

# 8. Sorting II

Heapsort, Quicksort, Mergesort

# 8.1 Heapsort

[Ottman/Widmayer, Kap. 2.3, Cormen et al, Kap. 6]

# Heapsort

Inspiration from selectsort: fast insertion

Inspiration from insertion sort: fast determination of position

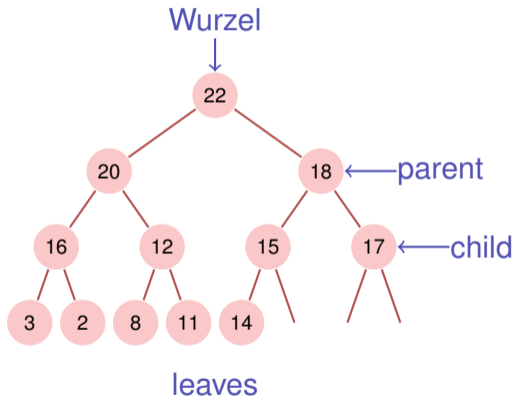
② Can we have the best of two worlds?

① Yes, but it requires some more thinking...

# [Max-]Heap<sup>6</sup>

Binary tree with the following properties

- 1 complete up to the lowest level
- 2 Gaps (if any) of the tree in the last level to the right
- 3 *Heap-Condition:*  
Max-(Min-)Heap: key of a child smaller (greater) than that of the parent node

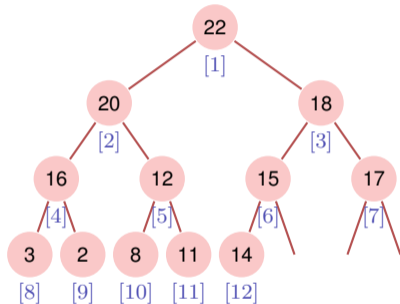
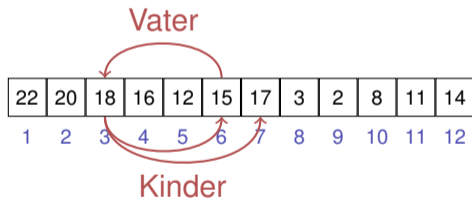


<sup>6</sup>Heap(data structure), not: as in "heap and stack" (memory allocation)

# Heap and Array

Tree  $\rightarrow$  Array:

- $\text{children}(i) = \{2i, 2i + 1\}$
- $\text{parent}(i) = \lfloor i/2 \rfloor$

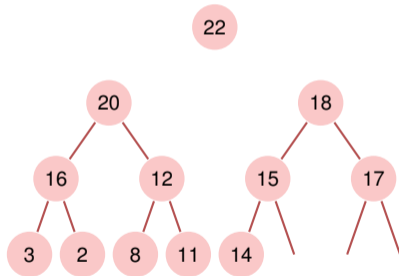


Depends on the starting index<sup>7</sup>

<sup>7</sup>For array that start at 0:  $\{2i, 2i + 1\} \rightarrow \{2i + 1, 2i + 2\}$ ,  $\lfloor i/2 \rfloor \rightarrow \lfloor (i - 1)/2 \rfloor$

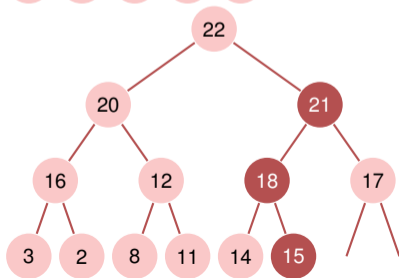
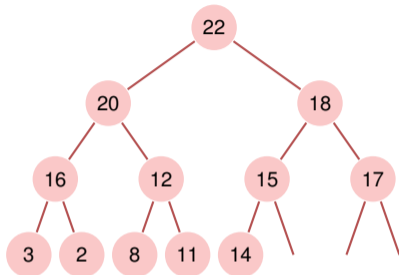
# Recursive heap structure

