# Adv. Data Structures: Functional queues

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# **Project**

In this project we perform an experimental study of functional data structures. The implementations are done in Haskell, which is a lazy language. Due to this we have to make sure that our results are actually used, so the evaluations are not postponed.

We implement a queue using

- 1. A standard Haskell list.
- 2. A pair of lists, with amotized O(1) guarantee if the same queue will never be argument to repeated queue operations.
- 3. A pair of lists, exploiting the properties of lazy evaluation of list concatenation to guranteed amortized O(1) per queue operation.
- 4. A O(1) list, with a worst-case gurantee of O(1) per queue operation (if executed strictly).

We then design and perform experiements comparing the different implementations, where we cover the worst-case scenario for every queue

# Queue implementations

A list in Haskell is nothing more than a singly linked list. So in other words, it's cheap to add or remove the first element, but very expensive to add an element to the tail, or remove the last element. Throughout the description of the queues, we will omit talking about trivial cases for the remove operation when we have an empty queues. This is because the only sensible action is to return nothing.

#### A standard Haskell list

Using a standard Haskell list is a very simple data structure. Unfortunately it is only usable as a queue if you gurantee that it only contain a few elements. Because of how Haskell represent lists, when inserting at the end of the queue, it will have to traverse the entire list, before appending the new element. Thus the complexity for inserting is O(n). Removing on the other hand, is fairly simple. You can simply remove the first element in constant time. Thus O(1) for remove.

#### Worst-case input

The worst-case test is fairly straight forward. Simply insert n elements and that is it. Note that the lazy nature of append, require us to force strict behaviour **HOW DO WE DO THIS?** 

### A pair of lists

Using a pair of lists, we can obtain amotized O(1) as long as we do not repeat expensive operations. The queue works by having a pair of lists, a left and a right. When inserting, we will add the element to the head of right. Removing on the other hand is a bit more complicated. There are 2 cases:

- 1. left contain 1 or more elements.
- 2. left is empty and right contains 1 or more elements.

For (1), the remove operation will simply remove the head of left. For (2) it's a bit more involved. The list right is reversed, become left and the first element is removed, this makes sense, because that will be the oldest element in the queue. The running time of the insert operation is still O(1), however the remove operation can be O(n), especially, the first time er remove an element, we will reverse right. If we gurantee that we do not repeat the expensive operation, we have a O(1) amotized (for every element we have in the reverse operation, we will do a remove paying for that element). However, it is common in the functional paradigm to reuse functions. Therefore one should take care to not use expensive operation more than once. This also provide us with the worst-case behaviour.

#### Worst-case input

Just as with the single list queue, the worst-case test is straight forward: insert n elements and repeat the remove of the first element. Note for this to be true, we have to avoid memoirzation.

### A pair of lists, exploiting laziness

As described in the paper Simple and Efficient Purely Functional Queues and Deques by Chris Okasaki, we can exploit lazyness to achieve the same O(1) amotized bound, but with an improved worst-case to  $O(\log n)$ .

To achieve this, we want to calculate the reverse of right in an incremental fashion, i.e. left ++ rev(right), where ++ is the append operation. Since append is lazy, we have to make the reverse lazy as well. To do this, we replace  $\langle left, right \rangle$  with  $\langle left ++ rev(right), \parallel \rangle$  periodically. This is called a rotation, denoted rot, and is a fusion of append and reverse. The function rot takes as input 3 lists, left, right and acc - it will then peel of the elements of left one by one. For each element it will also take off an element from right and append it to acc. Assuming that left and right have the same length, once the recursion have removed all elements from left, the content of acc will be the reversed right. Because append operation is lazy, the recursive calls to rot will be lazy as well.

The details of this can be found in the before mentioned paper.

The data structure works by maintaining an invariant,  $|right| \leq |left|$ . We

maintain this invariant by calling rot whenever right become strictly larger than left, i.e. when |right| = |left| + 1. So, when inserting we make a call to rot every time we have inserted |left| + 1 elements. But the lazy nature of rot means the actual evaluation will be deferred until needed. This makes insert  $O(\log 1)$ . When we call remove, it is possible that left consists of a series of deferred rot calls, all of which have to be evaluated. We know that each of those calls will reduce the size of left in half, thus the recursion will have depth of log n. Again the lazy nature of rot ensure that only log n work will be done, giving the O(log n) bound.

#### Worst-case input

We have already mentioned how left should be, to enforce the worst-case behaviour. By inserting  $1+2^i$  for any i will leave us with such a senarios for the first time we call remove. Note, due to memoization repeating this call will not be expensive for any successive calls. This is importent when timing several runs to gain an average.

### A O(1) list

In the paper Okasaki also describe a queue that have O(1) worst-case bounds, and the idea is an improvement of the previous queue structure. The expensive operation of the previous queue, was evaluating the tail of left, thus we need to make sure this have already been evaluated and memoized before a call to remove is made. This is called pre-evaluation, and works in the following way: The queue is a triple of lists, left, right, and lefthat, where lefthat is some tail of left representing the unevaluated part of left. Thus if lefthat = [] all tails of left have been evaluated. Every call to insert and remove, that does not call cause a rotation, will evaluate one step of lefthat. This gurantee us that when we do call rot we have evaluated all tails, and memoirzing will cause it to be constant in time. After a rotation, lefthat = left, and right = 0. It's easy to see that |lefthat| = |left| - right.

