

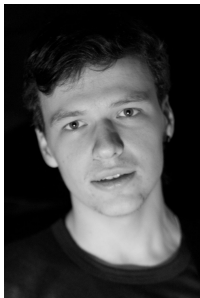
Advanced FP

Functional Data Structures

**Daniil Berezun**

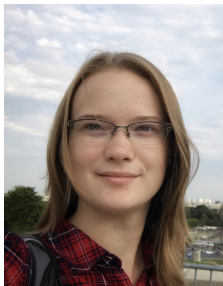
2023

# Us; Contacts



Daniil Berezun

- telegram: @DaniilBerezun
- danya.berezun@gmail.com



Ekaterina Verbitskaia

- telegram: @kajigor
- kajigor@gmail.com

JetBrains Research, Programming Languages and Tools Lab

Telegram: Advanced Functional Programming, CUB fall 2023



# Syllabus

- Functional data structures, persistency (C. Okasaki):
  - RB-trees in functional setting
  - Queues in functional setting:
    - Pure Functional Queue
    - Banker's queue
    - Physicist's queue
    - Real-time Queue
- Quality assurance; Property-based testing
- Lazy programming; Profiling and debugging of Haskell programs
- Functional concepts and pearls (R. Bird):
  - Smallest missing number
  - Remove repeats
  - Burrows-Wheeler transform
  - All prefixes
  - Knuth-Morris-Pratt algorithm
  - Puzzles: rush hour and sudoku
  - Hylomorphisms and nexyses
  - Code with no cycles
- Effect systems; Using them for logging, error catching, mutability
- Zippers, type algebra, comonads, and Pearl: Scrap Your Zippers
- New Pearls: More Fixpoints!, monoids and 'vector reverse'
- Functional Design Patterns. Design and implementation of DSLs:
  - GADTs
  - Existential type
  - Rank N types
  - DSL
- Type level programming in Haskell and beyond. Type families, tagless-final
- Template Haskell, Lens
- Interfacing with the real world: working with databases, IORef, concurrency



# Today: Reminder about Persistency

## Main today's concepts

- *Immutable* data structures
- *Persistent* data structures

## Remarks

- We can use *old* nodes (*share*) in new version of the data structure
- Non-persistent data structures are called *ephemeral*

## Important remark

During **this lecture** we **do not** assume our language to be **lazy**



## Definition (Linked List)

Who knows?

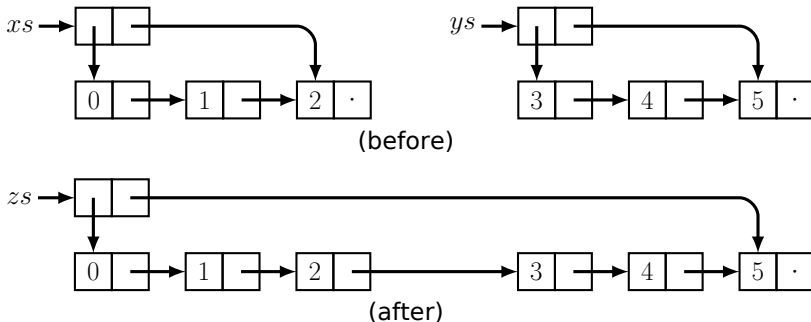
## Definition (Linked List)

Who knows?

## Definition (List) [One of possible definitions]

A data structure such that from some predefined side (for example, list head) deletion and insertion of element has complexity  $O(1)$

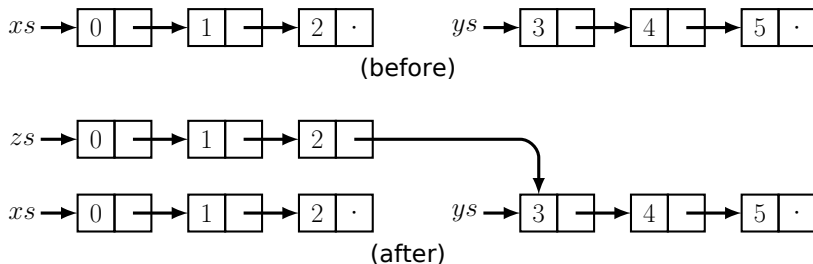
# List Concatenation in the Imperative Paradigm



