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Machines & Models

16 February 2021

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 ≠ 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



input \rightarrow algo \rightarrow output

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

- ▶ fast running time can be complicated in distributed systems
- ▶ 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

Clicker Question



What is the cost of *adding* two d -digit integers?
For example, for $d = 5$, what is $45\,235 + 91\,342$?

- A** constant time
- B** logarithmic in d
- C** proportional to d
- D** quadratic in d
- E** no idea what you are talking about

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Clicker Question



What is the cost of *adding* two d -digit integers?

For example, for $d = 5$, what is $45\,235 + 91\,342$?

- A** constant time ✓ if fit into an int (64 bit)
- B** ~~logarithmic in d~~
- C** proportional to d ✓ if #bits $\Rightarrow 64$
(BigInteger)
- D** ~~quadratic in d~~
- E** no idea what you are talking about ✓

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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 $r = 100$)
- ▶ memory cells $\text{MEM}[i]$ and registers R_i store w -bit integers, i. e., numbers in $[0..2^w - 1]$
 w is the word width/size; typically $w \propto \lg n \rightsquigarrow 2^w \approx n$ $w = 64 \quad 128$
- ▶ Instructions:
 - ▶ load & store: $R_i := \text{MEM}[R_j] \quad \text{MEM}[R_j] := R_i$
 - ▶ 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

machine grows with input n

or $R_i = 0$

we will see further models later

\rightsquigarrow 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.

restart at noon

.. 1 more min

1.3 Asymptotics & Big-Oh

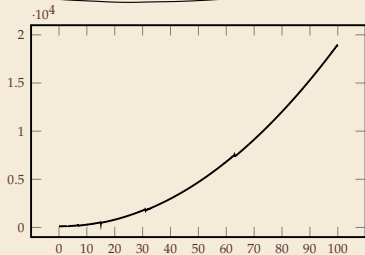
Why asymptotics?

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 ∞

Example: Consider a function $f(n)$ given by
 $2n^2 - 3n \lfloor \log_2(n+1) \rfloor + 7n - 3 \lfloor \log_2(n+1) \rfloor + 120$



Why asymptotics?

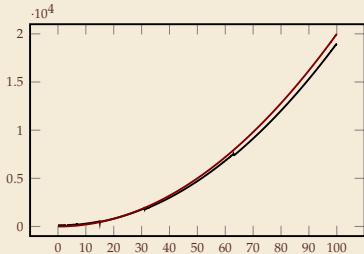
Algorithm analysis focuses on (the limiting behavior for infinitely) **large inputs**.

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

Asymptotics = approximation around ∞

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

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



Asymptotic tools – Formal & definitive definition

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

Asymptotic tools – Formal & definitive definition

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

- ▶ “Big-Oh Notation:” $f(n) \in O(g(n))$ ^{also write ‘=’ instead} $\iff \left| \frac{f(n)}{g(n)} \right| < \infty$ is bounded for $n \geq n_0$ (g(n) could be 0)

need supremum since limit might not exist!

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

Variants:

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

Asymptotic tools – Formal & definitive definition

▶ “Tilde Notation:” $f(n) \sim g(n)$ if, and only if $\iff \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 1$
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▶ “Big-Oh Notation:” $f(n) \in O(g(n))$ also write ‘=’ instead $\iff \left| \frac{f(n)}{g(n)} \right|$ is bounded for $n \geq n_0$

need supremum since limit might not exist!
 $\iff \limsup_{n \rightarrow \infty} \left| \frac{f(n)}{g(n)} \right| < \infty$

Variants: “Big-Omega”

▶ $f(n) = \Omega(g(n))$ $\iff g(n) = O(f(n))$

▶ $f(n) = \Theta(g(n))$ $\iff f(n) = O(g(n))$ **and** $f(n) = \Omega(g(n))$

“Big-Theta”

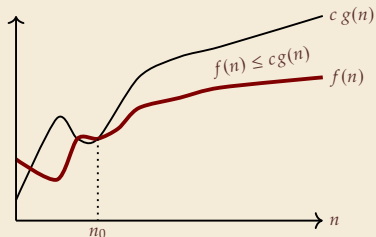
▶ “Little-Oh Notation:” $f(n) = o(g(n))$ $\iff \lim_{n \rightarrow \infty} \left| \frac{f(n)}{g(n)} \right| = 0$

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

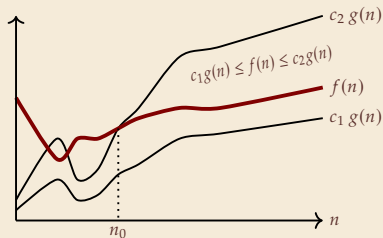
Asymptotic tools – Intuition

- ▶ $f(n) = O(g(n))$: $f(n)$ is **at most** $g(n)$ up to constant factors and for sufficiently large n

$$f \leq g$$



- ▶ $f(n) = \Theta(g(n))$: $f(n)$ is **equal to** $g(n)$ up to constant factors and for sufficiently large n



Plots can be misleading!

