

# Algorithms and Data Structures

## A5. Runtime Analysis: Introduction and Selection Sort

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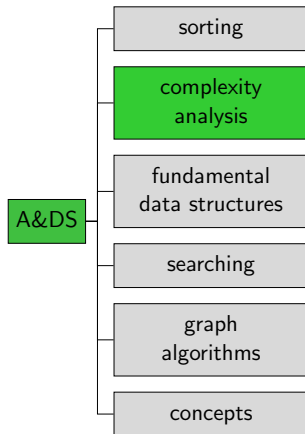
## A5.1 Runtime Analysis in General

## A5.2 Example: Selection Sort

## A5.3 Summary

# A5.1 Runtime Analysis in General

# Content of the Course



# Exact Runtime Analysis Unrealistic

- ▶ **Would be nice:** formula that determines for a specific input how long the computation will take.
- ▶ **Exact runtime prediction is hard** because of too many influencing factors.
  - ▶ Speed and architecture of the computer
  - ▶ Programming language
  - ▶ Compiler version
  - ▶ Current load (what else is running?)
  - ▶ Caching behavior

We neither can nor want to consider all this in a formula.

# Runtime Analysis: 1st Simplification

## Don't measure time but count operations

What is an operation?

- ▶ Ideally: one line of machine code or – even more precisely – one processor cycle
- ▶ Instead: constant-time operations
  - ▶ Constant time: running time independent of input.
  - ▶ Ignore runtime differences of different operations.
  - ▶ E.g. addition, assignments, branching, function call.
  - ▶ **Roughly**: operation = one line of code.
  - ▶ **But**: also consider what's behind it  
e.g. steps inside the called function.

Running time roughly proportional to the number of operations

# Runtime Analysis: 2nd Simplification

Don't count exactly but use bounds!

- ▶ Mostly considering upper bounds  
How many steps does it take at most?
- ▶ Sometimes also lower bound  
How many steps are at least executed?

„running time“ for bound on number of executed operations

# Runtime Analysis: 3rd Simplification

## Bounds only relative to the input size

- ▶  $T(n)$ : running time for input of size  $n$
- ▶ For adaptive algorithms we distinguish
  - ▶ Best case  
running time for best possible input of size  $n$
  - ▶ Worst case  
running time for worst possible input of size  $n$
  - ▶ Average case  
average running time over all inputs of size  $n$



# Cost Models

Sometimes: analysis wrt. **cost model**

- ▶ Identify fundamental operations for the algorithm class  
e.g. for sorting algorithms.
  - ▶ Key comparison
  - ▶ Swap of two elements or movement of an element
- ▶ Analyze number of these operations.

# Example from C++ Reference

function template

**std::sort**

<algorithm>

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```

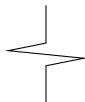
default (1) template <class RandomAccessIterator>
              void sort (RandomAccessIterator first, RandomAccessIterator last);
custom (2)  template <class RandomAccessIterator, class Compare>
              void sort (RandomAccessIterator first, RandomAccessIterator last, Compare comp);
  
```

## Sort elements in range

Sorts the elements in the range `[first, last)` into ascending order.

The elements are compared using operator< for the first version, and *comp* for the second.

Equivalent elements are not guaranteed to keep their original relative order (see `stable_sort`).



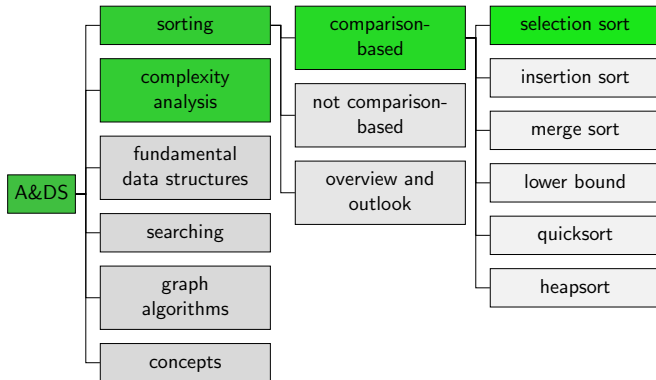
## Complexity

On average, linearithmic in the `distance` between *first* and *last*: Performs approximately  $N \cdot \log_2(N)$  (where  $N$  is this distance) comparisons of elements, and up to that many element swaps (or moves).

<http://www.cplusplus.com/reference/algorithm/sort/>

## A5.2 Example: Selection Sort

# Content of the Course



# Selection Sort: Algorithm

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```
1 def selection_sort(array):
2     n = len(array)
3     for i in range(n - 1): # i = 0, ..., n-2
4         # find index of minimum element at positions i, ..., n-1
5         min_index = i
6         for j in range(i + 1, n): # j = i+1, ..., n-1
7             if array[j] < array[min_index]:
8                 min_index = j
9         # swap element at position i with minimum element
10        array[i], array[min_index] = array[min_index], array[i]
```

