Recurrences and Recursion

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- Recursion is one of the most versatile techniques in programming as well as algorithm design
- Closely related to induction:
 - Consists of a base case and recursive case, similar to base and inductive steps.
 - · Correctness of recursive algorithms is proved by induction
- Example: Computing Fibonacci numbers:

Base case(s): F(0) = 0, F(1) = 1Recursive case: F(n) = F(n-1) + F(n-2)

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Recursive data types: e.g., lists, trees,...

Base Case: empty list (for lists), leaf (for trees),...

Recursive Case: pair of element and rest of list, pair of trees, etc.

Recurrences: Tower of Hanoi Problem



Goal: Move all disks from one post to another. Rules:

- Only the top-most disk can be moved.
- No disk can be placed on a smaller disk.

Questions:

- How do you solve the puzzle?
- How many moves will be needed?









A Recursive Algorithm for Tower of Hanoi Problem





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- *MoveStack*(*n* − 1, 1, 2)
- *MoveDisk*(*n*, 1, 3)
- *MoveStack*(*n* − 1, 2, 3)



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Base Case: MoveStack(1, x, y)• MoveDisk(1, x, y)

A Recurrence for the Runtime of Towers of Hanoi Algorithm



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Solving Recurrences: Plug and Chug

- Expand the recurrence out for a few steps
- Identify the pattern
- Guess a solution based on the pattern
- Check the solution for a few small values of *n*
- Verify using induction

Plug and Chug for Tower of Hanoi Recurrence

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

Examples of Recurrences for Runtimes: Exponentiation

$$exp(x, n) = \begin{cases} 1, & \text{if } n = 0 \\ x * exp(x, n - 1), & \text{otherwise} \end{cases}$$

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Examples of Recurrences for Runtimes: Fast Exponentiation

$$fexp(x, n) = \begin{cases} 1, & \text{if } n = 0\\ x, & \text{if } n = 1\\ fexp(x * x, n/2), & \text{if } n \text{ is even}\\ fexp(x * x, n/2) * x, & \text{if } n \text{ is odd} \end{cases}$$

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Solving Linear Recurrences

• Homogeneous linear recurrences are of the form

$$f(n) = \sum_{i=1}^{d} a_i f(n-i)$$

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$$\sum_{i=0}^{d} a_{i} x^{d-i} = 0 \quad \text{(Rearrange terms to arrive at a polynomial, with } a_{0} = 1\text{)}$$

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- Note: if the polynomial has fewer than *d* roots, the general form of the solution gets more complicated we will ignore this case here.

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1. Substitute $f(n) = x^n$ in this equation, simplify to get *characteristic equation* $x^2 = x + 1$

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- 4. Plug in f(0) = 0 and f(1) = 1 to obtain the following equations:

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$$k_1 p^0 + k_2 q^0 = f(0) = 0$$

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- 5. Thus, the solution is

$$f(n) = \frac{1}{\sqrt{5}} \left(\frac{1+\sqrt{5}}{2}\right)^n - \frac{1}{\sqrt{5}} \left(\frac{1-\sqrt{5}}{2}\right)^n$$

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- Note that $|q| = |\frac{1-\sqrt{5}}{2}| = 0.6180 < 1$ so q^n rapidly approaches zero. For instance, $q^{20} \approx 0.00006$, and the error in f(n) due to ignoring q is less than one in 10^{-8} .

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- So, for larger *n*, f(n) is determined almost entirely by the first term $\frac{1}{\sqrt{5}} \left(\frac{1+\sqrt{5}}{2}\right)^n$ • $p^n/\sqrt{5}$ is very close to an integer value, although *p* is irrational!

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- So, for larger n, f(n) is determined almost entirely by the first term ¹/_{√5} (^{1+√5}/₂)ⁿ
 pⁿ/√5 is very close to an integer value, although p is irrational!
- The ratio between successive Fibonacci numbers converges to *p* = 1.618, which is called the *golden ratio*

Asymptotic Complexity

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 - Each such operation may in fact take a different amount of time
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- Expressing complexity in terms of "number of steps" is a simplification
 - Each such operation may in fact take a different amount of time
 - But it is too complex to worry about the details, esp. because they differ across programming languages, processor types, etc.
- Why not simplify further?
 - Capture just the growth rate of T(n) as a function of n
 - Ignore constant factors
 - No need to count operations in a loop (their number should be bounded by a constant)
 - Ignore exceptions from the formula for small values of *n*

Asymptotic Complexity: Big-O notation

Definition

Given functions $f, g : \mathbb{R} \longrightarrow \mathbb{R}$, we say f = O(g), i.e., "f grows no faster than g," iff $\lim_{x \to \infty} f(x)/g(x) < c$ for some constant c



Big-O notation: Examples

- 10n = O(n)
- $0.0001n^3 + n = O(n^3)$
- $2^n + 10^n + n^2 + 2 = O(10^n)$
- $0.0001n \log n + 10000n = O(n \log n)$

Solving Divide-and-Conquer Recurrences: Master Theorem

If $T(n) = aT\left(\frac{n}{b}\right) + O(n^d)$ for constants a > 0, b > 1, and $d \ge 0$, then

$$T(n) = egin{cases} O(n^d), & ext{if } d > \log_b a \ O(n^d \log n) & ext{if } d = \log_b a \ O(n^{\log_b a}) & ext{if } d < \log_b a \end{cases}$$

Solving Recurrences: Examples Using Master Theorem

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$$T(n)=4T(n/2)+n^3$$

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