A heap consists of two heaps:



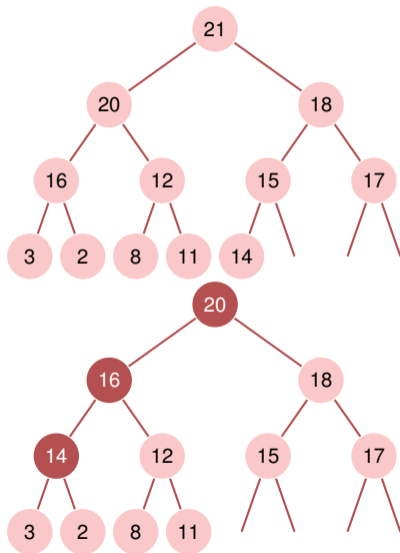
# Insert

- Insert new element at the first free position. Potentially violates the heap property.
- Reestablish heap property: climb successively
- Worst case number of operations:  $\mathcal{O}(\log n)$



# Remove the maximum

- Replace the maximum by the lower right element
- Reestablish heap property: sink successively (in the direction of the greater child)
- Worst case number of operations:  $\mathcal{O}(\log n)$





# Algorithm Sink( $A, i, m$ )

**Input :** Array  $A$  with heap structure for the children of  $i$ . Last element  $m$ .

**Output :** Array  $A$  with heap structure for  $i$  with last element  $m$ .

**while**  $2i \leq m$  **do**

$j \leftarrow 2i$ ; //  $j$  left child

**if**  $j < m$  and  $A[j] < A[j + 1]$  **then**

$j \leftarrow j + 1$ ; //  $j$  right child with greater key

**if**  $A[i] < A[j]$  **then**

        swap( $A[i], A[j]$ )

$i \leftarrow j$ ; // keep sinking

**else**

$i \leftarrow m$ ; // sinking finished

# Sort heap

$A[1, \dots, n]$  is a Heap.

While  $n > 1$

- $\text{swap}(A[1], A[n])$
- $\text{Sink}(A, 1, n - 1)$ ;
- $n \leftarrow n - 1$

swap  $\Rightarrow$

sink  $\Rightarrow$

swap  $\Rightarrow$

sink  $\Rightarrow$

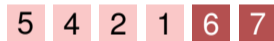
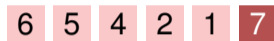
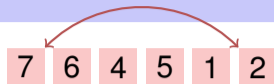
swap  $\Rightarrow$

sink  $\Rightarrow$

swap  $\Rightarrow$

sink  $\Rightarrow$

swap  $\Rightarrow$



# Heap creation

**Observation:** Every leaf of a heap is trivially a correct heap.

**Consequence:** Induction from below!

# Algorithm HeapSort( $A, n$ )

**Input :** Array  $A$  with length  $n$ .

**Output :**  $A$  sorted.

**for**  $i \leftarrow n/2$  **downto** 1 **do**

└ Sink( $A, i, n$ );

// Now  $A$  is a heap.

**for**  $i \leftarrow n$  **downto** 2 **do**

└ swap( $A[1], A[i]$ )

└ Sink( $A, 1, i - 1$ )

// Now  $A$  is sorted.

# Analysis: sorting a heap

Sink traverses at most  $\log n$  nodes. For each node 2 key comparisons.  $\Rightarrow$  sorting a heap costs is the worst case  $2 \log n$  comparisons.

Number of memory movements of sorting a heap also  $\mathcal{O}(n \log n)$ .

# Analysis: creating a heap

Calls to sink:  $n/2$ . Thus number of comparisons and movements:  
 $v(n) \in \mathcal{O}(n \log n)$ .