Concatenation of lists *xs* and *ys* in the imperative paradigm

- Destroys argument lists *xs* and *ys* (one can't use them further)
- Complexity:  $O(1)$

# Pure Functional Lists Concatenation



Execution of  $zs = xs ++ ys$  in functional world

- $xs$  and  $ys$  remain intact
- we copied **a lot** but the first list only
  - i.e. *persistency* through copying (memory)
  - *shared* parts



## Pure Functional Lists Concatenation --- 2

### How to implement concatenation ++ of lists xs and ys?

- If `xs` is empty then `ys` is the answer
- Otherwise `xs` consists of `h` as a head and `tl` as a tail then the answer is a list with head `h` and tail `tl ++ ys`.

Complexity:  $O(\text{length}(xs))$

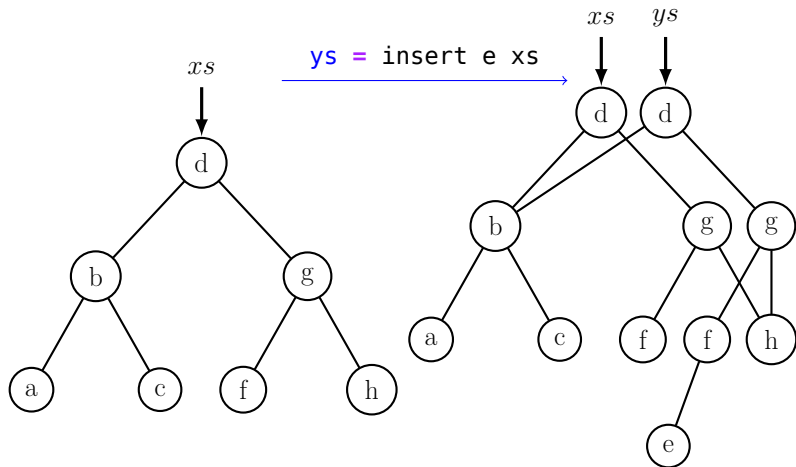
```
1  (++) []      ys = ys
2  (++) (h:tl) ys = h : (tl ++ ys)
```

### How to update the n-th list element?

```
1  update []    i y = error "i is greater than list length"
2  update x:xs  0 y = y : xs
3  update x:xs  i y = x : update xs (i-1) y
```

- $O(n)$ ... **very sad** ;(
- We copy the element being modified **and** all elements that have direct or indirect pointers to it

## Another example: Trees



➤ Usually, the number of nodes to be copied is at most  $\log_2 n$

# On Concatenation Associativity

In theory list concatenation is associative

$$(((a_1 \mathbin{++} a_2) \mathbin{++} a_3) \mathbin{++} \dots \mathbin{++} a_n) \equiv (a_1 \mathbin{++} (a_2 \mathbin{++} (a_3 \mathbin{++} (\dots \mathbin{++} a_n))))$$

In practise left-hand side is much slower than right-hand side

## Note for developers

Sometimes, for an efficient implementation one needs to redesign algorithms in a way such that shorter lists are concatenated with longer lists; Ideally, always concatenate one element with a list

# Pearl: RB-Trees: definition

```
1 data Colour = R | B
2 data Tree a = E | T Colour (Tree a) a (Tree a)
```

## Invariants

- 0  $T \_ a \times b \Rightarrow \forall i \in a, j \in b. i < x \leq j$
- 1 No red node has a red parent
- 2 Every path from the root to an empty node contains the same number of black nodes

