Lecture 8: Trees and Heap Sort COMS10007 - Algorithms

Dr. Christian Konrad

19.02.2019

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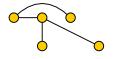
Data Structures

- Data storage format that allows for efficient access and modification
- Building block of many efficient algorithms
- For example, an array is a data structure

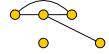
Definition: A tree T = (V, E) of size n is a tuple consisting of

$$V = \{v_1, v_2, \dots, v_n\}$$
 and $E = \{e_1, e_2, \dots, e_{n-1}\}$

with |V| = n and |E| = n - 1 with $e_i = \{v_j, v_k\}$ for some $j \neq k$ such that for every node v_i there is at least one edge e_j such that $v_i \in e_j$. V are the nodes/vertices and E are the edges of T.



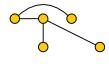




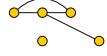
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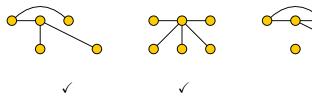




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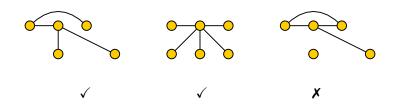




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Rooted Trees

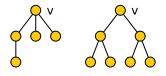
Definition: (rooted tree) A *rooted* tree is a triple T = (v, V, E) such that T = (V, E) is a tree and $v \in V$ is a designed node that we call the *root* of T.





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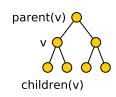
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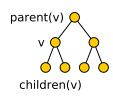
Definition: (leaf, internal node) A *leaf* in a tree is a node with exactly one incident edge. A node that is not a leaf is called an *internal node*.

Further Definitions:

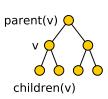
 The parent of a node v is the closest node on a path from v to the root.
 The root does not have a parent.



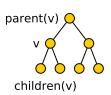
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- The *height* of a tree is the length of a longest root-to-leaf path.
- The degree deg(v) of a node v is the number of incident edges to v. Since every edge is incident to two vertices we have

$$\sum_{v\in V} \deg(v) = 2\cdot |E| = 2(n-1).$$

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• The *level* of a vertex v is the length of the unique path from the root to v plus 1.

Property:

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Proof

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Proof Let $L \subseteq V$ be the subset of leaves.

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Proof Let $L \subseteq V$ be the subset of leaves. Suppose that there is at most 1 leaf, i.e., $|L| \le 1$. Then:

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a contradiction to the fact that $\sum_{v \in V} \deg(v) = 2(n-1)$ in every tree.

Binary Trees

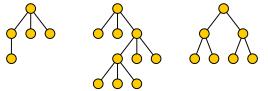
Definition: (k-ary tree) A (rooted) tree is k-ary if every node has at most k children. If k=2 then the tree is called binary. A k ary tree is

- full if every internal node has exactly k children,
- complete if all levels except possibily the last is entirely filled (and last level is filled from left to right),
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complete 3-ary tree full 3-ary tree perfect binary tree

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Remark: The runtime of many algorithms that use tree data structures depends on the height of these trees. We are therefore interested in using complete/perfect trees.

Priority Queues

Priority Queue:

Data structure that allows the following operations:

- Build(.): Create data structure given a set of data items
- Extract-Max(.): Remove the maximum element from the data structure
- others...

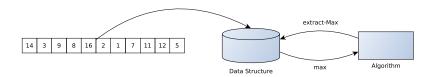
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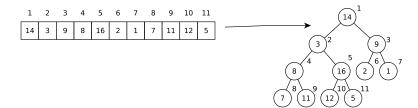
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Sorting using a Priority Queue

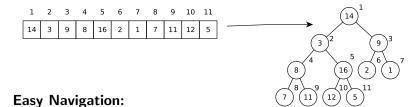


Interpretation of an Array as a Complete Binary Tree

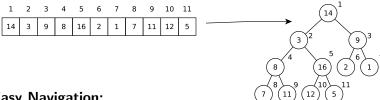
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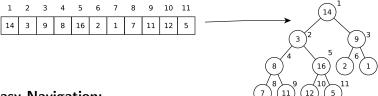
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Easy Navigation:

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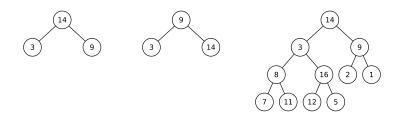
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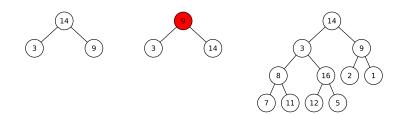
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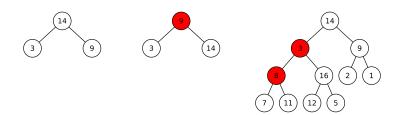
The Heap Property



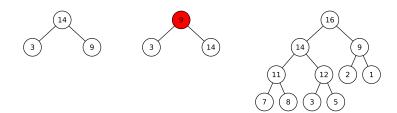
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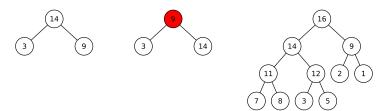


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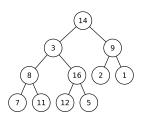
Key of nodes larger than keys of their children



Heap Property \rightarrow Maximum at root Important for Extract-Max(.)

