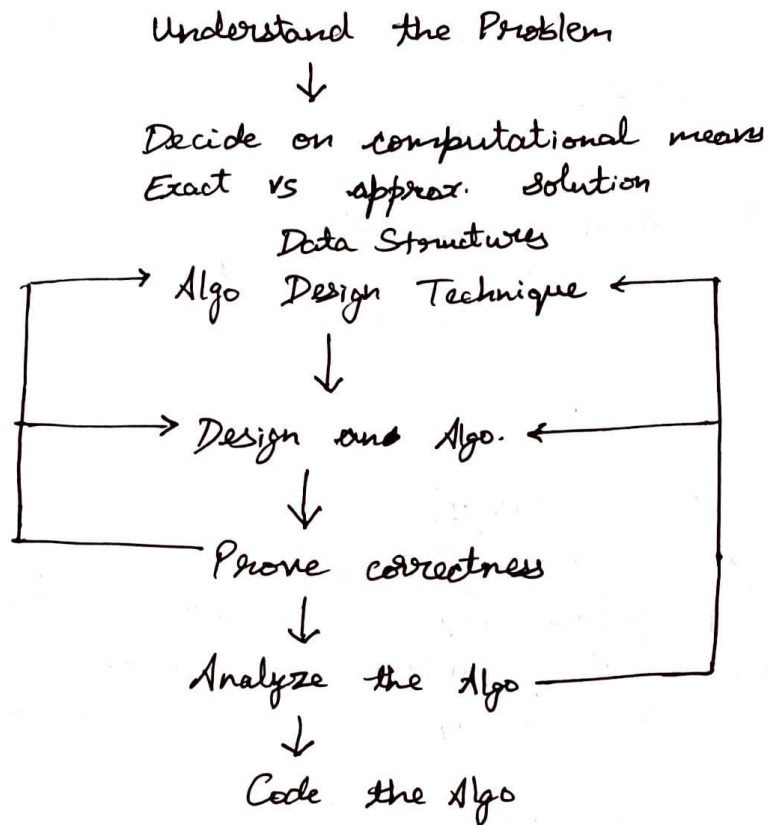


Algo. Design and Analysis Problem



→ Specifying an algo

- Natural Language
- Pseudo code
- Flowchart

→ Analyzing an algo

- Efficiency
 - Time
 - Space
- Simplicity
- Generality

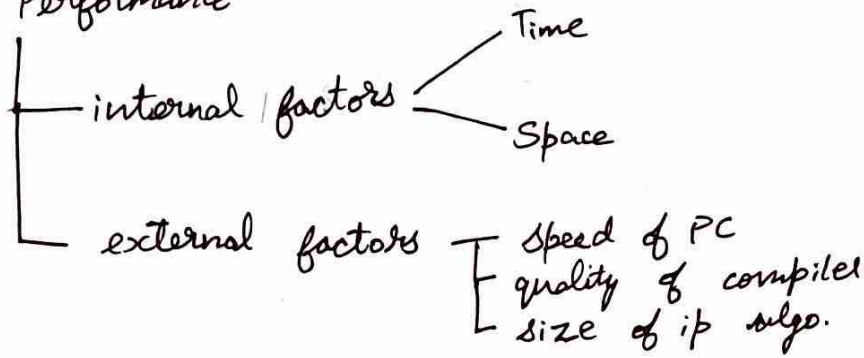
→ Coding an algo

- Efficient implementation
- Correctness of program
- Code optimization

Analysis Framework

→ Complexity of an algo.

Algo's Performance



→ Space Complexity

$$S(P) = C + SP(I)$$

Fixed space requirements (C)
independent of the characteristics
of ips and ops

Variable Space Requirements
(SP(I))
dependent on instance
characteristic I

→ Time Complexity

Experimental Study

Theoretical Analysis

$$T(P) = C + T_P(I)$$

Compile Time (C)
independent of instance
characteristics

run (execution) time T_P

→ Theoretical Analysis

1. Order of magnitude / Asymptotic categorization
2. Estimation of running Time.

1. Operation counts
2. Step counts

Basic operation count

$$T(n) \approx C_{op} C(n)$$

Running Time execution time for basic operation no. of times basic operation is executed

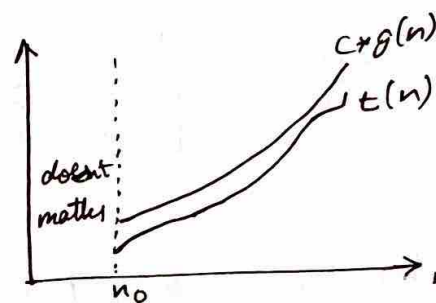
Asymptotic Notations

- $O(g(n))$: class of functions $f(n)$ that grow no faster than $g(n)$ upper bound
 $\Omega(g(n))$: at least as fast as $g(n)$ - lower bound
 $\Theta(g(n))$: at same rate as $g(n)$ - average bound
 $o(g(n))$: slower rate than $g(n)$
 $\omega(g(n))$: faster rate than $g(n)$

class: $1 < \log n < \sqrt{n} < \textcircled{n} < n \log n < n^2 < n^3 < \dots < 2^n < 3^n < \dots < n^n$
Lower bound avg bound upper bound

