CS660: Algorithms - Lecture 1

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- Textbook: Algorithms by Jeff Erickson. The book is available online for free at http://jeffe.cs.illinois.edu/teaching/algorithms/.
- You can also buy a hard copy on Amazon for about \$30 (some additional chapters are not in the hard copy).
- Please do the reading assignment before class. The reading assignment for each lecture can be found on the course's blackboard.

Topics:

- Divide and conquer
- Dynamic programming
- Greedy algorithms
- Graph algorithms (shortest paths, bipartite matching, maximum flow, etc.)
- NP-Completeness and approximation algorithms
- Randomized algorithms.

- An introduction course in data structures and algorithms is strongly advised. You should be familiar with basic data structures and algorithms.
- A good background in algebra and discrete math is also necessary. In particular, you should be familiar with how to write and read proofs.

- There will be 2 midterm exams, 1 final exam, and homework. The course grade is broken down as follows:
 - Midterms (25 % each)
 - Final exam (35 %)
 - Homework (40%)
- See the university's website for the final exam schedule.

- The lowest score homework will be dropped. No late homework will be accepted. If you need to do make-up exams, please notify me in advance.
- Do not search for solutions to the homework online.
- You may discuss the homework with other students. However, you must write up the solutions on your own.
- Cheating and plagiarism will be dealt with according to the university policy.

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What is this course about?

- Design efficient algorithms
 - Efficiency is measured by running time, memory use, communication & number of rounds (in distributed algorithms).
 - In this course, we will mostly focus on the running time.
- What are the steps to design an algorithm?
 - Specify the algorithm (the best way to describe an algorithm is via structured English & pseudo-code). Read page 13 in the textbook and the examples we cover in class. You should describe algorithms in this fashion.
 - 2 Show that the algorithm is correct.
 - Analyze the running time.

A bit about myself

- I do research in theoretical computer science, mostly algorithms.
- If you are interested in doing research or an MS project with me, please talk to me (I do have funding).
- Some topics: graph algorithms, data science algorithms, or applying algorithms to suitable applications.

Abstraction: Makes it easier to study efficiency

- Making an algorithm fast involves many considerations that are architecture specific.
- For the sake of simplicity and generality, we will assume that: any location in memory can be accessed a unit cost and measure running time in basic steps such as pairwise arithmetic operation and memory accesses.

Example: Selection sort

- Suppose we want to sort a list of integers $A[1], A[2], \ldots, A[n]$.
- Selection sort algorithms:
 - **1** For i = 1, ..., n:
 - **1** Scan through $A[i], A[i+1], \ldots, A[n]$ to find the smallest number.
 - ② Let A[j] be the smallest among $A[i], A[i+1], \ldots, A[n]$.
 - 3 Swap A[j] and A[i].
- Proof of correctness: After the t-th iteration, A[i] is the i-th smallest number for $1 \le i \le t$. Therefore, after the n-th iteration, the list is sorted.
- Running time: in the i-th iteration, we need to access $A[i], A[i+1], \ldots, A[n]$ to find the next smallest element. This takes n-i+1 unit time cost. The swapping step can also be done in 3 unit time cost. Therefore, the running time is

$$\sum_{i=1}^{n}((n-i+1)+3)=3n+\sum_{i=1}^{n}i=3n+\frac{n(n+1)}{2}=O(n^{2}).$$

Big-Oh

- To simplify the running time analysis, we do not count the steps exactly.
- The running time is measured "asymptotically".
- Definition: f(n) = O(g(n)) if there exists constants c and n_0 such that

$$f(n) \leq c \cdot g(n)$$
 for all $n \geq n_0$.

• Selection sort running time:

$$3 + \frac{n(n+1)}{2} = O(n^2) .$$

• Exercise: show the above by definition.

Big-Oh

• Definition: f(n) = o(g(n)) if there exists constants c and n_0 such that for all k > 0

$$f(n) \le ck \cdot g(n)$$
 for all $n \ge n_0$.

• Exercise: Show that $n^2 = o(n^{2.5})$ by definition.

Slightly easier characterization

•
$$f(n) = O(g(n))$$
 if

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}<\infty.$$

•
$$f(n) = o(g(n))$$
 if

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}=0.$$

Big-Oh exercises

- Show that 100 = O(1).
- Show that $3n^3 + n = O(n^3)$.
- Show that $n \log_2 n = O(n^{0.01})$.
- Show that $\log_b n = O(\log_a n)$ for constants a, b.
- Show that f(n) = O(g(n)) and $\ell(n) = O(k(n))$ then $f(n) + \ell(n) = O(g(n) + k(n))$.

Slightly easier characterization

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L'Hospital's rule

• If f' and g' exist, then

$$\lim_{n\to\infty}\frac{f(n)}{g(n)}=\lim_{n\to\infty}\frac{f'(n)}{g'(n)}.$$

- Use L'Hospital rule to show that $10n^{10} 1000n^9 = O(n^{10})$.
- Exercise (will be in homework 1) Show that $n^3 = O(e^{\sqrt{n}})$.

Other notations

- $f(n) = \Theta(g(n))$ if f(n) = O(g(n)) and g(n) = O(f(n)).
- $f(n) = \Omega(g(n))$ if g(n) = O(f(n)).
- $f(n) = \omega(g(n))$ if g(n) = o(f(n)).

Divide and Conquer (Recursion) - Exponential

- Assuming *n* is a power of 2. We want to compute $A(n) = 3^n$.
- Algorithm 1: Set x to 1. For i = 1 to n: $x \leftarrow 3x$.
- Algorithm 2:
 - ① If n = 1, return 3.
 - 2 Else x = A(n/2). Return $x \cdot x$.
- Assuming multiplication takes O(1) time, the second algorithm runs in $O(\log n)$ time. Why?

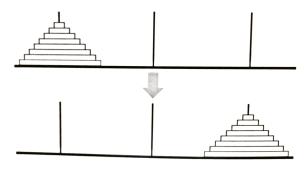


Figure 1.1. The (8-disk) Tower of Hanoi puzzle

- Move the discs from the original needle to the target needle.
- During the process, no bigger disc can be on top of a smaller disc.
- Design an algorithm that output a solution to the problem.

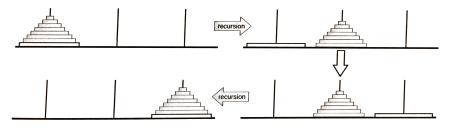


Figure 1.2. The Tower of Hanoi algorithm; ignore everything but the bottom disk.

- First, move the n-1 smallest discs to the non-target needle.
- Move the largest disc to the target needle.
- Move the n-1 smallest discs to the target needle.

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\frac{\text{Hanoi}(n, src, dst, tmp):}{\text{if } n > 0}
\text{Hanoi}(n-1, src, tmp, dst) \quad \langle \langle \text{Recurse!} \rangle \rangle
\text{move disk } n \text{ from } src \text{ to } dst
\text{Hanoi}(n-1, tmp, dst, src) \quad \langle \langle \text{Recurse!} \rangle \rangle
```

Figure 1.4. A recursive algorithm to solve the Tower of Hanoi

The running time is given by the recurrence

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

- Claim: $T(n) = 2^n 1$. Prove by induction.
- $T(1) = 2^1 1 = 1$ which is true since we only need to move one disc.
- Suppose $T(i) = 2^i 1$. Then $T(i+1) = 2T(i-1) + 1 = 2(2^i 1) + 1 = 2^{i+1} 1$.
- Therefore, the running time is $O(2^n + 1) = O(2^n)$.