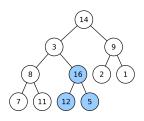
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- If node does not fulfill Heap Property: Heapify()



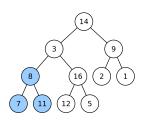
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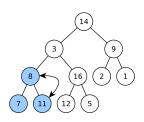
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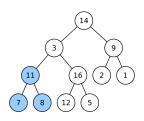
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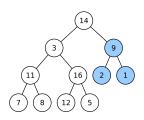
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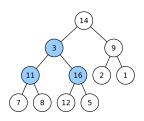
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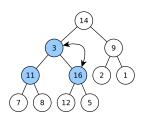
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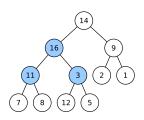
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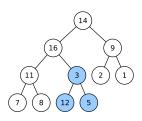
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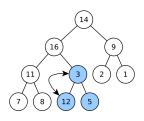
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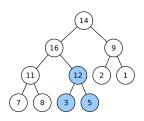
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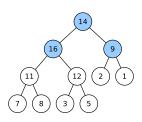
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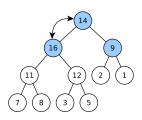
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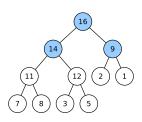
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Runtime of Heapify()

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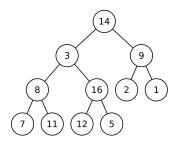
Constructing a Heap: Time $O(n \log n)$

More Precise Analysis of the Heap Construction Step

- Heapify(x): $O(\text{depth of subtree rooted at } x) = O(\log n)$
- Observe: Most nodes close to the "bottom"

Analysis:

- Let *i* be the largest integer such that $n' := 2^i 1$ and n' < n
- There are at most n' internal nodes (candidates for Heapify())
- These nodes are contained in a perfect binary tree
- This tree has height i-1

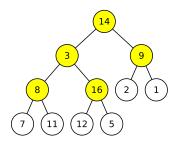


More Precise Analysis of the Heap Construction Step

- Heapify(x): $O(\text{depth of subtree rooted at } x) = O(\log n)$
- Observe: Most nodes close to the "bottom"

Analysis:

- Let *i* be the largest integer such that $n' := 2^i 1$ and n' < n
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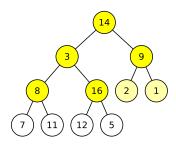


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Analysis

We sum over all levels, count the number of nodes per level, and multiply with the depth of their subtrees:

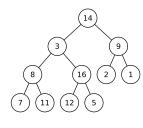
$$\sum_{j=1}^{i} \underset{\text{nodes in level } i-j}{\underbrace{2^{i-j}}} \cdot \underbrace{j}_{\text{depth of subtree}}$$

$$\sum_{j=1}^{i} 2^{i-j} \cdot j = 2^{i} \cdot \sum_{j=1}^{i} \frac{j}{2^{j}} = O(2^{i}) = O(n^{i}) = O(n).$$

We'll prove $\sum_{j=1}^{i} \frac{j}{2^{j}} = O(1)$ very soon...!

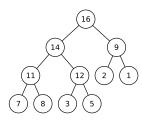
	14	3	9	8	16	2	1	7	11	12	5
--	----	---	---	---	----	---	---	---	----	----	---

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- Repeat n times:
 - Swap root with last element
 - ② Decrease size of heap by 1
 - Heapify()



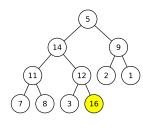
16	14	a	11	12	2	1	7	Ω	3	5
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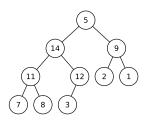
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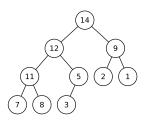
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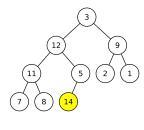
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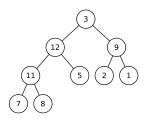
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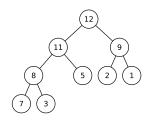
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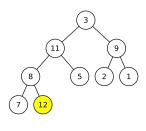
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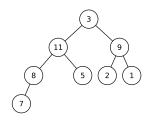
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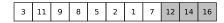


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Putting Everything Together



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Putting Everything Together



- Build-heap() O(n)
- 2 Repeat *n* times:
 - **1** Swap root with last element O(1)
 - ② Decrease size of heap by 1 O(1)
 - **3** Heapify() $O(\log n)$

Runtime: $O(n \log n)$

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$$= \sum_{i=0}^{n-1} \frac{1}{2^i} + \frac{n-1}{2^n} = \frac{\frac{1}{2^n} - \frac{1}{2}}{\frac{1}{2} - 1} + \frac{n-1}{2^n} = O(1).$$