# **Amortized Analysis**

#### **Definition 1**

A data structure with operations  $op_1(), \ldots, op_k()$  has amortized running times  $t_1, \ldots, t_k$  for these operations if the following holds.

Suppose you are given a sequence of operations (starting with an empty data-structure) that operate on at most n elements, and let  $k_i$  denote the number of occurences of  $op_i()$  within this sequence. Then the actual running time must be at most  $\sum_i k_i \cdot t_i(n)$ .



## **Potential Method**

#### Introduce a potential for the data structure.

- $\Phi(D_i)$  is the potential after the *i*-th operation.
- Amortized cost of the *i*-th operation is

 $\hat{c}_i = c_i + \Phi(D_i) - \Phi(D_{i-1}) \ . \label{eq:ci}$ 

• Show that  $\Phi(D_i) \ge \Phi(D_0)$ .

Then

$$\sum_{i=1}^{k} c_i \le \sum_{i=1}^{k} c_i + \Phi(D_k) - \Phi(D_0) = \sum_{i=1}^{k} \hat{c}_i$$

This means the amortized costs can be used to derive a bound on the total cost.



# **Example: Stack**

### Stack

- S. push()
- ▶ S. pop()
- S. multipop(k): removes k items from the stack. If the stack currently contains less than k items it empties the stack.
- The user has to ensure that pop and multipop do not generate an underflow.

### Actual cost:

- S. push(): cost 1.
- S. pop(): cost 1.
- ► S. multipop(k): cost min{size, k} = k.



### **Example: Stack**

Use potential function  $\Phi(S)$  = number of elements on the stack.

#### Amortized cost:

S. push(): cost

 $\hat{C}_{\text{push}} = C_{\text{push}} + \Delta \Phi = 1 + 1 \le 2$ .

S. pop(): cost

$$\hat{C}_{\rm pop} = C_{\rm pop} + \Delta \Phi = 1 - 1 \le 0$$
 .

Note that the analysis becomes wrong if pop() or multipop() are called on an empty stack.

► S. multipop(k): cost

 $\hat{C}_{\rm mp} = C_{\rm mp} + \Delta \Phi = \min\{\text{size}, k\} - \min\{\text{size}, k\} \le 0$ .



# **Example: Binary Counter**

#### Incrementing a binary counter:

Consider a computational model where each bit-operation costs one time-unit.

Incrementing an n-bit binary counter may require to examine n-bits, and maybe change them.

#### Actual cost:

- Changing bit from 0 to 1: cost 1.
- Changing bit from 1 to 0: cost 1.
- Increment: cost is k + 1, where k is the number of consecutive ones in the least significant bit-positions (e.g, 001101 has k = 1).



### **Example: Binary Counter**

Choose potential function  $\Phi(x) = k$ , where k denotes the number of ones in the binary representation of x.

#### Amortized cost:

Changing bit from 0 to 1:

$$\hat{C}_{0\to 1} = C_{0\to 1} + \Delta \Phi = 1 + 1 \le 2$$
.

• Changing bit from 1 to 0:

$$\hat{C}_{1\to 0} = C_{1\to 0} + \Delta \Phi = 1 - 1 \le 0$$
.

Increment: Let k denotes the number of consecutive ones in the least significant bit-positions. An increment involves k (1 → 0)-operations, and one (0 → 1)-operation.

Hence, the amortized cost is  $k\hat{C}_{1\rightarrow 0} + \hat{C}_{0\rightarrow 1} \le 2$ .

Collection of trees that fulfill the heap property.

Structure is much more relaxed than binomial heaps.





#### Additional implementation details:

- Every node x stores its degree in a field x. degree. Note that this can be updated in constant time when adding a child to x.
- Every node stores a boolean value x.marked that specifies whether x is marked or not.



#### The potential function:

- t(S) denotes the number of trees in the heap.
- m(S) denotes the number of marked nodes.
- We use the potential function  $\Phi(S) = t(S) + 2m(S)$ .



The potential is  $\Phi(S) = 5 + 2 \cdot 3 = 11$ .



We assume that one unit of potential can pay for a constant amount of work, where the constant is chosen "big enough" (to take care of the constants that occur).

