Week 3

Searching & Sorting

Thorben Klabunde www.th-kl.ch

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Agenda

- Mini-Quiz
- 2 Assignment
- Theory Recap
- 4 Additional Practice
- 6 Peer Grading

Mini-Quiz

Assignment

Feedback Assignment 2

- **Fibonacci Exercise**: Remember to cover all required cases in the B.C. and, consequently, also in the I.H. At the latest, when writing your proof and using the I.H. you should check that all the assumed cases are covered.
- **Limits**: Generally well done but please double check your calculations for arithmetic errors. When you apply a Theorem, get in the habit of briefly stating it.
- Bounds: Well done! General feedback on proof structure: Try to keep your proofs clearly structured (left to right, top to bottom) and guide the reader on how your statements are connected.

Ex. 3.1.b (3)

(3) Prove that for $n \geq 3$

$$\frac{n^{1/n} - 1}{n} = \Theta\left(\frac{\ln(n)}{n^2}\right).$$

You may use results from the earlier exercises.

You may use the following fact:

Let $f,g:\mathbb{R}^+\to\mathbb{R}$. Suppose that $\lim_{n\to\infty}g(n)=\infty$ and there exists some constant $C\in\mathbb{R}$ such that $\lim_{n\to\infty}f(n)=C$. Then $\lim_{n\to\infty}f(g(n))=C$.

Exercise 3.2 Substring counting.

Given a n-bit bitstring S[0..n-1] (i.e. $S[i] \in \{0,1\}$ for i=0,1,...,n-1), and an integer $k \geq 0$, we would like to count the number of nonempty contiguous substrings of S with exactly k ones. Assume $n \geq 2$.

For example, when S="0110" and k=2, there are 4 such substrings: "011", "11", "110", and "0110".

Ex. 3.2c)

(c) Consider an integer $m \in \{0, 1, \dots, n-2\}$. Using Prefixtable and Suffixtable, design an algorithm spanning (m, k, S) that returns the number of substrings S[i..j] of S that have exactly k ones and such that $i \leq m < j$.

For example, if S="0110", k=2, and m=0, there exist exactly two such strings: "011" and "0110". Hence, SPANNING(m,k,S)=2.

Describe the algorithm using pseudocode. Mention and justify the runtime of your algorithm (you don't need to provide a formal proof, but you should state your reasoning).

Hint: Each substring S[i..j] with $i \le m < j$ can be obtained by concatenating a string S[i..m] that is a suffix of S[0..m] and a string S[m+1..j] that is a prefix of S[m+1..n-1].

Ex. 3.2d)

(d)* Using spanning, design an algorithm with a runtime 1 of at most $O(n \log n)$ that counts the number of nonempty substrings of a n-bit bitstring S with exactly k ones. (You can assume that n is a power of two.)

Justify its runtime. You don't need to provide a formal proof, but you should state your reasoning.

Hint: Use the recursive idea from the lecture.

Ex. 3.4a)

(a) Design an O(n) algorithm that computes the nth Fibonacci number f_n for $n \in \mathbb{N}$. Describe the algorithm using pseudocode. Justify the runtime (you don't need to provide a formal proof, but you should state your reasoning).

Remark: As shown in part Exercise 2.2, f_n grows exponentially (e.g., at least as fast as $\Omega(1.5^n)$). For this exercise, you can assume that all addition operations can be performed in constant time.

Ex. 3.4b)

(b) Given an integer $k \geq 2$, design an algorithm that computes the largest Fibonacci number f_n such that $f_n \leq k$. The algorithm should have complexity $O(\log k)$. Describe the algorithm using pseudocode and formally prove its runtime is $O(\log k)$.

Hint: Use the bound proved in Exercise 2.2.

Theory Recap

Sorting Searching: The Big Picture

This week, the three key things you should learn are how to:

- Characterize the Search & Sort Problem Sorting data makes searching much faster $(O(n) \rightarrow O(\log n))$. Different sorting algorithms have fundamentally different runtimes $(O(n \log n) \text{ vs. } O(n^2))$.
- Proving Theoretical Limits to discuss Optimality
 Using decision trees, we proved a lower bound for an entire class of algorithms. For comparison-based search, this limit is $\Omega(\log n)$, making Binary Search optimal.
- Prove Correctness using Loop Invariants
 Loop invariants provide a formal way to prove iterative algorithms are correct using a simple inductive structure (base case, maintenance, termination).

