

Algorithmen / Algorithms II

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2 Advanced Data Structures

Here using the example of priority queues.

further examples:

Perfect Hashing

Monotone integer priority queues

chapter: shortest paths

chapter: randomized algorithms

- search trees with advanced operations see book
 - External priority queues

chapter: external algorithms

Geometric data structures

chapter: geom. algorithms

2.1 Addressable Priority Queues – Operations

Procedure build($\{e_1, \ldots, e_n\}$) $M := \{e_1, \ldots, e_n\}$ **Function** size **return** |M| **Procedure** insert(e) $M := M \cup \{e\}$ **Function** min **return** minM **Function** deleteMin $e := \min M$; $M := M \setminus \{e\}$; **return** e **Function** remove(h : Handle) e := h; $M := M \setminus \{e\}$; **return** e **Procedure** decreaseKey(h : Handle, k : Key) **assert** key(h) $\ge k$; key(h):= k**Procedure** merge(M') $M := M \cup M'$

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Addressable Priority queues – Use Cases

- Dijkstra's algorithm shortest paths
- Jarník-Prim algorithm for minimum spanning trees
- Here: hierarchy construction for route planning
- Here: graph partitioning
- Here: disk scheduling

In general:

Greedy algorithms, where priorities change (within limits).



Basic Data Structure

A forest of heap-ordered trees



Generalization of binary heaps:

Tree \rightarrow forest

Binary \rightarrow arbitrary node-degrees



Processing Forests

Cut:



Link:

union(a,b): link(min(a,b),max(a,b))



Pairing Heaps

[Fredman Sedgewick Sleator Tarjan 1986]

```
\begin{array}{l} \textbf{Procedure } \text{insertItem}(h: \text{Handle}) \\ \text{newTree}(h) \end{array}
```

```
Procedure newTree(h : Handle)
forest:= forest \cup \{h\}
if *h < \min then minPtr:= h
```



Attention: Simple implementation from the 1. English edition.

Further editions differ (e.g. the German edition).

Pairing Heaps

Procedure decreaseKey(h : Handle, k : Key)
key(h):= k
if h is not a root then cut(h)
else update minPtr
Procedure cut(h : Handle)
remove the subtree rooted at h
newTree(h)





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Pairing Heaps

Function deleteMin : Handle

m:= minPtr

forest:= forest $\setminus \{m\}$

foreach child h of m do newTree(h)



perform pair-wise union operations on the roots in forest update minPtr

return m





Pairing Heaps

Procedure merge(*o* : AdressablePQ)

if *minPtr > *(o.minPtr) then minPtr:= o.minPtr forest:= forest $\cup o$.forest o.forest:= Ø



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Pairing Heaps – Representation





Pairing Heaps – Analysis

insert, merge: O(1)

deleteMin, remove: $O(\log n)$ amortized

decreaseKey: unknown! $O(\log \log n) \le T \le O(\log n)$ amortized, but fast in practice.

Proofs: not here.



Fibonacci Heaps [Fredman Tarjan 1987]

Rank: Save the number of (immediate) children.

Union-by-rank: Only call union on roots with the same rank.

Mark: Mark nodes that have lost a child.

Cascading cuts: Cut at marked nodes

(i.e. nodes that have lost two children)

Theorem: Amortised complexity $O(\log n)$ for deleteMin and remove O(1) for all other operations.

(i.e. *total time* = $O(o + d \log n)$ if

d = #deleteMin, o = #otherOps, $n = \max |M|$)



Fibonacci Heaps – Representation

Roots: in a doubly linked list

(and a temporary array for deleteMin)



insert, merge: as before, in time O(1)



deleteMin with Union-by-Rank

Function deleteMin : Handle

```
m := \min Ptr
```

forest:= forest $\setminus \{m\}$



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foreach child *h* of *m* **do** newTree(*h*) while $\exists a, b \in$ forest : rank(*a*) = rank(*b*) **do**

union(a,b) // increments rank of surviving root

update minPtr

return m



Fast Union-by-Rank

An array that is addressed by the rank.

Execute link until a free entry is found.



Analysis: Time O(#unions + |forest|)



Amortised Analysis for deleteMin

```
maxRank:= \max_{a \in \text{forest}} \operatorname{rank}(a) (after)
```

Lemma: $T_{\text{deleteMin}} = O(\text{maxRank})$

Proof: Using the accounting method. One token per Root rank(minPtr) \leq maxRank

 \rightsquigarrow costs of O(maxRank) for newTrees and a new token.

Union-by-rank: token pays for

 \Box union operations (a token becomes free) and

iterating through roots (old and new).

At the end there are $\leq \max \text{Rank}$ roots.



Why is maxRank logarithmic? – Binomial Trees

 $2^k + 1 \times \text{insert}, 1 \times \text{deleteMin} \rightsquigarrow \text{rank } k$



[Vuillemin 1978] PQ (only) with binomial trees, $T_{decreaseKey} = O(\log n)$. Problem: Cuts can lead to high ranking trees.

Cascading Cuts

Procedure decreaseKey(h : Handle, k : Key) key(h):= kcascadingCut(h)

Procedure cascadingCut(h)
 if h is not a root then
 p:= parent(h)
 unmark h
 cut(h)
 if p is marked then
 cascadingCut(p)
 else mark p



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We will show: cascading cuts keep maxRank logarithmic



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Lemma: decreaseKey has amortised complexity O(1). Accounting Method: (\approx 1 Token per cut or union)

- 1 token per root
- 2 tokens for every marked node

Looking at decreaseKey with k consecutive marked predecessors:

- 2k token becomes free (nodes become unmarked)
- 2 token needed for new marks
- k+1 tokens needed for the new roots
- k+1 tokens pay for the cuts

Thus, there remains a cost of 4 tokens +O(1) time for decreaseKey



Here is where Mr. Fibonacci comes in.

$$F_i := \begin{cases} 0 & \text{für i=0} \\ 1 & \text{für i=1} \\ F_{i-2} + F_{i-1} & \text{else} \end{cases}$$

Known: $F_{i+1} \ge ((1+\sqrt{5})/2)^i \ge 1.618^i$ for all $i \ge 0$.

We show:

A subtree with root *v* and rank(*v*) = *i* contains $\geq F_{i+2}$ elements. \Rightarrow

logarithmic time for deleteMin.

Proof:

Looking at the moment when the *j*-th child w_j of v was added:

$$\begin{split} w_j \text{ and } v \text{ had the same rank} &\geq j-1 \quad (v \text{ already had } j-1 \text{ children}) \\ \operatorname{rank}(w_j) \text{ was reduced by at most one} & (\operatorname{cascading cuts}) \\ &\Rightarrow \operatorname{rank}(w_j) \geq j-2 \text{ and } \operatorname{rank}(v) \geq j-1 \end{split}$$



Addressable Priority Queues – More

Lower bound $\Omega(\log n)$ for deleteMin (comparison based) Proof: exercise

- Worst case Bounds: not here
- Monotone PQs with integer keys (stay tuned)

Open Problems:

Analysis of pairing heaps (simplification of Fibonacci Heaps)



Recap Data Structures

- In this lecture, we focused on the example of priority queues (see shortest path algorithms and external algorithms).
 - Heap concept can take you far.
- Sibling-pointers can be used to represent arbitrary trees with a constant number of pointers per item.
- Fibonacci heaps a non-trivial example for amortised analysis