CS 344: Design & Analysis of Algorithms

Lecture 1

Sep 3, 2019

Staff

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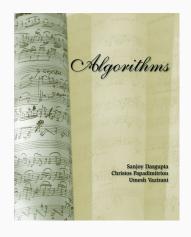
Recitations

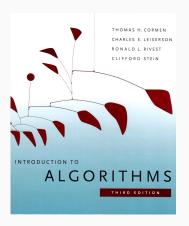
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Tue 8:25 PM - 9:20 PM, SEC-205 (Yuwei)
Thu 8:55 AM - 9:50 AM, BE-250 (Zelong)
Thu 8:25 PM - 9:20 PM, SEC-209 (Jianchao)
Tue 8:25 PM - 9:20 PM, SEC-207 (Chengguizi)
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Homeworks

- 5-6 written homeworks
- 1–2 weeks for each
- ullet small groups are allowed (\leq 3 people)

Textbooks





DPV CLRS

Exams

- Midterm 1: Oct. 17
- Midterm 2: Nov. 14
- Final: Dec. 17 (8 pm 11 pm)

Grading (tentative)

20%
25%
25%
30%

Topics (tentative)

- Big-O notation, asymptotics
- Divide and conquer
- Dynamic programming
- Greedy algorithms
- Sorting
- Graph algorithms

- Linear programming
- Number theoretic algorithms
- Computational geometry
- String matching
- NP-completeness
- Approximation algorithms

Why algorithms?

Fibonacci numbers: $\{0, 1, 1, 2, 3, 5, 8, 13, 21, ...\}$

$$F_n = F_{n-1} + F_{n-2}$$

This grows quickly:

$$F_n \approx 2^{0.649n}$$

```
def fib(n):
    if n == 0:
        return 0
    elif n == 1:
        return 1
    else:
        return fib(n-1) + fib(n-2)
```

Three questions:

- Is it correct?
- How much time does it take?
- Can we do better?

For $n \leq 1$, 1 operation.

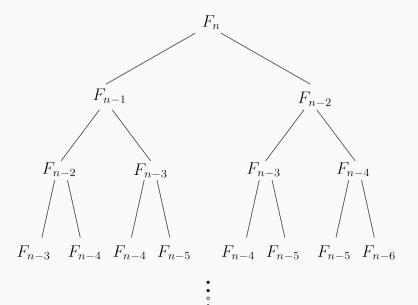
For n > 1,

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

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

$$F_n = F_{n-1} + F_{n-2}$$

$$F_n \approx 2^{0.649n}$$



```
def fib2(n):
    fibs = [0, 1]
    for i in range(2, n+1):
        fibs.append(fibs[i-1] + fibs[i-2])
    return fibs[n]
```

Sum numbers 1 through *n*:

```
sum = 0
for i in range(1, n+1):
    sum += i
```

- Is it correct?
- How long does it take?
- Can we do better?

Sum numbers 1 through n:

```
sum = n * (n + 1) / 2
```

- Is it correct?
- How long does it take?
- Can we do better?

Base case (n = 0):

$$\sum_{i=1}^{0} i = 0 = \frac{0(0+1)}{2}$$

Inductive step:

Assume that

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

and prove that

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

$$\sum_{i=1}^{n} i = \sum_{i=1}^{n-1} i + n$$

$$= \frac{(n-1)n}{2} + n$$

$$= \frac{(n-1)n + 2n}{2}$$

$$= \frac{n^2 - n + 2n}{2}$$

$$= \frac{n^2 + n}{2}$$

$$= \frac{n(n+1)}{2}$$

How long does it take?

- One addition
- One multiplication
- One division

We generally want to abstract out certain details:

- Per-operation cost
- Specialized CPU instructions
- Compiler optimizations
- Memory layout
- Cache effects

This goes from exponential to polynomial, but is it that important? We have Moore's law...

- How many new Fibonacci numbers could we compute?
- Exponential algorithm ⇒ polynomial benefit
- Polynomial algorithm ⇒ exponential benefit

We care mainly about the growth of the running time, in terms of the input size, n.

•
$$42n^2 + 3n + 4$$

•
$$7n + 2$$

•
$$2^{0.649n} + n^4 + 2n^2$$

And we care mainly about the dominant factor in the running time.