Searching Algorithms at a Glance

The goal of a search algorithm is to find the index of an element b in an array A. The **efficiency** of the search **depends** heavily **on whether the array is sorted**.

Feature	Linear Search	Binary Search	
ldea	Iterate through the array from the beginning until the element is found.	Compare the target with the mid- dle element and discard half of the remaining array in each step.	
Requirement	None. The array can be unsorted.	The array must be sorted.	
Runtime	$\mathcal{O}(n)$ comparisons in the worst case.	$\mathcal{O}(\log n)$ comparisons.	
Optimality	Best possible for an unsorted array.	Asymptotically optimal for comparison-based search.	

Key Takeaway: Searching is significantly faster on sorted data.

Theoretical Lower Bound for Search: The Decision Tree Model

How can we prove that Binary Search, with its $\mathcal{O}(\log n)$ runtime, is the best we can do for searching in a sorted array?

Key Idea: The Decision Tree

Any comparison-based search algorithm can be modeled as a **decision tree**.

- Each **internal node** represents a comparison (e.g., b < A[i]?).
- Each leaf node represents a final outcome or result.

The worst-case number of comparisons for an algorithm is the height of its decision tree.

For an array A of size n, how many possible outcomes are there?

- The element *b* could be at any of the *n* indices.
- The element b might not be in the array at all.

This gives a total of n + 1 possible outcomes.

Deriving the Lower Bound

Theorem: Lower Bound for Comparison-Based Search

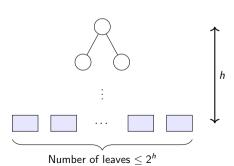
Any comparison-based search algorithm \mathcal{A} on a sorted array of size n requires at least $\Omega(\log n)$ comparisons in the worst case.

Proof Sketch:

- Every path from root to leaf is an execution path. The tree's height is the worst-case number of comparisons.
- **2** A must distinguish between n + 1 possible outcomes, each a unique leaf.
- **3** A binary tree of height h has at most 2^h leaves.
- Therefore, the number of leaves must be at least the number of outcomes:

$$n+1 \le 2^h \implies \log_2(n+1) \le h$$

n+1 possible outcomes must map to leaves



Bubble Sort

Idea: In each pass, let the largest unsorted element "bubble up" into its correct position.

Algorithm

Loop invariant (after *j* **passes):**

The largest j elements are in their final positions at the right end, i.e., the suffix A[n-j+1..n] is sorted in asc. order.

Comparisons:

$$\sum_{j=1}^{n-1}\sum_{i=1}^{n-1}1=\sum_{j=1}^{n-1}(n-1)=(n-1)^2=\Theta(n^2)$$

$$\sum_{j=1}^{n-1} \sum_{i=1}^{n-j} 1 = \sum_{j=1}^{n-1} (n-j) = \sum_{j=1}^{n-1} j \overset{\textit{Gauss}}{=} \Theta(n^2)$$

Swaps:

 $O(n^2)$ in the worst case.

Selection Sort

Idea: In each pass, find the largest unsorted element and swap it into its correct final position at the end of the array.

Algorithm

Invariant I(j)

After j iterations of the outer loop, the last j elements of the array contain the j largest values in sorted order.

$$A[1 \dots n-j] \mid A[n-j+1 \dots n]$$
 (unsorted) (sorted)

Comparisons:

$$\sum_{i=1}^{n-1} (n-1-j) = \frac{n(n-1)}{2} = \Theta(n^2)$$

Swaps: O(n) (only once per iteration)

Insertion Sort (with Binary Search for the Slot)

Idea: Build a sorted subarray at the beginning. Take the next element from the unsorted part and insert it into its correct place within the already sorted part.