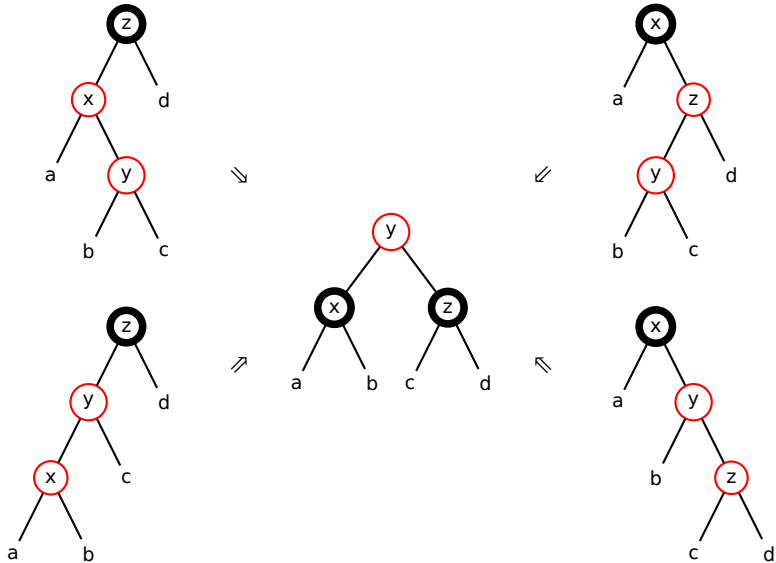
## Simple set operations

```
1 empty = E -- we assume them to be black
2
3 member x E = False
4 member x (T _ a y b)
5   | x < y    = member x a
6   | x > y    = member x b
7   | otherwise = True
```

# Pearl: RB-Trees: Insertions

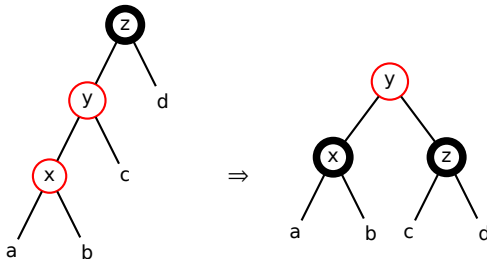
```
1 insert :: Ord a => a -> Tree a -> Tree a
2 insert x s = makeBlack (ins s) -- root is always black
3 where
4   ins E = T R E x E -- new node is red
5   ins s@(T colour a y b)
6     | x < y      = balance colour (ins a) y b
7     | x > y      = balance colour a      y (ins b)
8     | otherwise = s
9   makeBlack (T _ a y b) = T B a y b
```

# Pearl: RB-Trees: Possible Invariant Violations



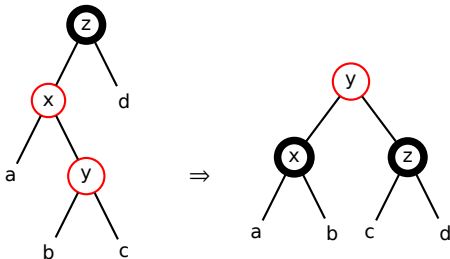
## RB-Trees: Balance (1/4)

```
1 balance :: Colour -> Tree a -> a -> Tree a -> Tree a
2 balance B (T R (T R a x b) y c) z d
3 = T R (T B a x b) y (T B c z d)
```



## RB-Trees: Balance (2/4)

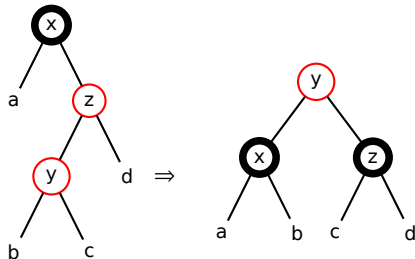
```
1 balance :: Colour -> Tree a -> a -> Tree a -> Tree a
2 balance B (T R a x (T R b y c)) z d
3 = T R (T B a x b) y (T B c z d)
```





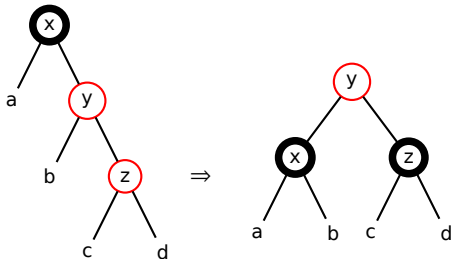
## RB-Trees: Balance (3/4)

```
1 balance :: Colour -> Tree a -> a -> Tree a -> Tree a
2 balance B a x (T R (T R b y c) z d)
3 = T R (T B a x b) y (T B c z d)
```



