Advanced Algorithms 1.1. Algorithm Design and Analysis

Henry Förster

Chair of Efficient Algorithms

Algorithmic Revolution



KIND OF NEW: Machines learn algorithms based on information without explicit instructions



Even difficult problems can be solved that way!



Application scenarios where the definition is fuzzy can be solved! (E.g., the machine should provide a "human-like" response)



Why do we still discuss classic algorithms?

Shooting Sparrows with Cannons



- ► Efficient algorithms...
 - ▶ solve *problems directly*
 - require less computational ressources
 - ► can run *on restrictive hardware*
 - ► can be *cheaper to operate*
 - require *no expensive training* (cost, emissions, ...)
- ► Also, it is fun to solve puzzles!
 - ► And understand *why* the solution works!

Welcome to Advanced Algorithms!

- ► Hi, I'm *Henry*!
 - ▶ PhD and PostDoc at University of Tübingen
 - ► Research Focus: Efficient Algorithms for Visualizing Relational Data,
 Properties of Geometric and Topological Graphs,
 Evaluating Visualization Techniques
 - ► This course is *not* (completely) ad-hoc: taught similar classes in Tübingen
- ► How about you?
- Disclaimer: Today is my very first day here!
 - ► moodle will be available later today
 - Exam date will be announced later today on moodle and Thursday in class

Course Organization

- ► *Lecture:* Tuesday 10:15 11:45
- Again, there will be a *moodle*
 - ▶ Slides, assignment sheets, forum, announcements, ...
- Assignment sheets
 - Available after lecture (this week later today when moodle is running)
 - ► Voluntary, but graded for feedback
 - ► Hand-in via moodle until Tuesday before lecture, possible in groups
- ▶ *Discussion Session:* Thursday, 10:15 11:45 (same room)
 - ► Per request: Solutions for assignment sheet from last week, questions regarding the lecture, guidance on current assignment sheet
 - ► This week: Announcements regarding moodle and exam date
- ► Written Exam: Date TBA, 120 mins
 - Cheat sheet allowed (A4, both sides)

Course Contents

I. Algorithm Design and Analysis	 Recap: Run time and Correctness, Landau Notation, Divide and Conquer Advanced Design Concepts: Dynamic Programming, Greedy Amortized Analysis
II. Advanced Data Structures	1. Fibonacci-Heaps 2. Union-Find Data Structure
III. Graph Algorithms	1. Maximum Flow & Maximum Matching 2. Push-Relabel Algorithm
IV. Geometric Algorithms	1. Sweep Line Method 2. Randomized Incremental Construction
V. Linear Programming	1. Properties and Duality 2. Algorithms for Linear Programming
VI. Approximation Algorithms	1. APX 2. PTAS
Q VII. Parameterized Algorithms	1. FPT 2. W[1]

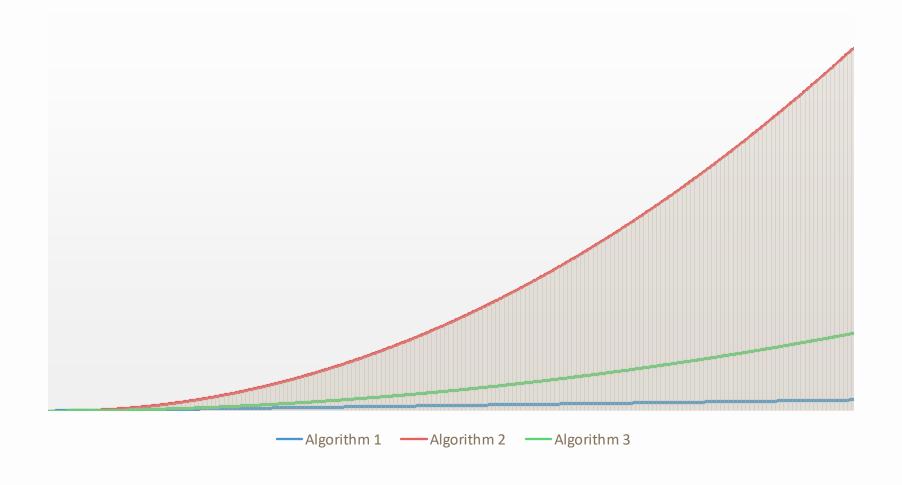
Goals for Today

At the end of this lecture, you can ...