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# Selection Sort with Cost Model

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```
1 def selection_sort(array):
2     n = len(array)
3     for i in range(n - 1): # i = 0, ..., n-2
4         # find index of minimum element at positions i, ..., n-1
5         min_index = i
6         for j in range(i + 1, n): # j = i+1, ..., n-1
7             if array[j] < array[min_index]:
8                 min_index = j
9         # swap element at position i with minimum element
10        array[i], array[min_index] = array[min_index], array[i]
```

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→  $n-1$  swaps of two elements (“linear”)

→  $0.5(n-1)n$  key comparisons (“quadratic”)

# Selection Sort: Analysis I

We show:  $T(n) \leq c' \cdot n^2$  for  $n \geq 1$  and some constant  $c'$

- ▶ Outer loop (3-10) and inner loop (6-8)
- ▶ Number of operations for each iteration of the outer loop:
  - ▶ Constant  $a$  for no. of operations in lines 7 and 8
  - ▶ Constant  $b$  for no. of operations in lines 5 and 10

$i$	# operations
0	$a(n-1) + b$
1	$a(n-2) + b$
	...
n-2	$a \cdot 1 + b$

- ▶ Total:  $T(n) = \sum_{i=0}^{n-2} (a(n - (i + 1)) + b)$

## Selection Sort: Analysis II

$$\begin{aligned}T(n) &= \sum_{i=0}^{n-2} (a(n - (i + 1)) + b) \\&= \sum_{i=1}^{n-1} (a(n - i) + b) \\&= a \sum_{i=1}^{n-1} (n - i) + b(n - 1) \\&= 0.5a(n - 1)n + b(n - 1) \\&\leq 0.5an^2 + b(n - 1) \\&\leq 0.5an^2 + b(n - 1)n \\&\leq 0.5an^2 + bn^2 \\&= (0.5a + b)n^2\end{aligned}$$

$\Rightarrow$  with  $c' = (0.5a + b)$  it holds for  $n \geq 1$  that  $T(n) \leq c' \cdot n^2$



# Selection Sort: Analysis III

Too generous bound?

We show for  $n \geq 2$ :  $T(n) \geq c \cdot n^2$  for some constant  $c$

$$\begin{aligned} T(n) &= \dots = 0.5a(n-1)n + b(n-1) \\ &\geq 0.5a(n-1)n \\ &\geq 0.25an^2 \quad (n-1 \geq 0.5n \text{ for } n \geq 2) \end{aligned}$$

$\Rightarrow$  with  $c = 0.25a$  it holds for  $n \geq 2$  that  $T(n) \geq c \cdot n^2$

## Theorem

Selection sort has **quadratic running time**, i.e., there are constants  $c > 0$ ,  $c' > 0$ ,  $n_0 > 0$  such that for  $n \geq n_0$ :  $cn^2 \leq T(n) \leq c'n^2$ .

## Selection Sort: Analysis IV

**Quadratic running time:** twice as large input, fourfold running time

What does this mean in practice?

- ▶ Assumption:  $c = 1$ , one operation takes on average  $10^{-8}$  sec.
- ▶ With 1000 elements, we wait  
 $10^{-8} \cdot (10^3)^2 = 10^{-8} \cdot 10^6 = 10^{-2} = 0.02$  seconds.
- ▶ With 10 thousand elements, we wait  
 $10^{-8} \cdot (10^4)^2 = 1$  second.
- ▶ With 100 thousand elements  $10^{-8} \cdot (10^5)^2 = 100$  seconds.
- ▶ With 1 million elements  $10^{-8} \cdot (10^6)^2$  seconds = 2.77 hours.
- ▶ With 1 billion elements  $10^{-8} \cdot (10^9)^2$  seconds = 317 years.  
1 billion numbers with 4 bytes/number are „only“ 4 GB.

**Quadratic running time problematic for large inputs**

## A5.3 Summary

# Summary

- ▶ Runtime analysis considers **bounds** on the **number of executed operations**.
  - ▶ We don't count exactly.
  - ▶ We ignore how long each operation actually takes.
  - ▶ Running time should be roughly proportional to the number of operations.
- ▶ **Selection sort** has **quadratic running time** and performs a linear number of swaps and a quadratic number of key comparisons.