Algorithm

```
insertionSort(A)
   int n = A.length;
   for j = 1 ... n-1:
   // bin. search for A[j+1] in A[1...j]
   // to find insertion idx as a by-product
      tmp = A[j+1]
      shift A[k...j] to A[k+1...j+1]
      A[k] = tmp
end for
```

Invariant I'(j)

After j iterations of the outer loop, the prefix of the array $A[1\ldots j]$ is sorted.

$$A[1 \dots j] \mid A[j+1 \dots n]$$
 (sorted)

Comparisons:

$$\sum_{j=1}^{n-1} \lceil \log_2 j \rceil \le O(n \log n)$$

Swaps: $\sum_{i=1}^{n-1} O(j) = \Theta(n^2)$ (shifts dominate).

Merge Sort

Idea: Divide and Conquer by splitting the array in half, sorting each half, and then merging the two sorted halves back together.

Algorithm

```
MergeSort(A, left, right)
  if left < right
    middle = (left + right) / 2
    MergeSort(A, left, middle)
    MergeSort(A, middle+1, right)
    Merge(A, left, middle, right)
  end if</pre>
```

Merge invariant:

At any time, B holds the sorted merge of the current prefixes of the two halves; after merging, A[L..R] is sorted.

Time (comparisons/moves):

$$T(n) = 2T(n/2) + cn \Rightarrow O(n \log n).$$
 (see last week's DnC Runtime Analysis)

Extra space: $\Theta(n)$ for the auxiliary array during merge (MS does *not* act in-place)

Proving Correctness with Invariants

The analysis of an algorithm always includes a proof of correctness and runtime considerations.

One technique for iterative algorithms is to use a **loop invariant**, which allows us to make statements about the state of some property with every iteration.

Proof by Induction with Loop Invariants

We must show three things about our invariant:

- Initialization (Base Case): The invariant is true before the first iteration of the loop.
- Maintenance (Inductive Step): If the invariant is true before an iteration (our I.H.), it remains true after the iteration.
- Termination: When the loop terminates, the invariant (combined with the loop's termination condition) guarantees the desired outcome.

Caution: The most critical part is choosing a precise and useful invariant!

Correctness of Bubble Sort

Let's prove the correctness of Bubble Sort using the invariant we discussed.

Bubble Sort Invariant

After j outer iterations, the suffix A[n-j..n-1] contains the j largest elements in increasing order.

Your Turn: Proving Bubble Sort Correct

Invariant I(j): After j outer iterations, the suffix A[n-j..n-1] contains the j largest elements in increasing order.

Base Case: Let j = 1. The invariant holds because ...

I.H.: Assume now that ...

I.S.: Then in the (j+1)-st pass of the outer loop ...

Termination: The loop finishes after j = n - 1 iterations. The whole array must be sorted since ...

Questions?

Additional Practice

Counting Loop Iterations - FS21

Algorithm 2

```
\begin{array}{l} \mathbf{for} \ j=1,\ldots,n \ \mathbf{do} \\ \mathbf{for} \ k=1,\ldots,n \ \mathbf{do} \\ \ell \leftarrow 1 \\ \mathbf{while} \ k+\ell \leq j \ \mathbf{do} \\ f() \\ \ell \leftarrow 2 \cdot \ell \end{array}
```

Theoretical Lower Bound for Sorting

Prove that the theoretical lower bound for any **comparison-based** sorting algorithm A is $\Omega(n \log n)$.

Proof.

Hint: Proceed analogously to the proof for search. Characterize the algorithm as a binary decision trees, count the leaves and conclude the result.

Quiz: Searching and Sorting

Claim	True	False
1. Bubble Sort algorithm always performs $\mathcal{O}(n^2)$ comparisons.		
2. The invariant for Selection Sort is that after j passes, the first j elements are sorted.		
3. In the worst case, Selection Sort performs $\mathcal{O}(n)$ swaps.		
4. Linear search can take $\mathcal{O}(\log n)$ operations on some arrays.		

Frame Title

- ii) Assume $n \ge 10$. For which of the following arrays does InsertionSort take time $\Omega(n^2)$ (to sort the array in ascending order)?
 - (a) $[2,1,3,4,5,\ldots,n-2,n-1,n]$
 - (b) $[n, n-1, n-2, \ldots, 3, 2, 1]$
 - (c) For both
 - (d) For neither

Peer Grading

Peer-Grading Exercise

This week's peer-grading exercise is **Exercise 3.1**

Each group grades the group below in the table I sent you (resp. the last one grades the first one). Please send the other group your solution. If you don't get their solution, please contact me so I can send it to you.