- prove the *correctness* of algorithms using suitable *invariants* and induction
- provide bounds on run times using the Landau notation
- resolve recurrences by applying the master theorem
- □ design *recursive* and *divide-and-conquer* algorithms

Comparing Run Times

Run Times of Three Algorithms



Landau Notation (or Big-O Notation)

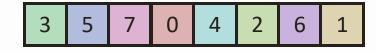
- Analysis Goals: Give a good estimate for the run time!
 - ► No bias on size of data
 - ► Independent of assumptions on data distribution
 - ► Stable with respect to implementation details
- Worst-Case Asymptotic Bounds expressed in Landau Notation
 - ▶ Let $f, g: \mathbb{N} \to \mathbb{R}_+$
 - ightharpoonup f is at most order of g or $f \in O(g)$ or f = O(g) iff $\exists c > 0 : \exists n_0 > 0 : \forall n \ge n_0 : f(n) \le c \cdot g(n)$ $\Leftrightarrow \limsup_{n \to \infty} \frac{f(n)}{g(n)} < \infty$
 - ightharpoonup We also say g is upper bound for f
- ▶ Does this kind of definition fulfill the goals?

Landau Notation (or Big-O Notation)

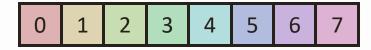
- ► More Useful Definitions:
 - ► f is at least order of g or $f \in \Omega(g)$ or $f = \Omega(g)$ iff $\exists c > 0 : \exists n_0 > 0 : \forall n \ge n_0 : f(n) \ge c \cdot g(n) \Leftrightarrow \liminf_{n \to \infty} \frac{f(n)}{g(n)} > 0$
 - ▶ f is order of g or $f \in \Theta(g)$ or $f = \Theta(g)$ iff $f \in O(g)$ and $f \in \Omega(g)$
 - $ightharpoonup \Omega$ is used to establish *lower bounds*, Θ for *tight bounds*
- ► Notation for imprecise bounds:
 - ► f is order strictly greater than g or $f \in \omega(g)$ or $f = \omega(g)$ iff $\forall c > 0$: $\exists n_0 > 0$: $\forall n \ge n_0$: $f(n) > c \cdot g(n) \Leftrightarrow \liminf_{n \to \infty} \frac{f(n)}{g(n)} = \infty$
 - ► f is order strictly lesser than g or $f \in o(g)$ or f = o(g) iff $\forall c > 0$: $\exists n_0 > 0$: $\forall n \ge n_0$: $f(n) < c \cdot g(n) \Leftrightarrow \limsup_{n \to \infty} \frac{f(n)}{g(n)} = 0$

Running Example for Today: SORTING

- ightharpoonup Input: An unsorted set A of n distinct comparable elements
 - ► Assumption: A is implemented an array
 - ightharpoonup Elements are taken from universe U with associated abstract total order \leq



- ► Output: A *n*-tuple $S = (s_1, s_2, ..., s_n)$ such that
 - ▶ *S* contains exactly the elements of *A*, i.e., $\bigcup_{i=1}^{n} s_i = A$
 - ▶ *S* is *sorted w.r.t. to* \leq , i.e., $\forall i \in \{1, ..., n-1\}$: $s_i \leq s_{i+1}$
 - ► Assumption: S is implemented as an array



▶ Which sorting algorithms do you know? What are their run times?