But mean length of sinking paths is much smaller:

$$v(n) = \sum_{h=0}^{\lfloor \log n \rfloor} \left\lceil \frac{n}{2^{h+1}} \right\rceil \cdot c \cdot h \in \mathcal{O}\left(n \sum_{h=0}^{\lfloor \log n \rfloor} \frac{h}{2^h}\right)$$

$s(x) := \sum_{k=0}^{\infty} kx^k = \frac{x}{(1-x)^2}$  ( $0 < x < 1$ ). With  $s(\frac{1}{2}) = 2$ :

$$v(n) \in \mathcal{O}(n).$$

## 8.2 Mergesort

[Ottman/Widmayer, Kap. 2.4, Cormen et al, Kap. 2.3],

# Intermediate result

Heapsort:  $\mathcal{O}(n \log n)$  Comparisons and movements.

## ❓ Disadvantages of heapsort?

- ❗ Missing locality: heapsort jumps around in the sorted array (negative cache effect).
- ❗ Two comparisons before each required memory movement.

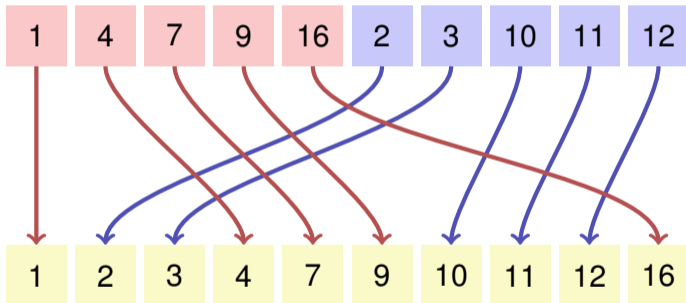


# Mergesort

Divide and Conquer!

- Assumption: two halves of the array  $A$  are already sorted.
- Minimum of  $A$  can be evaluated with two comparisons.
- Iteratively: sort the presorted array  $A$  in  $\mathcal{O}(n)$ .

# Merge



# Algorithm Merge( $A, l, m, r$ )

**Input :** Array  $A$  with length  $n$ , indexes  $1 \leq l \leq m \leq r \leq n$ .  $A[l, \dots, m]$ ,  
 $A[m + 1, \dots, r]$  sorted

**Output :**  $A[l, \dots, r]$  sortiert

1  $B \leftarrow$  new Array( $r - l + 1$ )

2  $i \leftarrow l$ ;  $j \leftarrow m + 1$ ;  $k \leftarrow 1$

3 **while**  $i \leq m$  and  $j \leq r$  **do**

4     **if**  $A[i] \leq A[j]$  **then**  $B[k] \leftarrow A[i]$ ;  $i \leftarrow i + 1$

5     **else**  $B[k] \leftarrow A[j]$ ;  $j \leftarrow j + 1$

6      $k \leftarrow k + 1$ ;

7 **while**  $i \leq m$  **do**  $B[k] \leftarrow A[i]$ ;  $i \leftarrow i + 1$ ;  $k \leftarrow k + 1$

8 **while**  $j \leq r$  **do**  $B[k] \leftarrow A[j]$ ;  $j \leftarrow j + 1$ ;  $k \leftarrow k + 1$

9 **for**  $k \leftarrow l$  **to**  $r$  **do**  $A[k] \leftarrow B[k - l + 1]$

# Correctness

Hypothesis: after  $k$  iterations of the loop in line 3  $B[1, \dots, k]$  is sorted and  $B[k] \leq A[i]$ , if  $i \leq m$  and  $B[k] \leq A[j]$  falls  $j \leq r$ .

Proof by induction:

*Base clause:* the empty array  $B[1, \dots, 0]$  is trivially sorted.

*Induction step* ( $k \rightarrow k + 1$ ):

- wlog  $A[i] \leq A[j]$ ,  $i \leq m, j \leq r$ .
- $B[1, \dots, k]$  is sorted by hypothesis and  $B[k] \leq A[i]$ .
- After  $B[k + 1] \leftarrow A[i]$   $B[1, \dots, k + 1]$  is sorted.
- $B[k + 1] = A[i] \leq A[i + 1]$  (if  $i + 1 \leq m$ ) and  $B[k + 1] \leq A[j]$  if  $j \leq r$ .
- $k \leftarrow k + 1, i \leftarrow i + 1$ : Statement holds again.

