## 4 Modelling Issues

#### 1 What do you measure?

- Memory requirement
- Running time
- Number of comparisons
- Number of multiplications
- Number of hard-disc accesses
- Program size
- Power consumption



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# 4 Modelling Issues

#### Input length

The theoretical bounds are usually given by a function  $f: \mathbb{N} \to \mathbb{N}$  that maps the input length to the running time (or storage space, comparisons, multiplications, program size etc.).

The input length may e.g. be

- the size of the input (number of bits)
- ▶ the number of arguments

#### Example 1

Suppose n numbers from the interval  $\{1,\ldots,N\}$  have to be sorted. In this case we usually say that the input length is n instead of e.g.  $n\log N$ , which would be the number of bits required to encode the input.

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## 4 Modelling Issues

#### How do you measure?

- Implementing and testing on representative inputs
  - How do you choose your inputs?
  - May be very time-consuming.
  - Very reliable results if done correctly.
  - Results only hold for a specific machine and for a specific set of inputs.
- ▶ Theoretical analysis in a specific model of computation.
  - Gives asymptotic bounds like "this algorithm always runs in time  $\mathcal{O}(n^2)$ ".
  - Typically focuses on the worst case.
  - Can give lower bounds like "any comparison-based sorting algorithm needs at least  $\Omega(n \log n)$  comparisons in the worst case".



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# **Model of Computation**

## How to measure performance

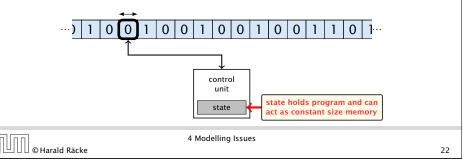
- 1. Calculate running time and storage space etc. on a simplified, idealized model of computation, e.g. Random Access Machine (RAM), Turing Machine (TM), . . .
- 2. Calculate number of certain basic operations: comparisons, multiplications, harddisc accesses, ...

Version 2. is often easier, but focusing on one type of operation makes it more difficult to obtain meaningful results.

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## **Turing Machine**

- Very simple model of computation.
- ▶ Only the "current" memory location can be altered.
- Very good model for discussing computability, or polynomial vs. exponential time.
- ▶ Some simple problems like recognizing whether input is of the form *xx*, where *x* is a string, have quadratic lower bound.
- ⇒ Not a good model for developing efficient algorithms.



## **Random Access Machine (RAM)**

#### Operations

- ▶ input operations (input tape  $\rightarrow R[i]$ )
  - ► READ *i*
- ▶ output operations  $(R[i] \rightarrow \text{output tape})$ 
  - ▶ WRITE *i*
- register-register transfers
  - ightharpoonup R[i] := R[i]
  - ightharpoonup R[j] := 4
- indirect addressing
  - ▶ R[j] := R[R[i]] loads the content of the R[i]-th register into the j-th register
  - ▶ R[R[i]] := R[j]
     loads the content of the j-th into the R[i]-th register

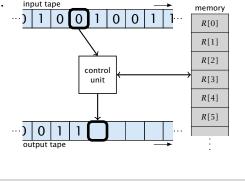
## **Random Access Machine (RAM)**

- ► Input tape and output tape (sequences of zeros and ones; unbounded length).
- ▶ Memory unit: infinite but countable number of registers  $R[0], R[1], R[2], \ldots$
- ► Registers hold integers.
- Indirect addressing.

Note that in the picture on the right

the tapes are one-directional, and that

a READ- or WRITE-operation always ad-



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vances its tape.

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## **Random Access Machine (RAM)**

#### **Operations**

- branching (including loops) based on comparisons
  - jump x jumps to position x in the program; sets instruction counter to x; reads the next operation to perform from register R[x]
  - jump z x R[i]jump to x if R[i] = 0if not the instruction of

if not the instruction counter is increased by 1;

- jumpi i
  jump to R[i] (indirect jump);
- ▶ arithmetic instructions: +, -, ×, /
  - ► R[i] := R[j] + R[k];R[i] := -R[k];

The jump-directives are very close to the jump-instructions contained in the assembler language of real machines.



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# **Model of Computation**

- uniform cost modelEvery operation takes time 1.
- logarithmic cost model

The cost depends on the content of memory cells:

- ► The time for a step is equal to the largest operand involved:
- ► The storage space of a register is equal to the length (in bits) of the largest value ever stored in it.

**Bounded word RAM model:** cost is uniform but the largest value stored in a register may not exceed  $2^w$ , where usually

$$w = \log_2 n$$
.

The latter model is quite realistic as the word-size of a standard computer that handles a problem of size n must be at least  $\log_2 n$  as otherwise the computer could either not store the problem instance or not address all its memory.



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#### There are different types of complexity bounds:

best-case complexity:

$$C_{\rm bc}(n) := \min\{C(x) \mid |x| = n\}$$

Usually easy to analyze, but not very meaningful.

worst-case complexity:

$$C_{WC}(n) := \max\{C(x) \mid |x| = n\}$$

Usually moderately easy to analyze; sometimes too pessimistic.

average case complexity:

$$C_{\text{avg}}(n) := \frac{1}{|I_n|} \sum_{|x|=n} C(x)$$

more general: probability measure  $\mu$ 

$$C_{\operatorname{avg}}(n) := \sum_{x \in I_n} \mu(x) \cdot C(x)$$

C(x) cost of instance x input length of instance x set of instances of length n

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## 4 Modelling Issues

#### Example 2

#### **Algorithm 1** RepeatedSquaring(n)

1:  $r \leftarrow 2$ ;

2: for  $i = 1 \rightarrow n$  do

3:  $r \leftarrow r^2$ 

4: return  $\gamma$ 

- running time:
  - uniform model: n steps
  - logarithmic model:  $1 + 2 + 4 + \cdots + 2^n = 2^{n+1} 1 = \Theta(2^n)$
- space requirement:
  - uniform model:  $\mathcal{O}(1)$
  - ▶ logarithmic model:  $\mathcal{O}(2^n)$



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#### There are different types of complexity bounds:

amortized complexity:

The average cost of data structure operations over a worst case sequence of operations.

randomized complexity:

The algorithm may use random bits. Expected running time (over all possible choices of random bits) for a fixed input

x. Then take the worst-case over all x with |x| = n.

# 4 Modelling Issues Bibliography [MS08] Kurt Mehlhorn, Peter Sanders: Algorithms and Data Structures — The Basic Toolbox, Springer, 2008 [CLRS90] Thomas H. Cormen, Charles E. Leiserson, Ron L. Rivest, Clifford Stein: Introduction to algorithms (3rd ed.), McGraw-Hill, 2009 Chapter 2.1 and 2.2 of [MS08] and Chapter 2 of [CLRS90] are relevant for this section. 4 Modelling Issues © Harald Räcke 29