•
$$42n^2 + 3n + 4 \rightarrow 42n^2$$

•
$$7n+2 \rightarrow 7n$$

•
$$2^{0.649n} + n^4 + 2n^2 \rightarrow 2^{0.649n}$$

Finally, we want to eliminate constant coefficients.

•
$$42n^2 + 3n + 4 \rightarrow n^2$$

•
$$7n + 2 \rightarrow n$$

•
$$2^{0.649n} + n^4 + 2n^2 \rightarrow 2^n$$

Big-O notation:

•
$$42n^2 + 3n + 4 = O(n^2)$$

•
$$7n + 2 = O(n)$$

•
$$2^{0.649n} + n^4 + 2n^2 = O(2^n)$$

Aside: "equality" in big-O notation

- $O(n) = O(n^2)$
- $O(n^2) \neq O(n)$

Aside: "equality" in big-O notation

- $O(n) \in O(n^2)$
- $O(n^2) \notin O(n)$

What does f(n) = O(g(n)) mean?

- g(n) is a kind of upper bound on f(n)
- for sufficiently large *n*

What does f(n) = O(g(n)) mean?

Attempt 1:

For all n, $f(n) \leq g(n)$.

What does f(n) = O(g(n)) mean?

Attempt 2:

There exists some constant c, such that for all n, $f(n) \le c \cdot g(n)$.

What does
$$f(n) = O(g(n))$$
 mean?

There exists some constants c and N such that, for all n > N, $f(n) \le c \cdot g(n)$.

$$\exists c, N \ \forall n > N, f(n) \leq c \cdot g(n)$$

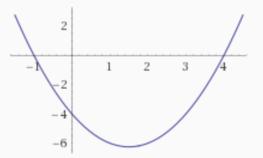
$$42n^2 + 3n + 4 = O(n^2)$$

means $42n^2 + 3n + 4 \le c \cdot n^2$ for n > N.

$$42n^2 + 3n + 4 \le c \cdot n^2$$
 for $n > N$.
Let $c = 43$. When is $42n^2 + 3n + 4 \le 43n^2$?

$$42n^{2} + 3n + 4 \le 43n^{2}$$
$$3n + 4 \le n^{2}$$
$$0 \le n^{2} - 3n - 4$$





$$42n^2 + 3n + 4 = O(n^2)$$

since for all n > 4,

$$42n^2 + 3n + 4 \le 43n^2$$

Rules of thumb:

- Constants can be omitted: $7n \rightarrow n$
- Higher exponents dominate: $n^3 + n^2 \rightarrow n^3$
- Exponentials dominate polynomials: $2^n + n^2 \rightarrow 2^n$
- Polynomials dominate logarithms: $n + \log n \rightarrow n$

We also need to consider different inputs, since they may cause the algorithm to perform more or less work.

- Worst case
- Best case
- Average case

Insertion sort: work from left to right, sorting as we go, and "insert" a value where it belongs in the already-sorted region.

- Say we're looking at element i
- Copy arr[i] to a temporary variable
- Find where that value belongs and insert it

Insertion sort:

for each position i:

- walk backwards from i, shifting values as you go, until you find a value less than a[i]
- put a[i] into this new position





sorted region





Insertion sort:

- Worst case: $O(n^2)$
- Best case: O(n)

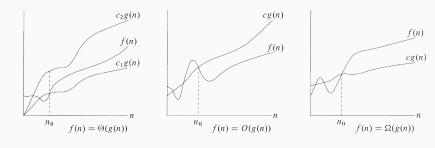
Big-O notation is similar to a "less-than" relation.

What about "greater-than" or "equal"?

- $O(\cdot)$: "less-than"
- $\Omega(\cdot)$: "greater-than"
- $\Theta(\cdot)$: "equal"

$$f(n) = \Omega(g(n))$$
 means $g(n) = O(f(n))$

$$f(n) = \Theta(g(n))$$
 means $f(n) = O(g(n))$ and $g(n) = O(f(n))$



- $7n^2 + 3 = O(n^2)$
- $7n^2 + 3 = O(n^3)$
- $7n^2 + 3 = \Omega(n)$
- $7n^2 + 3 = \Theta(n^2)$