# Analysis (Merge)

## Lemma

*If: array  $A$  with length  $n$ , indexes  $1 \leq l < r \leq n$ .  $m = \lfloor (l + r)/2 \rfloor$  and  $A[l, \dots, m]$ ,  $A[m + 1, \dots, r]$  sorted.*

*Then: in the call of  $\text{Merge}(A, l, m, r)$  a number of  $\Theta(r - l)$  key movements and comparison are executed.*

Proof: straightforward (Inspect the algorithm and count the operations.)

# Mergesort

5 2 6 1 8 4 3 9

5 2 6 1 | 8 4 3 9

5 2 | 6 1 | 8 4 | 3 9

5 | 2 | 6 | 1 | 8 | 4 | 3 | 9

2 5 | 1 6 | 4 8 | 3 9

1 2 5 6 | 3 4 8 9

1 2 3 4 5 6 8 9

Split

Split

Split

Merge

Merge

Merge

# Algorithm recursive 2-way Mergesort( $A, l, r$ )

**Input :** Array  $A$  with length  $n$ .  $1 \leq l \leq r \leq n$

**Output :** Array  $A[l, \dots, r]$  sorted.

**if**  $l < r$  **then**

```
     $m \leftarrow \lfloor (l + r) / 2 \rfloor$            // middle position
    Mergesort( $A, l, m$ )                   // sort lower half
    Mergesort( $A, m + 1, r$ )               // sort higher half
    Merge( $A, l, m, r$ )                   // Merge subsequences
```

# Analysis

Recursion equation for the number of comparisons and key movements:

$$C(n) = C\left(\left\lceil \frac{n}{2} \right\rceil\right) + C\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + \Theta(n) \in \Theta(n \log n)$$



# Algorithm StraightMergesort( $A$ )

*Avoid recursion:* merge sequences of length 1, 2, 4, ... directly

**Input :** Array  $A$  with length  $n$

**Output :** Array  $A$  sorted

$length \leftarrow 1$

**while**  $length < n$  **do** // Iteriere über die Längen  $n$

$right \leftarrow 0$

**while**  $right + length < n$  **do** // Iteriere über die Teilfolgen

$left \leftarrow right + 1$

$middle \leftarrow left + length - 1$

$right \leftarrow \min(middle + length, n)$

        Merge( $A, left, middle, right$ )

$length \leftarrow length \cdot 2$

# Analysis

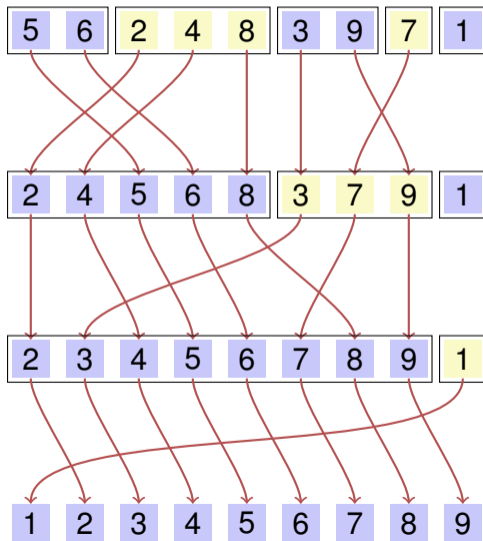
Like the recursive variant, the straight 2-way mergesort always executed a number of  $\Theta(n \log n)$  key comparisons and key movements.

# Natural 2-way mergesort

Observation: the variants above do not make use of any presorting and always execute  $\Theta(n \log n)$  memory movements.