Recursive Bubble Sort

- One strategy for sorting:
 - 1. Find the *largest element* a^* of A and put it at position |A|
 - 2. Sort the elements of $A \setminus \{a^*\}$ recursively (i.e., the *first* |A| 1 *elements* of the array)
 - 3. Return A



- ► Paradigm: An algorithm that calls itself to solve related problems or subproblems is called recursive algorithm or recursion.
 - ► Base Case: Trivial solution without self-call

```
BubbleSort(A,n)
   if (n=1)
      return A;
   for i←1 to n-1
      if A[i] > A[i+1]
          swap(A[i],A[i+1]);
   BubbleSort(A,n-1);
   return A;
```

Recursive Bubble Sort – Run Time

► Run Time:

- ightharpoonup n comparisons
- ▶ potentially *n swaps*
- ightharpoonup Recursive call on $n \leftarrow n-1$
- Expression as recurrence: $T(n) \le T(n-1) + 2n$ $T(1) \le 2$
- ► Solving the recurrence:
 - ► *Insert recursive formulation* a few times:

$$T(n) = T(n-1) + 2n$$

$$= T(n-2) + 2(n-1+n)$$

$$= T(n-3) + 2(n-2+n-1+n) = \cdots$$

► Educatedly *guess closed form*:

$$T(n) = 2\sum_{i=1}^{n} i = n(n+1)$$

```
BubbleSort(A,n)
   if (n=1)
      return A;
   for i←1 to n-1
      if A[i] > A[i+1]
          swap(A[i],A[i+1]);
   BubbleSort(A,n-1);
   return A;
```

Recursive Bubble Sort – Run Time

► Status Quo:

- ► Given: T(n) = T(n-1) + 2n T(1) = 2
- Guess: T(n) = n(n+1)
- ► *Next Up:* Prove inductively!
 - ▶ Induction Base: T(1) = 1(1+1) = 2
 - ► Induction Hypothesis: For a fixed $n \ge 1$, T(n) = n(n + 1)
 - ► Inductive Step:

```
T(n+1) = T(n) + 2(n+1)
= {}^{IH} n(n+1) + 2(n+1)
= (n+2)(n+1)
```

```
BubbleSort(A,n)
   if (n=1)
      return A;
   for i←1 to n-1
      if A[i] > A[i+1]
          swap(A[i],A[i+1]);
   BubbleSort(A,n-1);
   return A;
```

Recursive Bubble Sort – Correctness

- Still to show: The algorithm is correct!
- ► Definition: A loop invariant is a property which remains true if it was true before entering the loop.
- Useful tool for proving correctness!
 - ► *Initialisation:* Establish that the invariant holds before the loop
 - ▶ Continuation: Show that the loop invariant remains true in the i-th iteration assuming that it holds after the (i-1)-th.
 - ► *Termination:* Use the loop invariant for the remainder of the proof.

```
BubbleSort(A,n)
   if (n=1)
      return A;
   for i←1 to n-1
      if A[i] > A[i+1]
          swap(A[i],A[i+1]);
   BubbleSort(A, n-1);
   return A;
```

Recursive Bubble Sort – Correctness

- For our for-loop: After i-th iteration, A[i+1] is the greatest of the first i+1 elements.
- ▶ *Initialisation:* Before the first iteration (i = 1), A[1] is greatest of the first 1 elements.
- **►** Continuation:
 - ► A[i] is greatest amongst the first i elements by assumption of the invariant before the iteration
 - ► Thus, either A[i] or A[i+1] is the greatest
 - ▶ We swap if necessary!
- ► Termination: A[n] is greatest element of A
 - ► Remainder of the proof: Induction

```
BubbleSort(A,n)
   if (n=1)
      return A;
   for i←1 to n-1
      if A[i] > A[i+1]
          swap(A[i],A[i+1]);
   BubbleSort(A, n-1);
   return A;
```

Recursive vs. Iterative Algorithms

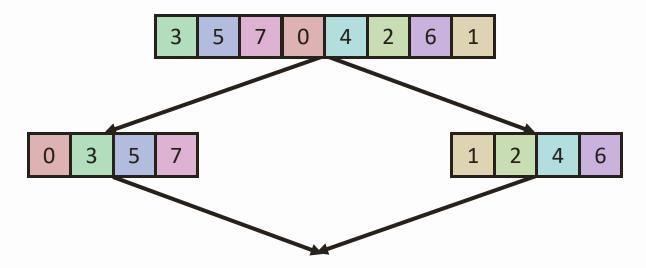
- Iterative version: Recursive calls are replaced by a loop!
- ► Works because the *signatures of calls are known in advance*
- ► *In general:* Recursion is *more powerful* than iteration!
 - ► *Recall:* Halting problem and Incomputability
 - But also more difficult to analyze, implement,

