

APPLIED ALGORITHMIC \$ APPLIED  
APPLIED ALGORITHMIC \$  
CS \$ APPLIED ALGORITHMIC  
DALGORITHMIC \$ APPLIE  
ED ALGORITHMIC \$ APPLI  
GORITHMIC \$ APPLIEDAL  
HMICS \$ APPLIEDALGORIT  
ICS \$ APPLIEDALGORITHM  
IEDALGORITHMIC \$ APPL  
ITHMIC \$ APPLIEDALGOR  
LGOITHMIC \$ APPLIEDA  
LIEDALGORITHMIC \$ AP  
MICS \$ APPLIEDALGORITH  
ORITHMIC \$ APPLIE  
PLIEDALGORITHMIC \$ AP  
PPLIEDALGORITHMIC \$ A  
RITHMIC \$ APPLIEDALGO  
S \$ APPLIEDALGORITHMIC  
THMIC \$ APPLIEDALGORI

# 1

# Machines & Models

27 January 2020

Sebastian Wild

# Outline

## 1 Machines & Models

- 1.1 Algorithm analysis
- 1.2 The RAM Model
- 1.3 Asymptotics & Big-Oh

# What is an algorithm?

An algorithm is a sequence of instructions.

think: recipe

More precisely:

e. g. Java program

1. mechanically executable  
↔ no “common sense” needed
2. finite description  $\neq$  finite computation!
3. solves a problem, i. e., a class of problem instances

$x + y$ , not only  $17 + 4$

typical example: *bubblesort*

not a specific program but underlying idea



# What is a data structure?

A data structure is

1. a rule for encoding data  
(in computer memory), plus
2. algorithms to work with it  
(queries, updates, etc.)

typical example: binary search tree



## **1.1 Algorithm analysis**

# Good algorithms

**Our goal:** Find good (best?) algorithms and data structures for a task.

Good “usually” means

can be complicated in distributed systems

- ▶ fast running *time*
- ▶ moderate memory *space* usage

*Algorithm analysis* is a way to

- ▶ compare different algorithms,
- ▶ predict their performance in an application

# Running time experiment

Why not simply run and time it?

- ▶ results only apply to
  - ▶ single *test* machine
  - ▶ tested inputs
  - ▶ tested implementation
  - ▶ ...

≠ *universal truths*

- ▶ instead: consider and analyze algorithms on an abstract machine
  - ↪ provable statements for model
  - ↪ testable model hypotheses

survives Pentium 4



↪ Need precise model of machine (costs), input data and algorithms.

# Data Models

Algorithm analysis typically uses one of the following simple data models:

- ▶ **worst-case performance:**  
consider the *worst* of all inputs as our cost metric
- ▶ **best-case performance:**  
consider the *best* of all inputs as our cost metric
- ▶ **average-case performance:**  
consider the average/expectation of a *random* input as our cost metric

Usually, we apply the above for *inputs of same size  $n$* .

↪ performance is only a **function of  $n$** .



## **1.2 The RAM Model**

# Machine models

The machine model decides

- ▶ what algorithms are possible
- ▶ how they are described (= programming language)
- ▶ what an execution *costs*

**Goal:** Machine model should be detailed and powerful enough to reflect actual machines, abstract enough to unify architectures, simple enough to analyze.

# Random Access Machines

## Random access machine (RAM)

more detail in §2.2 of *Sequential and Parallel Algorithms and Data Structures*  
by Sanders, Mehlhorn, Dietzfelbinger, Dementiev

- ▶ unlimited *memory*  $\text{MEM}[0], \text{MEM}[1], \text{MEM}[2], \dots$
- ▶ fixed number of *registers*  $R_1, \dots, R_r$  (say  $r = 100$ )
- ▶ every memory cell  $\text{MEM}[i]$  and register  $R_i$  stores a  $w$ -bit integer, i. e., a number in  $[0..2^w - 1]$   
 $w$  is the word width; typically  $2^w \approx n$
- ▶ Instructions:
  - ▶ load & store:  $R_i := \text{MEM}[R_j]$      $\text{MEM}[R_j] := R_i$
  - ▶ operations on registers:  $R_k := R_i + R_j$  (arithmetic is *modulo*  $2^w$ !)  
also  $R_i - R_j, R_i \cdot R_j, R_i \text{ div } R_j, R_i \bmod R_j$   
C-style operations (bitwise and/or/xor, left/right shift)
  - ▶ conditional and unconditional jumps
- ▶ cost: number of executed instructions

we will see further models later

~> The RAM is the standard model for sequential computation.