- ② How can partially presorted arrays be sorted better?
- ① Recursive merging of previously sorted parts (*runs*) of  $A$ .

# Natural 2-way mergesort



# Algorithm NaturalMergesort( $A$ )

**Input :** Array  $A$  with length  $n > 0$

**Output :** Array  $A$  sorted

**repeat**

$r \leftarrow 0$

**while**  $r < n$  **do**

$l \leftarrow r + 1$

$m \leftarrow l$ ; **while**  $m < n$  **and**  $A[m + 1] \geq A[m]$  **do**  $m \leftarrow m + 1$

**if**  $m < n$  **then**

$r \leftarrow m + 1$ ; **while**  $r < n$  **and**  $A[r + 1] \geq A[r]$  **do**  $r \leftarrow r + 1$

            Merge( $A, l, m, r$ );

**else**

$r \leftarrow n$

**until**  $l = 1$

# Analysis

In the best case, natural merge sort requires only  $n - 1$  comparisons.

❓ Is it also asymptotically better than StraightMergesort on average?

❗ No. Given the assumption of pairwise distinct keys, on average there are  $n/2$  positions  $i$  with  $k_i > k_{i+1}$ , i.e.  $n/2$  runs. Only one iteration is saved on average.

Natural mergesort executes in the worst case and on average a number of  $\Theta(n \log n)$  comparisons and memory movements.

## 8.3 Quicksort

[Ottman/Widmayer, Kap. 2.2, Cormen et al, Kap. 7]

# Quicksort

② What is the disadvantage of Mergesort?

⚠ Requires  $\Theta(n)$  storage for merging.

② How could we reduce the merge costs?

⚠ Make sure that the left part contains only smaller elements than the right part.

② How?

⚠ Pivot and Partition!



# Quicksort (arbitrary pivot)

2 4 5 6 8 3 7 9 1

2 1 3 6 8 5 7 9 4

1 2 3 4 5 8 7 9 6

1 2 3 4 5 6 7 9 8

1 2 3 4 5 6 7 8 9

1 2 3 4 5 6 7 8 9

# Algorithm Quicksort( $A[l, \dots, r]$ )

**Input :** Array  $A$  with length  $n$ .  $1 \leq l \leq r \leq n$ .

**Output :** Array  $A$ , sorted between  $l$  and  $r$ .

**if**  $l < r$  **then**

    Choose pivot  $p \in A[l, \dots, r]$

$k \leftarrow \text{Partition}(A[l, \dots, r], p)$

    Quicksort( $A[l, \dots, k - 1]$ )

    Quicksort( $A[k + 1, \dots, r]$ )

# Reminder: algorithm Partition( $A[l, \dots, r], p$ )

**Input :** Array  $A$ , that contains the sentinel  $p$  in  $[l, r]$  at least once.

**Output :** Array  $A$  partitioned around  $p$ . Returns the position of  $p$ .

**while**  $l < r$  **do**

**while**  $A[l] < p$  **do**

$l \leftarrow l + 1$

**while**  $A[r] > p$  **do**

$r \leftarrow r - 1$

    swap( $A[l], A[r]$ )

**if**  $A[l] = A[r]$  **then**

$l \leftarrow l + 1$

// Only for keys that are not pairwise different

**return**  $l-1$

# Analysis: number comparisons

*Best case.* Pivot = median; number comparisons:

$$T(n) = 2T(n/2) + c \cdot n, T(1) = 0 \quad \Rightarrow \quad T(n) \in \mathcal{O}(n \log n)$$

*Worst case.* Pivot = min or max; number comparisons:

$$T(n) = T(n - 1) + c \cdot n, T(1) = 0 \quad \Rightarrow \quad T(n) \in \Theta(n^2)$$

# Analysis: number swaps

Result of a call to partition (pivot 3):

2 1 3 6 8 5 7 9 4

② How many swaps have taken place?