#### Worst-case input

This data structure do give worst-case O(1) gurantees, and there is no particular input that will make it blow up. Therefore it will serve as reference

when performing the benchmarking of the other data structures worst-case inputs.

# Experiments with worst-case

#### A standard Haskell list

The worst-case test for this queue is simply a timing of how long it takes to insert n elements and then removing them again. Since either insert or remove will be O(n), the graph is expected to grow in a super linear shape. This is consistent with figure 1. All other queues are obviously much faster, and that is in line with the O(1) gurantees.

### A pair of lists

The worst-case here, is reusing the function to repeately extract the element that will cause the queue to reverse the list. This is seen on figure 3. All other queues are much faster in this aspect. The simple list because the remove operation is constant, the paired lazy list because it gurantee a  $O(\log n)$  worst-case the the pre-evaluated list because it gurantee a O(1).

The list still gurantee a amotized O(1) if we do not repeat expensive operations, and that is seen in figure 2

# A pair of lists, exploiting laziness

The worst-case test for this queue is a bit more tricky. We didn't get it quite right, and the graph in figure 4 was the best we managed to produce. Unfortunately it's not really useful, but what we expect to see is that the lazy pair queue will have a  $O(\log n)$  growth, with list of pairs queue O(n) as described above, and the other two queues to have a constant time remove.

The amotized O(1) should be obvious on figure 2, but it seems to grow in a super linear shape. We do not have an explanation for this, but maybe a bigger input will normalize the graph, or something fundamental in the Haskell VM causes this.

### A O(1) list

Since the queue gurantee worst-case O(1) for both operations, we can use it for comparison with the other queues, and see that it works fast in all cases. For the standard insert/remove benchmark, it unfortunately show something that looks like super liniar growth, just as the lazy pairs queue. We do not think cache effects are the cause since we're well beyond 1 million elements in the queue. This again leave our best bet to be the nature of the Haskell VM (garbage collector and other mechanics).

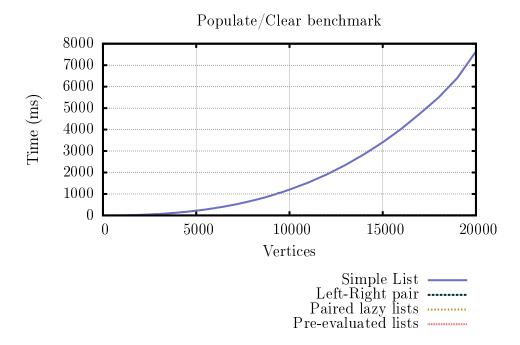


Figure 1: blah

The graph below looks unusual because of the way our benchmark works. In order to force the worst-case scenario of the paired lazy lists, the size of the left list has to be exactly one less than the size of the right list. Given the nature of our benchmarking framework we are stuck with whatever size we are given. Therefore we fill up the queue until the two lists are exactly the same. For any given list size its actual size is going to be one less than the subsequent exponent of 2. Since the number of insertions shrink as the size increases, the lines are declining slightly only to jump at the point where we have reached a new exponent of 2.

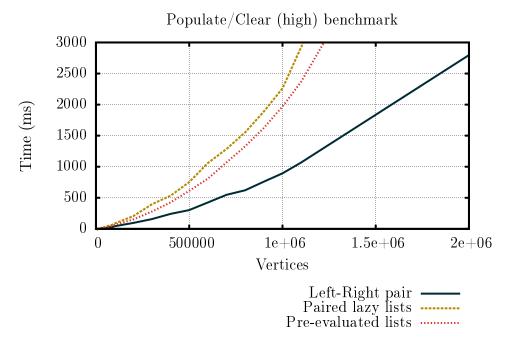


Figure 2: blah

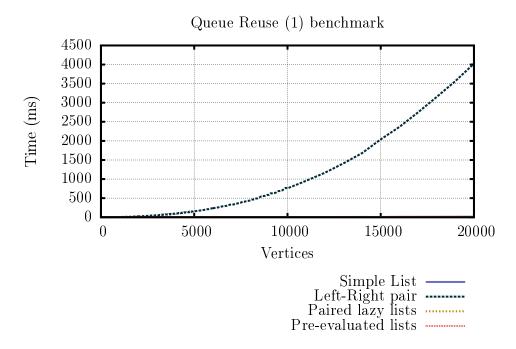


Figure 3: blah

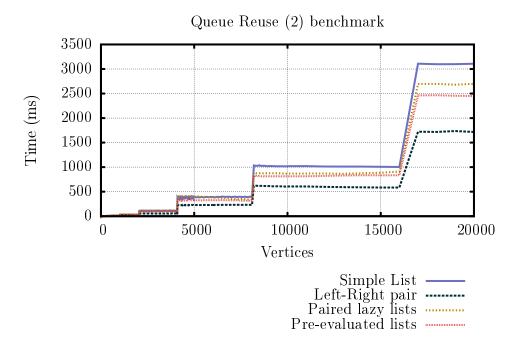


Figure 4: blah