## RB-Trees: Balance (4/4)

```
1 balance :: Colour -> Tree a -> a -> Tree a -> Tree a
2 balance B a x (T R b y (T R c z d))
3 = T R (T B a x b) y (T B c z d)
```



## Pearl: RB-Trees: All together

```
1 balance :: Colour -> Tree a -> a -> Tree a -> Tree a
2 balance B (T R (T R a x b) y c) z d
3     = T R (T B a x b) y (T B c z d)
4 balance B (T R a x (T R b y c)) z d
5     = T R (T B a x b) y (T B c z d)
6 balance B a x (T R (T R b y c) z d)
7     = T R (T B a x b) y (T B c z d)
8 balance B a x (T R b y (T R c z d))
9     = T R (T B a x b) y (T B c z d)
10 balance c t1 a t2 = T c t1 a t2
```

# Pearl: RB-Trees: All together

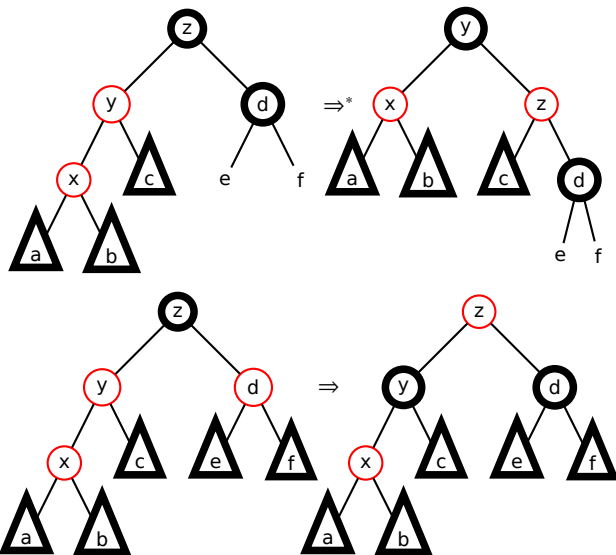
```
1 balance :: Colour -> Tree a -> a -> Tree a -> Tree a
2 balance B (T R (T R a x b) y c) z d
3     = T R (T B a x b) y (T B c z d)
4 balance B (T R a x (T R b y c)) z d
5     = T R (T B a x b) y (T B c z d)
6 balance B a x (T R (T R b y c) z d)
7     = T R (T B a x b) y (T B c z d)
8 balance B a x (T R b y (T R c z d))
9     = T R (T B a x b) y (T B c z d)
10 balance c t1 a t2 = T c t1 a t2
```

```
1 balance :: Colour -> Tree a -> a -> Tree a -> Tree a
2 balance B (T R (T R a x b) y c) z d
3     || B (T R a x (T R b y c)) z d
4     || B a x (T R (T R b y c) z d)
5     || B a x (T R b y (T R c z d))
6     = T R (T B a x b) y (T B c z d)
7 balance c t1 a t2 = T c t1 a t2
```

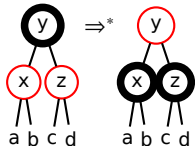
# Pearl: RB-Trees: What happened to all the mess?

In imperative world we usually consider more cases (like uncle's colour) trying to *minimize assigns*

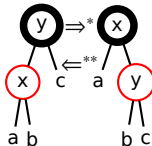
While in FP we *construct new tree anyway*



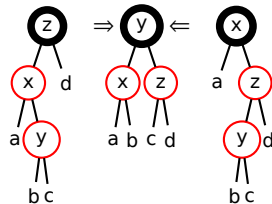
# Pearl: RB-Trees: Alternative imperative-like



Colour flip;  
\* one of  $a - d$  red



Single Rotation;  
\*  $a$  red; \*\*  $c$  red



Double Rotation

```

1  -- color flips
2  balance B (T R a@(T R _ _ _ ) x b) y (T R c z d)
3  || B (T R a x b@(T R _ _ _ )) y (T R c z d)
4  || B (T R a x b) y (T R c@(T R _ _ _ ) z d)
5  || B (T R a x b) y (T R c z d@(T R _ _ _ ))
6  = T R (T B a x b) y (T B c z d)
7  -- single rotations
8  balance B (T R a@(T R _ _ _ ) x b) y c = T B a x (T R b y c)
9  balance B a x (T R b y c@(T R _ _ _ )) = T B (T R a x b) y c
10 -- double rotations
11 balance B (T R a x (T R b y c)) z d
12 || B a x (T R (T R b y c) z d)
13 = T B (T R a x b) y (T R c z d)
14 -- no balancing necessary
15 balance color a x b = T color a x b
    
```