# Pseudocode

Typical simplifications for convenience:

- ▶ more abstract *pseudocode* to specify algorithms  
code that humans understand (easily)
- ▶ count *dominant operations* (e. g. array accesses) instead of all operations

In both cases: can go to full detail if needed.

## 1.3 Asymptotics & Big-Oh

# Why asymptotics?

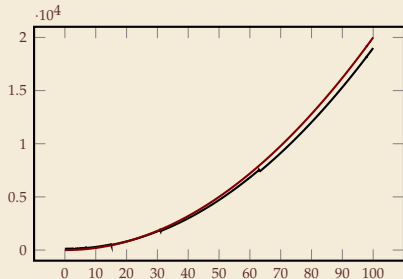
Algorithm analysis focuses on (the limiting behavior for infinitely) **large** inputs.

- ▶ abstracts from unnecessary detail
- ▶ simplifies analysis
- ▶ often necessary for sensible comparison

Asymptotics = approximation around  $\infty$

**Example:** Consider a function  $f(n)$  given by

$$2n^2 - 3n \lfloor \log_2(n+1) \rfloor + 7n - 3 \lfloor \log_2(n+1) \rfloor + 120 \sim 2n^2$$



# Asymptotic tools

- ▶ “Tilde Notation:”  $f(n) \sim g(n)$  if, and only if  $\text{iff } \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 1$   
„ $f$  and  $g$  are asymptotically equivalent”

- ▶ “Big-Oh Notation:”  $f(n) \in O(g(n))$  also write '=' instead  $\text{iff } \left| \frac{f(n)}{g(n)} \right|$  is bounded for  $n \geq n_0$

need supremum since limit might not exist!

$$\text{iff } \limsup_{n \rightarrow \infty} \left| \frac{f(n)}{g(n)} \right| < \infty$$

## Variants:

- ▶  $f(n) = \Omega(g(n))$  “Big-Omega”  $\text{iff } g(n) = O(f(n))$
- ▶  $f(n) = \Theta(g(n))$  “Big-Theta”  $\text{iff } f(n) = O(g(n))$  **and**  $f(n) = \Omega(g(n))$

- ▶ “Little-Oh Notation:”  $f(n) = o(g(n))$   $\text{iff } \lim_{n \rightarrow \infty} \left| \frac{f(n)}{g(n)} \right| = 0$

$$f(n) = \omega(g(n)) \text{ if } \lim = \infty$$

# Asymptotics – Example 1

Basic examples:

- ▶  $20n^3 + 10n \lg(n) + 5 \sim 20n^3 = \Theta(n^3)$
- ▶  $3 \lg(n^2) + \lg(\lg(n)) = \Theta(\log n)$
- ▶  $10^{100} = O(1)$

Frequent orders of growth

- ▶ logarithmic  $\Theta(\log n)$       Note:  $a, b > 0$  constants  $\rightsquigarrow \Theta(\log_a(n)) = \Theta(\log_b(n))$
- ▶ linear  $\Theta(n)$
- ▶ linearithmic  $\Theta(n \log n)$
- ▶ quadratic  $\Theta(n^2)$
- ▶ polynomial  $O(n^c)$  for constant  $c$
- ▶ exponential  $O(c^n)$  for constant  $c$       Note:  $a > b > 0$  constants  $\rightsquigarrow b^n = o(a^n)$



## Asymptotics – Example 2

### Square-and-multiply algorithm

for computing  $x^m$  with  $m \in \mathbb{N}$

Inputs:

- ▶  $m$  as binary number (array of bits)
- ▶  $n = \#$ bits in  $m$
- ▶  $x$  a floating-point number

---

```
1 double pow(double base, boolean[] exponentBits) {
2     double res = 1;
3     for (boolean bit : exponentBits) {
4         res *= res;
5         if (bit) res *= base;
6     }
7     return res;
8 }
```

---

▶ Cost:  $C = \#$  multiplications

▶  $C = n$  (line 4) +  $\#$ one-bits binary representation of  $m$  (line 5)

$\rightsquigarrow n \leq C \leq 2n$