① 2. The maximum number of swaps is given by the number of keys in the smaller part.

# Analysis: number swaps

## *Intellectual game*

- Each key from the smaller part pay a coin when swapped.
- When a key has paid a coin then the domain containing the key is less or equal than half the previous size.
- Every key needs to pay at most  $\log n$  coins. But there are only  $n$  keys.

*Consequence:* there are  $\mathcal{O}(n \log n)$  key swaps in the worst case.

# Randomized Quicksort

Despite the worst case running time of  $\Theta(n^2)$ , quicksort is used practically very often.

Reason: quadratic running time unlikely if the choice of the pivot and the presorting is not very disadvantageous.

Avoidance: randomly choose pivot. Draw uniformly from  $[l, r]$ .

# Analysis (randomized quicksort)

Expected number of compared keys with input length  $n$ :

$$T(n) = (n - 1) + \frac{1}{n} \sum_{k=1}^n (T(k - 1) + T(n - k)), \quad T(0) = T(1) = 0$$

Claim  $T(n) \leq 4n \log n$ .

Proof by induction:

*Base clause* straightforward for  $n = 0$  (with  $0 \log 0 := 0$ ) and for  $n = 1$ .

*Hypothesis:*  $T(n) \leq 4n \log n$  für ein  $n$ .

*Induction step:*  $(n - 1 \rightarrow n)$



# Analysis (randomized quicksort)

$$\begin{aligned}T(n) &= n - 1 + \frac{2}{n} \sum_{k=0}^{n-1} T(k) \stackrel{H}{\leq} n - 1 + \frac{2}{n} \sum_{k=0}^{n-1} 4k \log k \\&= n - 1 + \sum_{k=1}^{n/2} 4k \underbrace{\log k}_{\leq \log n - 1} + \sum_{k=n/2+1}^{n-1} 4k \underbrace{\log k}_{\leq \log n} \\&\leq n - 1 + \frac{8}{n} \left( (\log n - 1) \sum_{k=1}^{n/2} k + \log n \sum_{k=n/2+1}^{n-1} k \right) \\&= n - 1 + \frac{8}{n} \left( (\log n) \cdot \frac{n(n-1)}{2} - \frac{n}{4} \left( \frac{n}{2} + 1 \right) \right) \\&= 4n \log n - 4 \log n - 3 \leq 4n \log n\end{aligned}$$

# Analysis (randomized quicksort)

## Theorem

*On average randomized quicksort requires  $\mathcal{O}(n \cdot \log n)$  comparisons.*

# Practical considerations

Worst case recursion depth  $n - 1$ <sup>8</sup>. The also memory consumption of  $\mathcal{O}(n)$ .

Can be avoided: recursion only on the smaller part. Then guaranteed  $\mathcal{O}(\log n)$  worst case recursion depth and memory consumption.

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<sup>8</sup>stack overflow possible!

# Quicksort with logarithmic memory consumption

**Input :** Array  $A$  with length  $n$ .  $1 \leq l \leq r \leq n$ .

**Output :** Array  $A$ , sorted between  $l$  and  $r$ .

**while**  $l < r$  **do**

    Choose pivot  $p \in A[l, \dots, r]$

$k \leftarrow \text{Partition}(A[l, \dots, r], p)$

**if**  $k - l < r - k$  **then**

        Quicksort( $A[l, \dots, k - 1]$ )

$l \leftarrow k + 1$

**else**

        Quicksort( $A[k + 1, \dots, r]$ )

$r \leftarrow k - 1$

The call of Quicksort( $A[l, \dots, r]$ ) in the original algorithm has moved to iteration (tail recursion!): the if-statement became a while-statement.

# Practical considerations.

Practically the pivot is often the median of three elements. For example:  $\text{Median3}(A[l], A[r], A[\lfloor l + r/2 \rfloor])$ .

There is a variant of quicksort that requires only constant storage. Idea: store the old pivot at the position of the new pivot.