

General Notation

\neg	→ Not
\exists	→ There exists
\vdots	→ Such that
\forall	→ For all
\equiv / \Leftrightarrow	→ Equivalent
\implies	→ Implies
\in	→ Is in/Belongs to
\wedge	→ And
\vee	→ Or
\nearrow	→ Strictly increasing
\searrow	→ Strictly decreasing
\nearrow	→ Eventually increasing
\searrow	→ Eventually decreasing

Sets of Numbers

\mathbb{Z}	→ Integers
\mathbb{R}	→ Real numbers
\mathbb{Q}	→ Rational numbers
\mathbb{N}	→ Natural numbers
\mathbb{P}	→ Prime numbers
\mathcal{U}	→ Universal set (everything)
$\emptyset, \{\}$	→ Empty set (nothing)

Common Operations

Floor	$\lfloor n \rfloor \rightarrow$ Truncate n
Ceiling	$\lceil n \rceil \rightarrow$ Round up n
Factorial	$n! \rightarrow 1 \times 2 \times 3 \times 4 \times \dots \times n$
Summation	$\sum_{k=1}^n f(k) \rightarrow \sum_{k=1}^5 2k = 2 + 4 + 6 + 8 + 10$
Binomial	$\binom{n}{k} \rightarrow \frac{n!}{k!(n-k)!}$
Divides	$n \mid k \rightarrow$ iff $\exists c \in \mathbb{Z} : k = nc$

Rules of Logarithms

$\log_b x = y$	$\iff b^y = x$
$b^{\log_b x}$	$\iff x$
$\log_b(b^x)$	$\iff x$
$\log_b(xy)$	$\iff \log_b x + \log_b y$
$\log_b(x/y)$	$\iff \log_b x - \log_b y$
$x^{\log_b n}$	$= n^{\log_b x}$
$\log_b(x^y)$	$\iff y \times \log_b x$
$x^{\log_b n}$	$= n^{\log_b x}$

Big-O, o, Ω and Θ ($\forall n > n_0$)

$f(n) \in O(g(n))$	$\rightarrow \exists C : 0 \leq f(n) \leq Cg(n)$
$f(n) \in o(g(n))$	$\rightarrow \forall C > 0 : 0 \leq f(n) < Cg(n)$
$f(n) \in \Omega(g(n))$	$\rightarrow \exists C : 0 \leq Cg(n) \leq f(n)$
$f(n) \in \Theta(g(n))$	$\rightarrow \exists A, B : Ag(n) \leq f(n) \leq Bg(n)$

Common Time-Complexities

Constant	$\rightarrow O(1)$
Logarithmic	$\rightarrow O(\log n)$
Linear	$\rightarrow O(n)$
Quasi-Linear	$\rightarrow O(n \log n)$
Quadratic	$\rightarrow O(n^2)$
Cubic	$\rightarrow O(n^3)$
Exponential	$\rightarrow O(2^n)$
Factorial	$\rightarrow O(n!)$

Proof Types (For $P \implies Q$)

Direct	\rightarrow Assume P , then use rules of logic to prove Q
Cases/Exhaustion	$\rightarrow P_1 \vee P_2 \vee \dots \vee P_n \implies Q$
Contradiction	\rightarrow Assume P and derive $P \not\Rightarrow Q$
Inductive	$\rightarrow P(1) \wedge P(k+1) \implies P(k)$

Common Series

$\sum_{k=1}^n k$	$\rightarrow 1 + 2 + 3 + \dots + n = \frac{n(n+1)}{2}$
$\sum_{k=1}^n k^2$	$\rightarrow 1 + 2^2 + 3^2 + \dots + n^2 = \frac{n(n+1)(2n+1)}{6}$
$\sum_{k=1}^n k^3$	$\rightarrow 1 + 2^3 + 3^3 + \dots + n^3 = \left[\frac{n(n+1)}{2} \right]^2$
$\sum_{k=1}^n ar^{k-1}$	$\rightarrow ar^0 + ar^1 + ar^2 + \dots + ar^{n-1} = a \left(\frac{1-r^n}{1-r} \right)$
$\sum_{k=0}^n \binom{n}{k}$	$\rightarrow \binom{n}{0} + \binom{n}{1} + \binom{n}{2} + \dots + \binom{n}{n} = 2^n$

Solving Recurrence Relations

The general steps for solving a recurrence relation are

- Find an explicit solution using the Characteristic Equation
- Verify solution using an inductive proof

First use Algebra to take a simple Recurrence Relation and turn it into a LHRCC

$$T_n = T_{n-1} + T_{n-2}$$

LHRCC is a Linear Homogeneous Recurrence Relation with Constant Coefficients.

$$T_n - T_{n-1} - T_{n-2} = 0$$

Use the Characteristic Equation, where C is the coefficient from the LHRCC:

$$C_0X^k + C_1X^{k-1} + C_2X^{k-2} + \dots + C_k = 0$$

For this example it would be

$$x^2 - x - 1 = 0 \implies x = \frac{1 \pm \sqrt{5}}{2}$$

Then solve for x and substitute into the General Solution

$$f_n = Ax_1^n + Bx_2^n$$

Next, use the initial conditions to solve for the explicit solution

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

Finally, verify the explicit solution that was found using an inductive proof.

Programs

BS Computer Science

MG Data Science

COS 485

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Divide and Conquer

If $f(n)$ is the number of operations required to solve an initial problem, then a Divide and Conquer Recurrence Relation will look like

$$f(n) = \underbrace{a}_{\text{subproblems}} f \underbrace{\left(\frac{n}{b} \right)}_{\text{branching factor}} + \underbrace{g(n)}_{\text{work per level}}$$

where $n = b^k$, $k \geq 1$

Theorem: Suppose $f \nearrow$ (is strictly increasing), $b \mid n$, $a \geq 1$, $b \geq 1$, $c \in \mathbb{R}^+$: $f(n) = af\left(\frac{n}{b}\right) + O(1)$, then

$$f(n) \in \begin{cases} O(n^{\log_b a}) & \text{if } a > 1 \\ O(\log_b n) & \text{if } a = 1 \end{cases}$$

Master Theorem

Suppose a complexity function $T(n)$ is eventually non-decreasing and satisfies

$$T(n) = aT\left(\frac{n}{b}\right) + O(n^d) \text{ for } n > 1, n \text{ a power of } b$$

$$T(1) = c$$

Where $b \geq 2$ and $d \geq 0$ are constant *integers*, and a and c are constant such that $a > 0$ and $c > 0$.

$$T(n) \in \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \lg n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Cheat Sheet

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Design/Analysis of Comp Algorithms

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Greedy Algorithms

Greedy algorithms are a class of algorithms that make **locally optimal** choices at each step with the hope of finding a **global optimum** solution. The idea is to select the best local choice at each step, leading to a solution that may or may not be the most optimal but is often good enough. Problems that are solved using Greedy Algorithms include

- Knapsack Problem
- Shortest Path (Graph)
- Minimal Spanning Tree

Dynamic Programming (DP)

DP solves problems by breaking them down into simpler **subproblems**. By solving each subproblem only once and storing the results, it avoids repeating computations and is more efficient. Dynamic Programming algorithms can use either a 'Top-Down' approach or a 'Bottom-Up' approach which use **memoization** or **tabulation** to store subproblems, respectively. Some examples of problems that use Dynamic Programming are:

- Fibonacci Sequence
- Shortest Path (Graph)
- Chain Matrix Multiplication

Contact

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