. . .

```
BubbleSort_rec(A,n)
   if (n=1)
      return A;
   for i←1 to n-1
          swap(A[i],A[i+1]);
   BubbleSort(A,n-1);
   return A;
```

Merge Sort

- ► Another strategy for sorting:
 - 1. Sort first half A_1 and second half A_2 independently yielding S_1 and S_2
 - 2. Compare first elements of S_1 and S_2 and move the lesser in S
 - 3. Repeat 2. until sorted



Merge Sort – Analysis

- ► Paradigm: A divide-and-conquer algorithm is a recursive algorithm that constructs a solution based on optimal recursively constructed subsolutions.
- Correctness (Sketch)
 - ► Prove correctness of *merge phase via* invariant
 - ightharpoonup Induction on n (assume power of 2)
- ► Run Time:

$$T(n) = 2T\left(\frac{n}{2}\right) + c \cdot n \qquad T(1) = 1$$

► Solvable with *Master Theorem*

```
MergeSort(A)
   if (n=1)
       return A;
   S1←MergeSort(A[1,...,n/2]);
   S2 \cdot Merge Sort(A[n/2+1,...,n]);
   i1, i2 \leftarrow 1;
   for (i←1 to n)
       if(S1[i1] < S2[i2])
           S[i] = S1[i1];
           i1 = i1+1;
       else
           S[i] = S2[i2];
           i2 = i2+1;
   return S;
```

Master Theorem

► Theorem: Let $a, b \ge 1$ constant and let $f(n), T(n) \ge 0$ such that $T(n) = a \cdot T\left(\frac{n}{b}\right) + f(n)$

- ▶ If $f(n) = O(n^{\log_b a \varepsilon})$ for $\varepsilon > 0$, then $T(n) = O(n^{\log_b a})$
- ▶ If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \cdot \log_b n)$
- ▶ If $f(n) = O\left(n^{\log_b a + \varepsilon}\right)$ for $\varepsilon > 0$ and $a \cdot f\left(\frac{n}{b}\right) \le c \cdot f(n)$ for c < 1 and n sufficiently large, then $T(n) = \Theta(f(n))$
- Merge Sort: $T(n) = 2T\left(\frac{n}{2}\right) + c \cdot n$
 - 1. Identify a, b and f(n): a = 2, b = 2, $f(n) = c \cdot n$
 - 2. Compute $\log_b a$: $\log_2 2 = 1$
 - 3. Compare f(n) and $n^{\log_b a}$: $f(n) = \Theta(n) = \Theta(n^{\log_b a})$
 - $\Rightarrow T(n) = \Theta(n \log n)$

Run Time Lower Bound for SORTING

► Theorem: Comparison-based sorting requires $\Omega(n \log n)$ time. $A[1] \leqslant A[2]$ ► *Proof:* Comparison-based sorting algorithms can $A[1] \leq A[3]$ $A[2] \leq A[3]$ Compare two elements Write an element to a different position ► Consider the *decision* (A[2],A[1],A[3])(A[1],A[2],A[3]) $A[1] \leq A[3]$ $A[2] \leq A[3]$ tree of any algorithm \triangleright n! leaves, one for each permutation ▶ thus, its *height is* $\Omega(\log(n!))$ ► At each inner node, there is one (A[1],A[3],A[2])(A[2],A[3],A[1])(A[3],A[2],A[1]) (A[3],A[1],A[2])comparison $\Rightarrow \Omega(n \log n)$ time

SORTING – Summary

- ▶ Theorem: SORTING can be solved in $\Theta(n \log n)$ time.
 - ▶ Bubble Sort takes $O(n^2)$ time.
- ► *Remark:* There are *non-comparison-based* sorting algorithms.
 - Examples: Radix Sort, Bucket Sort, Counting Sort
 - ▶ The *lower bound* can be extended to these!
 - ▶ But: More efficient if data contains mostly elements with short encoding or duplicates.