## Next for Today

- Reminder about *amortized* methods
- Persistent + amortization + laziness on FIFO queue example

# On Amortized Time Analysis

Standard complexity notation  $O(\cdot)$  -- worst case estimation

But actually, we may have more freedom:

- Let's perform  $n + 1$  action
- Most of actions will be ``cheap":  $O(1)$
- One ``expensive" action: for example,  $O(n)$
- Standard asymptotic complexity:  $O(n)$
- Average complexity of performing  $n$  actions (*amortized time complexity*) can be  $O(1)$  for an action

$$a = \frac{\sum_{i=1}^n t_i}{n}$$

This additional freedom degree sometimes allows a simpler and more efficient implementation to be designed



## Definition (*Accumulated Savings*)

A difference between total current amortized cost and total current fair value

- **NB:** accumulated savings must be **non-negative**
- I.e. ``expensive'' operations may take place iff accumulated savings are enough to cover their additional cost

$$a_i = t_i + c_i - \bar{c}_i$$

where  $t_i$  --- fair cost,  $c_i$  --- credit amount allocated by action  $i$ ,  
 $\bar{c}_i$  --- amount of credit spent by action  $i$

- Each credit unit must be allocated before being spent
- Credit cannot be used twice
- $\sum c_i \geq \sum \bar{c}_i \Rightarrow \sum a_i \geq \sum t_i$
- Amortized complexity is  $n * O(f(n, m)) \Leftrightarrow \forall n. a_i = O(f(n, m))$   
 $\Rightarrow a = \frac{\sum_{i=1}^n a_i}{n} = \frac{n * O(f(n, m))}{n} = O(f(n, m))$

$$\Phi : \text{Object}(d) \rightarrow \text{Potential}$$

Usually, initial potential is zero and is always non-negative

$$a_i = t_i + \Phi(d_i) - \Phi(d_{i-1})$$

$$\begin{aligned}\sum_{i=1}^j t_i &= \sum_{i=1}^j (a_i + \Phi(d_{i-1}) - \Phi(d_i)) \\ &= \sum_{i=1}^j a_i + \sum_{i=1}^j (\Phi(d_{i-1}) - \Phi(d_i)) \\ &= \sum_{i=1}^j a_i + \Phi(d_0) - \Phi(d_j)\end{aligned}$$

TODO: diff between methods

Interface:

- `empty`: `queue -> bool`
- `enqueue`: `queue * int`  
    `-> queue`
- `head`: `queue -> int`
- `tail`: `queue -> queue`

# Pure Functional Queues

Interface:

- > `empty`: queue -> bool
- > `enqueue`: queue \* int  
-> queue
- > `head`: queue -> int
- > `tail`: queue -> queue

## Simplest implementation

Via a pair of lists, `f` and `r`

- > `f` (front) contains the head elements of the queue in the initial (correct) order,
- > `r` (reversed) consist of tail elements in reverse order

For example, queue  
=`[1;2;3;4;5;6]` can be  
represented as two lists  
`f`=`[1;2;3]` and `r`=`[6;5;4]`

Question: When to move elements from the front to reversed list?