To make this more explicit we use *c* to denote the amount of work that a unit of potential can pay for.



#### S. minimum()

- Access through the min-pointer.
- ► Actual cost O(1).
- No change in potential.
- Amortized cost  $\mathcal{O}(1)$ .



- S. merge(S')
  - Merge the root lists.
  - Adjust the min-pointer

- In the figure below the dashed edges are replaced by red edges.
- The minimum of the left heap becomes
  - the new minimum of the merged heap.



### **Running time:**

- Actual cost  $\mathcal{O}(1)$ .
- No change in potential.
- Hence, amortized cost is  $\mathcal{O}(1)$ .



- S. insert(x)
  - Create a new tree containing x.
  - Insert x into the root-list.
  - Update min-pointer, if necessary.





#### **Running time:**

- Actual cost  $\mathcal{O}(1)$ .
- Change in potential is +1.
- Amortized cost is c + O(1) = O(1).



 $D(\min)$  is the number of children of the node that stores the minimum.

#### S. delete-min(x)

- ▶ Delete minimum; add child-trees to heap; time: D(min) · O(1).
- Update min-pointer; time:  $(t + D(\min)) \cdot O(1)$ .





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• Consolidate root-list so that no roots have the same degree. Time  $t \cdot O(1)$  (see next slide).



#### **Consolidate:**



During the consolidation we traverse the root list. Whenever we discover two trees that have the same degree we merge these trees. In order to efficiently check whether two trees have the same degree, we use an array that contains for every degree value d a pointer to a tree left of the current pointer whose root has degree d (if such a tree exist).



#### 8.3 Fibonacci Heaps









#### Consolidate:





8.3 Fibonacci Heaps





















t and t' denote the number of trees before and after the delete-min() operation, respectively.  $D_n$  is an upper bound on the degree (i.e., number of children) of a tree node.

#### Actual cost for delete-min()

- At most  $D_n + t$  elements in root-list before consolidate.
- ► Actual cost for a delete-min is at most O(1) · (D<sub>n</sub> + t). Hence, there exists c<sub>1</sub> s.t. actual cost is at most c<sub>1</sub> · (D<sub>n</sub> + t).

#### Amortized cost for delete-min()

- $t' \leq D_n + 1$  as degrees are different after consolidating.
- Therefore  $\Delta \Phi \leq D_n + 1 t$ ;
- We can pay  $\mathbf{c} \cdot (t D_n 1)$  from the potential decrease.
- The amortized cost is

 $c_1 \cdot (D_n + t) - \mathbf{c} \cdot (t - D_n - 1)$ 

 $\leq (c_1+c)D_n+(c_1-c)t+c \leq 2c(D_n+1) \leq \mathcal{O}(D_n)$ 

for  $\textbf{\textit{c}} \geq \textbf{\textit{c}}_1$  .



If the input trees of the consolidation procedure are binomial trees (for example only singleton vertices) then the output will be a set of distinct binomial trees, and, hence, the Fibonacci heap will be (more or less) a Binomial heap right after the consolidation.

If we do not have delete or decrease-key operations then  $D_n \leq \log n$ .





#### Case 1: decrease-key does not violate heap-property

Just decrease the key-value of element referenced by *h*.
 Nothing else to do.





Case 2: heap-property is violated, but parent is not marked

- Decrease key-value of element x reference by h.
- If the heap-property is violated, cut the parent edge of x, and make x into a root.
- Adjust min-pointers, if necessary.
- Mark the (previous) parent of x (unless it's a root).





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#### Case 3: heap-property is violated, and parent is marked

- Decrease key-value of element x reference by h.
- Cut the parent edge of *x*, and make *x* into a root.
- Adjust min-pointers, if necessary.
- Continue cutting the parent until you arrive at an unmarked node.