Example ↗

Clicker Question



Assume $f(n) \in O(g(n))$. What can we say about $g(n)$?

- A** $g(n) = O(f(n))$
- B** $g(n) = \Omega(f(n))$
- C** $g(n) = \Theta(f(n))$
- D** Nothing (it depends on f and g)

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Clicker Question



Assume $f(n) \in O(g(n))$. What can we say about $g(n)$?

A ~~$g(n) = O(f(n))$~~

B $g(n) = \Omega(f(n))$ ✓

C ~~$g(n) = \Theta(f(n))$~~

D ~~Nothing (it depends on f and g)~~

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Clicker Question



Assume $f(n) \in O(g(n))$. What can we say about $g(n)$?

- A** ~~$g(n) = O(f(n))$~~
- B** $g(n) = \Omega(f(n))$ ✓ (if $f(n) \neq 0$)
- C** ~~$g(n) = \Theta(f(n))$~~
- D** Nothing (it depends on f and g) ✓

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Asymptotics – Example 1

Basic examples:

▶ $20n^3 + 10n \ln(n) + 5 \sim 20n^3 = \Theta(n^3)$

▶ $\frac{3 \lg(n^2) + \lg(\lg(n))}{9} = \Theta(\log n)$

▶ $10^{100} = O(1)$

$$\frac{3.2 \lg(u) + \lg(\lg(u))}{\lg n} \xrightarrow{n \rightarrow \infty} 6$$

$$\begin{aligned} & \lim_{n \rightarrow \infty} \frac{20n^3 + 10n \ln n + 5}{20n^3} \\ &= \lim_{n \rightarrow \infty} \frac{20n^3}{20n^3} + \frac{10n \ln n + 5}{20n^3} \\ &= \overset{20}{1} + \underbrace{\lim_{n \rightarrow \infty} \frac{10n \ln n + 5}{20n^3}}_{=0} \end{aligned}$$

$20 < \infty \Rightarrow$

$f(n) = O(n^3)$

Use *wolfram alpha* to compute/check limits.

Clicker Question



Is $(\sin(n) + 2)n^2 = \Theta(n^2)$?

A Yes

B No

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Clicker Question



Is $(\sin(n) + 2)n^2 = \Theta(n^2)$?

A Yes ✓

B ~~No~~

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Asymptotics – Frequently used facts

▶ Rules:

▶ $c \cdot f(n) = \Theta(f(n))$ for constant $c \neq 0$

▶ $\Theta(f + g) = \Theta(\max\{f, g\})$ largest summand determines order of growth

$\approx \Theta$ class

▶ Frequently used 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)
- ▶ $n = \# \text{bits in } m$
- ▶ x a floating-point number

▶ Cost: $C = \# \text{multiplications}$

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

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

```
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 }
```

Clicker Question



We showed $n \leq C(n) \leq 2n$; what is the most precise asymptotic approximation for $C(n)$ that we can make?

Write e. g. $O(n^2)$ for $O(n^2)$ or $\Theta(\sqrt{n})$ for $\Theta(\sqrt{n})$.

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Asymptotics – Example 2

Square-and-multiply algorithm
for computing x^m with $m \in \mathbb{N}$

Inputs:

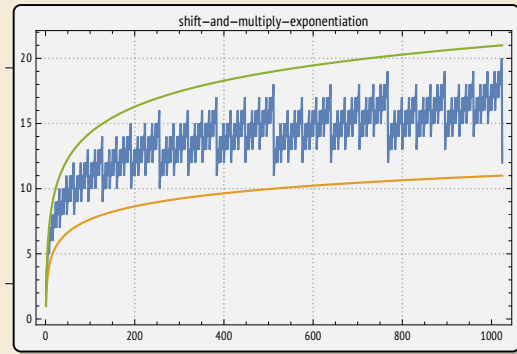
- ▶ m as binary number (array of bits)
- ▶ $n = \#$ bits in m
- ▶ x a floating-point number

▶ Cost: $C = \#$ multiplications

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

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

$\rightsquigarrow C = \Theta(n) = \Theta(\log m)$



Note: Often, you can pretend Θ is “like \sim with an unknown constant”
but in this case, no such constant exists!