## Definition (Queue Invariant)

List  $f$  may become empty iff list  $r$  is also empty (i.e., the queue is empty)

```
1 enqueue ([], _) x = ([x], []) ---  $O(1)$ 
2 enqueue (f, r) x = (f, x:r) ---  $O(1)$ 
3 tail ([x], r) = (rev r, []) ---  $O(n); O^*(1)$ 
4 tail (x:f, r) = (f, r) ---  $O(1)$ 
```

Thus, `head` and `enqueue` are  $O(1)$  instead of  $O(n)$ ,  
but `tail` is still  $O(n)$

## Definition (Invariant)

Each element in the *tail* list is associated with one credit unit

- > Each **enqueue** call performs the only real computational step and emits *additional* credit unit for an element in the tail list  
amortized complexity is 2
- > **tail**, if no list inversion happens, performs one step and spends no credit units  
amortized complexity is 1
- > **tail**, if list reverse happens, performs  $(m + 1)$  steps, where  $m$  is a tail list length, and spends  $m$  credit units  
amortized complexity is  $m + 1 - m = 1$

## Definition (Invariant)

Each element in the *tail* list is associated with one credit unit

- > Each **enqueue** call performs the only real computational step and emits *additional* credit unit for an element in the tail list  
amortized complexity is 2
- > **tail**, if no list inversion happens, performs one step and spends no credit units  
amortized complexity is 1
- > **tail**, if list reverse happens, performs  $(m + 1)$  steps, where  $m$  is a tail list length, and spends  $m$  credit units  
amortized complexity is  $m + 1 - m = 1$

*Physicist's method*: just let  $\Phi$  be rear list length

## Conclusion

- In case of purely functional queue, function `tail` worst case complexity is  $O(n)$  and amortized ---  $O(1)$
- Good if one do not need persistency, and amortized performance is good enough for the problem
- **Bad news:** we have implicitly assumed that we use queues *ephemerally* i.e. in *persistent setting* still  $O(n)$
- (Next) Lazy evaluations + amortized compexity = persistent queues with a very good amortized complexity



## Lazy Evaluations

Delays the evaluation of an expression until its value is needed  
(*non-strict evaluation*)

## ML

```
1  type  $\alpha$  sup
2  val delay : (unit ->  $\alpha$ ) ->  $\alpha$  sup
3  val force :  $\alpha$  sup ->  $\alpha$ 
```

We will use \$ denote that evaluation is suspended

# Lazy Evaluations

## Lazy Evaluations

Delays the evaluation of an expression until its value is needed  
(*non-strict evaluation*)

## ML

```
1  type  $\alpha$  sup
2  val delay : (unit ->  $\alpha$ ) ->  $\alpha$  sup
3  val force :  $\alpha$  sup ->  $\alpha$ 
```

We will use \$ denote that evaluation is suspended

## Memoization of lazy evaluations

Ones the value of expression is needed, evaluate it and *memoize* (remember, *sharing*) the result; If it will be needed further, just return the memoized result

## Definition (Stream)

is a list but evaluations of sublists are delayed

Example: Stream of all possible natural numbers

# Lazy Lists (Streams)

## Definition (Stream)

is a list but evaluations of sublists are delayed

Example: Stream of all possible natural numbers

## Notation

Add an element  $x$  to the tail  $xs$ :  $\$Cons\ x\ xs$

Empty stream:  $\$Nil$

Delay  $f$ :  $\$f$

## Remark

Stream may be both finite and infinite;  
One never knows until the end appears

# Streams in Standard ML

```
1 datatype  $\alpha$  StreamCell = Nil | Cons of  $\alpha \times \alpha$  Stream
2 withtype  $\alpha$  Stream =  $\alpha$  StreamCell susp
3
4 (Stream example *)
5 $(Cons(1, $Cons(2, $Cons(3, $Nil))))
6
7 (monolithic append aka. suspended list *)
8 fun s ++ t = $(force s @ force t)
9
10 (incremental append aka. stream *)
11 fun s ++ t = $case s of
12     $Nil => force t
13     | $Cons (x, s') => Cons (x, s' ++ t)
```

## Example: Fibonacci Numbers

Consider function  $\text{zip} : \text{stream} \times \text{stream} \rightarrow \text{stream}$ , which sums streams element by element

A stream of Fibonacci numbers:

$$\text{fibs} \equiv \$\text{Cons}(1, \$\text{Cons}(1, \text{zip}(\text{fibs}, \text{tail}(\text{fibs}))))$$

## Example: Fibonacci Numbers

Consider function  $\text{zip} : \text{stream} \times \text{stream} \rightarrow \text{stream}$ , which sums streams element by element

A stream of Fibonacci numbers:

$\text{fibs} \equiv \$\text{Cons}(1, \$\text{Cons}(1, \text{zip}(\text{fibs}, \text{tail}(\text{fibs}))))$

1	1			

## Example: Fibonacci Numbers

Consider function  $\text{zip} : \text{stream} \times \text{stream} \rightarrow \text{stream}$ , which sums streams element by element

A stream of Fibonacci numbers:

$\text{fibs} \equiv \$\text{Cons}(1, \$\text{Cons}(1, \text{zip}(\text{fibs}, \text{tail}(\text{fibs}))))$

1	1			
1				



## Example: Fibonacci Numbers

Consider function  $\text{zip} : \text{stream} \times \text{stream} \rightarrow \text{stream}$ , which sums streams element by element

A stream of Fibonacci numbers:

$\text{fibs} \equiv \$\text{Cons}(1, \$\text{Cons}(1, \text{zip}(\text{fibs}, \text{tail}(\text{fibs}))))$

1	1	2		
1				

## Example: Fibonacci Numbers

Consider function  $\text{zip} : \text{stream} \times \text{stream} \rightarrow \text{stream}$ , which sums streams element by element

A stream of Fibonacci numbers:

$\text{fibs} \equiv \$\text{Cons}(1, \$\text{Cons}(1, \text{zip}(\text{fibs}, \text{tail}(\text{fibs}))))$

1	1	2		
1	2			

## Example: Fibonacci Numbers

Consider function  $\text{zip} : \text{stream} \times \text{stream} \rightarrow \text{stream}$ , which sums streams element by element

A stream of Fibonacci numbers:

$\text{fibs} \equiv \$\text{Cons}(1, \$\text{Cons}(1, \text{zip}(\text{fibs}, \text{tail}(\text{fibs}))))$

1	1	2	3	
1	2			

and so on

# Persistent Banker's Queue

## Remark

This implementation has amortized complexity  $O(1)$  and is persistent

- 1 Use streams instead of lists
- 2 Store stream lengths explicitly
- 3 Invariant:  $|f| > |r|$

If streams  $f$  and  $r$  have the same length, define  $f$  as  $f \# reverse(r)$

## Why rotate when $|f| \approx |r|$ ?

- > *reverse* is monolithic, hence, it needs  $|r|$  steps
- > suspended *reverse* can be forced only after  $|f|$  calls of *tail*
- > *debits* instead credits:

each *tail* call discharges one debit

## Why is it persistent?

- > Lazy (suspended) evaluation  $\Rightarrow$  delayed until needed
- > Memoization  $\Rightarrow$  each suspension computes only ones

# Persistent Physicist's Queue

- $\Psi$  maps object with potential representing an upper bound on the accumulated debt  $\Psi = \min(2|W|, |F| - |R|)$
- We need only *monolithic suspensions*:
  - use suspended list instead of stream
- Front divided into two parts:
  - strict non-empty prefix
  - suspended front itself --- the only suspended part

```
1 datatype  $\alpha$  Queue = Queue of
2   {W :  $\alpha$  list, F :  $\alpha$  list susp, LenF : int, R :  $\alpha$  list, LenR : int}
```

```
1 (* invariante 1: W is no-empty whenever F is non-empty *)
2 fun checkW {W = [], F = f, LenF = lenF, R = r, LenR = lenR} =
3   Queue {W = force f, F = f, LenF = lenF, R = r, LenR = lenR}
4 | checkW q = Queue q
5 (* invariante 2: R is no longer than F *)
6 fun checkR (q as {W = w, F = f, LenF = lenF, R = r, LenR = lenR}) =
7   if lenR <= lenF then q
8   else let val w' = force f
9        in {W = w', F = $(w' @ rev r), LenF = lenF+lenR, R=[], LenR=0} end
10 (* check invariants *)
11 fun inv-queue q = checkW (checkR q)
```

```
1 fun enqueue (Queue {W = w, F = f, LenF = lenF, R = r, LenR = lenR}, x) =
2   inv-queue {W = w, F = f, LenF = lenF, R = x::r, LenR = lenR+1}
3 fun head (Queue {W = x::w, ...}) = x
4 fun tail (Queue {W = x::w, F = f, LenF = lenF, R = r, LenR = lenR}) =
5   inv-queue {W = w, F = $tl (force f), LenF=lenF-1, R=r, LenR=lenR}
```

## Problem Statement

- We produce  $n$  ``cheap" steps
- Then, one ``expensive" step  $O(n)$
- Thus, we can only state amortized complexity

## An Idea: *scheduling*

Instead of one ``expensive" step let's perform  $n$  smaller steps with constant complexity. Performing each ``cheap" step, we will also perform one of this ``smaller" steps.