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#### Case 3: heap-property is violated, and parent is marked

- Decrease key-value of element x reference by h.
- Cut the parent edge of *x*, and make *x* into a root.
- Adjust min-pointers, if necessary.
- Execute the following:

```
p \leftarrow parent[x];

while (p is marked)

pp \leftarrow parent[p];

cut of p; make it into a root; unmark it;

p \leftarrow pp;

if p is unmarked and not a root mark it;
```



Marking a node can be viewed as a first step towards becoming a

#### Actual cost:

- Constant cost for decreasing the value.
- Constant cost for each of  $\ell$  cuts.
- Hence, cost is at most  $c_2 \cdot (\ell + 1)$ , for some constant  $c_2$ .

#### Amortized cost:

- $t' = t + \ell$ , as every cut creates one new root.
- ▶  $m' \le m (\ell 1) + 1 = m \ell + 2$ , since all but the first cut unmarks a node; the last cut may mark a node.
- $\bullet \ \Delta \Phi \le \ell + 2(-\ell + 2) = 4 \ell$
- Amortized cost is at most  $c_2(\ell+1) + c(4-\ell) \le (c_2-c)\ell + 4c + c_2 = O(1)$ , trees before and after m and m': number of marked nodes before and after operation.



t and t': number of

### **Delete node**

#### *H*.delete(*x*):

- decrease value of x to  $-\infty$ .
- delete-min.

#### Amortized cost: $\mathcal{O}(D_n)$

- $\mathcal{O}(1)$  for decrease-key.
- $\mathcal{O}(D_n)$  for delete-min.



#### Lemma 2

Let x be a node with degree k and let  $y_1, ..., y_k$  denote the children of x in the order that they were linked to x. Then

degree
$$(y_i) \ge \begin{cases} 0 & \text{if } i = 1\\ i - 2 & \text{if } i > 1 \end{cases}$$

The marking process is very important for the proof of this lemma. It ensures that a node can have lost at most one child since the last time it became a non-root node. When losing a first child the node gets marked; when losing the second child it is cut from the parent and made into a root.



#### Proof

- ▶ When y<sub>i</sub> was linked to x, at least y<sub>1</sub>,..., y<sub>i-1</sub> were already linked to x.
- ► Hence, at this time degree(x) ≥ i − 1, and therefore also degree(y<sub>i</sub>) ≥ i − 1 as the algorithm links nodes of equal degree only.
- Since, then  $y_i$  has lost at most one child.
- Therefore, degree( $y_i$ )  $\ge i 2$ .



- Let s<sub>k</sub> be the minimum possible size of a sub-tree rooted at a node of degree k that can occur in a Fibonacci heap.
- $s_k$  monotonically increases with k
- $s_0 = 1$  and  $s_1 = 2$ .

Let x be a degree k node of size  $s_k$  and let  $y_1, \ldots, y_k$  be its children.

$$s_k = 2 + \sum_{i=2}^{k} \operatorname{size}(y_i)$$
$$\geq 2 + \sum_{i=2}^{k} s_{i-2}$$
$$= 2 + \sum_{i=0}^{k-2} s_i$$



8.3 Fibonacci Heaps

#### **Definition 3**

Consider the following non-standard Fibonacci type sequence:

$$F_{k} = \begin{cases} 1 & \text{if } k = 0\\ 2 & \text{if } k = 1\\ F_{k-1} + F_{k-2} & \text{if } k \ge 2 \end{cases}$$

#### Facts:

- 1.  $F_k \ge \phi^k$ .
- **2.** For  $k \ge 2$ :  $F_k = 2 + \sum_{i=0}^{k-2} F_i$ .

The above facts can be easily proved by induction. From this it follows that  $s_k \ge F_k \ge \phi^k$ , which gives that the maximum degree in a Fibonacci heap is logarithmic.



k=0:  
k=1:  
k-2,k-1 → k: 
$$F_k = F_{k-1} + F_{k-2} \ge \Phi^{k-1} + \Phi^{k-2} = \Phi^{k-2} (\Phi^{k-1}) = \Phi^k$$

k=2: 
$$3 = F_2 = 2 + 1 = 2 + F_0$$
  
k-1  $\rightarrow$  k:  $F_k = F_{k-1} + F_{k-2} = 2 + \sum_{i=0}^{k-3} F_i + F_{k-2} = 2 + \sum_{i=0}^{k-2} F_i$ 

