text { take } 0(0) \Rightarrow[1,1,1,1,1]
\end{aligned}
$$

## Exercise 3

Given the following tree type

$$
\begin{aligned}
\text { data Tree } & =\text { Empty } \\
& \text { | Node Tree Int Tree }
\end{aligned}
$$

define:

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


## Exercise 3: Solution

treeOnes $=$ Node treeOnes 1 treeOnes
treeFrom $n=$ Node (treeFrom $(n+1)$ ) n
(treeFrom (n + 1))
treeDepths $=$ treeFrom 0

## 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.

## Circular Programming (1)

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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?

## Circular Programming (1)

A non-empty tree type:

$$
\text { data Tree }=\text { Leaf Int | Node Tree Tree }
$$

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!

## 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:

$$
\begin{aligned}
& \text { fmr : Int }->\text { Tree }->\text { (Tree, Int) } \\
& \text { fmr } m \text { (Leaf } i)=\text { (Ieaf } m, i) \\
& \text { fmr m (Node tl tr) }= \\
& \text { (Node tl' tr', min ml mr) } \\
& \text { where } \\
& \left(t l^{\prime}, m l\right)=\text { fmr } m t l \\
& \left(t r^{\prime}, m r\right)=
\end{aligned}
$$

## Circular Programming (3)

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

$$
\left(t^{\prime}, m\right)=f m r m t
$$

Thus:

$$
\begin{aligned}
& \text { findMinReplace } t=t ' \\
& \text { where }
\end{aligned}
$$

$$
\left(t^{\prime}, m\right)=f m r m t
$$

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

## A Simple Spreadsheet Evaluator (1)

|  | a | b | C |
| :---: | :---: | :---: | :---: |
| 1 | $c 3+c 2$ |  |  |
| 2 | a3 * b2 | 2 | $\mathrm{a} 2+\mathrm{b} 2$ |
| 3 | 7 |  | $a 2+a 3$ |



$$
\begin{aligned}
s^{\prime}=\text { array } & \text { (bounds } s \text { ) } \\
& {\left[\left(r, \text { evalCell } s^{\prime} \quad(s \quad r)\right)\right.} \\
& \mid r<- \text { indices } s]
\end{aligned}
$$

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

## A Simple Spreadsheet Evaluator (2)

As it is quite instructive, let us develop this evaluator together. Some definitions to get us started:
type CellRef = (Char, Int)
type Sheet a = Array CellRef a
data BinOp = Add | Sub | Mul | Div
data Exp = Lit Double
| Ref CellRef
| App BinOp Exp Exp

## 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:
width $t i \quad$ 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).

## Breadth-first Numbering (3)

The following system of equations defines breadth-first numbering:

$$
\begin{align*}
\text { label } t 00 & =1  \tag{1}\\
\text { label } t(i+1) 0 & =\text { label } t i 0+\text { width } t i  \tag{2}\\
\text { label } t i(j+1) & =\text { label } t i j+1 \tag{3}
\end{align*}
$$

Note that label ti0 is defined for all levels $i$ (as long as the widths of all tree levels are finite).

## Breadth-first Numbering (4)

The code that follows sets up the defining system of equations:

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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.


## 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.


## 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.


## Breadth-first Numbering (6)


bfnAux :: [Integer] -> Tree a
-> ([Integer], Tree Integer)
bfnAux ns Empty $=$ (ns, Empty)

where

$$
\begin{aligned}
& \text { (ns', tl') }=\text { bfnAux ns tl } \\
& \text { (ns'r }{ }^{\prime} \text { tr') }=\text { bfnAux } n s^{\prime} \text { tr }
\end{aligned}
$$

## Breadth-first Numbering (7)



Breadth-first Numbering (8)


## 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 $v_{i}, v_{i+1}, \ldots, v_{i+s-1}$ (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 $3^{s-4}$ (non-trivial) calls.
- But for $n \geq 4$ vertices there are only $n(n-3)$ non-trivial subproblems!


## The Triangulation Problem (4)



## The Triangulation Problem (5)

- Let $C_{i s}$ denote the minimal triangulation cost of $S_{i s}$.
- Let $D\left(v_{p}, v_{q}\right)$ 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 $s \geq 4$ :

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

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


## 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
```


## 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.


## Attribute Grammars (2)

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

## 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
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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.