Reminder: banker's queue: we relied on calculation of  $f \mathbin{++} \text{reverse}(r)$

Now let's instead use a special function `rotate`

$$\text{rotate}(f, r, a) = f \mathbin{++} \text{reverse}(r) \mathbin{++} a$$

Third parameter is an accumulator which stores partially computed result of `reverse(r)`

Obviously

$$\text{rotate}(f, r, \$Nil) = f \mathbin{++} \text{reverse}(r)$$

## When to Reorder the Queue?

Let's reorder queue when  $|r| = |f| + 1$

This ratio will be maintained throughout the rebuilding

Let's prove it by induction on the length of front  $|f|$

Base:

$$\begin{aligned} \text{rotate}(\text{\$Nil}, \text{\$Cons}(y, \text{\$Nil}), a) &\equiv \text{\$Nil} \mathbin{++} \text{reverse}(\text{\$Cons}(y, \text{\$Nil})) \mathbin{++} a \\ &\equiv \text{\$Cons}(y, a) \end{aligned}$$

Induction step:

$$\begin{aligned} &\text{rotate}(\text{\$Cons}(x, f), \text{\$Cons}(y, r), a) \\ &\quad \equiv \text{\$Cons}(x, f) \mathbin{++} \text{reverse}(\text{\$Cons}(y, r)) \mathbin{++} a \\ &\quad \equiv \text{\$Cons}(x, f \mathbin{++} \text{reverse}(\text{\$Cons}(y, r)) \mathbin{++} a) \\ &\quad \equiv \text{\$Cons}(x, f \mathbin{++} \text{reverse}(r) \mathbin{++} \text{\$Cons}(y, a)) \\ &\quad \equiv \text{\$Cons}(x, \text{rotate}(f, r, \text{\$Cons}(y, a))) \end{aligned}$$



# Real-time Queue: code

```
1  (* S is a schedule -- suffix s.t. all nodes before S in F have already
2  been forced and !memoized! *)
3  datatype Queue = Queue of {F :  $\alpha$  stream, R :  $\alpha$  list, S :  $\alpha$  stream}
4
5  val empty = Queue {F = $Nil, R = [], S = $Nil}
6  fun isEmpty (Queue {F = f, ... }) = null f
7
8  fun rotate (f, r, a) = $case (f, r) of
9    ($Nil, $Cons (y, _)) => Cons (y, a)
10   | ($Cons (x, f'), $Cons (y, r')) => Cons (x, rotate (f', r', $Cons (y, a)))
11
12  (* Invariant: |S| = |F| - |R| *)
13  fun inv {F = f, R = r, S = $Cons (x, s)} = Queue {F = f, R = r, S = s}
14    | inv {F = f, R = r, S = $Nil} =
15      let val f' = rotate (f, r, $Nil) (* create a suspension *)
16      in Queue {F = f', R = [], S = f'} end (* store it in both F and S *)
17
18  fun enqueue (Queue {F=f, R=r, S=s}, x) = inv {F = f, R = x::r, S = s}
19
20  fun head (Queue {F = $Cons (x, f), ...}) = x
21
22  fun tail (Queue {F = $Cons (x, f), R=r, S=s}) = inv {F=f, R=r, S=s}
```

# Conclusion

Queue\ Operation	enqueue	head	tail
Banker's	$O(1)^*$	$O(1)^*$	$O(1)^*$
Physicist's	$O(1)^*$	$O(1)^*$	$O(1)^*$
Real-time	$O(1)$	$O(1)$	$O(1)$

Amortized estimations are marked with \*

Questions?