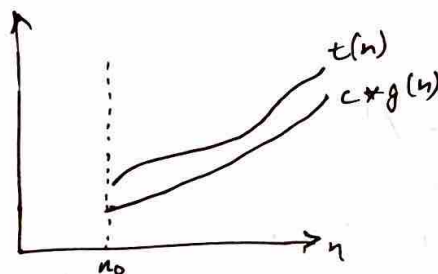
O -notation

$$t(n) \leq c * g(n) \text{ for all } n \geq n_0$$



Ω -notation

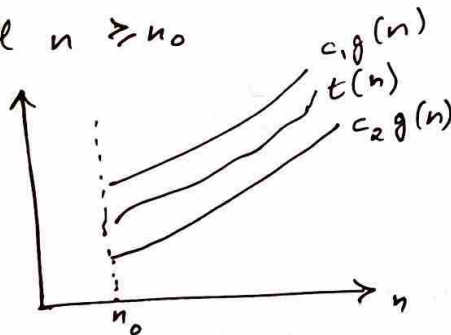
$$t(n) \geq c * g(n) \text{ for all } n \geq n_0$$



$$t(n) \in O(g(n))$$

Θ -notation

$$c_2 g(n) \leq t(n) \leq c_1 g(n) \text{ for all } n \geq n_0$$



Little-o notation

$$0 \leq t(n) < c * g(n) \text{ for all } n \geq n_0$$

Little Omega notation

$$t(n) > c * g(n) \geq 0 \text{ for all } n \geq n_0$$

Theorems

1. $t_1(n) \in O(g_1(n))$, $t_2(n) \in O(g_2(n))$
then $t_1(n) + t_2(n) \in O(\max\{g_1(n), g_2(n)\})$
2. $t_1(n) \in \Theta(g_1(n))$, $t_2(n) \in \Theta(g_2(n))$
then $t_1(n) + t_2(n) \in \Theta(\max\{g_1(n), g_2(n)\})$
3. $t_1(n) \in \Omega(g_1(n))$, $t_2(n) \in \Omega(g_2(n))$
then $t_1(n) + t_2(n) \in \Omega(\max\{g_1(n), g_2(n)\})$

Using limits to compare order of growth

$$\lim_{n \rightarrow \infty} \frac{t(n)}{g(n)} = \begin{cases} 0 & \Rightarrow t(n) \text{ has a smaller order of growth than } g(n) \\ c & \Rightarrow t(n) \text{ same } g(n) \\ \infty & \Rightarrow t(n) \text{ greater } g(n) \end{cases}$$

Time Efficiency of Non-recursive Algos.

8th February, 2024

eg: Max element

```

maxvalue ← A[0]
for i ← 1 to n-1 do
    if A[i] > maxvalue
        maxvalue ← A[i]
return maxvalue
    
```

$$C(n) = \sum_{i=1}^{n-1} 1 = (n-1 + 1) = n-1 \in O(n)$$

eg2: Unique Elements

```

for i ← 0 to n-2 do
    for j ← i+1 to n-1 do
        if A[i] == A[j] return false
    
```

$$\begin{aligned}
 C_{worst}(n) &= \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 = \sum_{i=0}^{n-2} ((n-1) - (i+1) + 1) \\
 &= \sum_{i=0}^{n-2} (n-1-i) \\
 &= \sum_{i=0}^{n-2} (n-1) - \sum_{i=0}^{n-2} i =
 \end{aligned}$$

$$\begin{array}{cccc}
 & \downarrow & \downarrow & & \\
 1 & 2 & 3 & 4 & \dots \\
 0 & 1 & 2 & 3 & \dots \\
 i & j & \uparrow & \uparrow &
 \end{array}$$

$$(n-1) \sum_{i=0}^{n-2} (i) - \sum_{i=0}^{n-2} i$$

$$= (n-1)(n-2-0+1) - \frac{(n-2)(n-1)}{2}$$

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

$$= (n-1) \left[(n-1) - \frac{(n-2)}{2} \right]$$

$$= (n-1) \left[\frac{2n-2-n+2}{2} \right]$$

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

Best - case : 1 comparison

Worst - case : $\frac{n^2}{2}$ comparisons

$$T(n)_{\text{worst case}} = O(n^2)$$

eg 3 Matrix Multiplication

for $i \leftarrow 0$ to $n-1$ do

for $j \leftarrow 0$ to $n-1$ do

$c[i, j] \leftarrow 0.0$

for $k \leftarrow 0$ to $n-1$ do

$c[i, j] \leftarrow c[i, j] + A[i, k] * B[k, j]$

return c

$$M(n) \in \Theta(n^3)$$

Mathematical analysis of Recursive Algo.

→ Important Recurrence Types

1. Decrease-by-one recurrences

eg: $n!$

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

linear

quadratic

2. Decrease-by-a-constant factor recurrences

eg: Binary search

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

→ Methods to solve recurrences

Substitution Method

Mathematical induction

Backward substitution

Recursion Tree Method

Master Method (decrease by constant factor recurrences)

eg: $n!$

$F(n)$

if $n == 0$ return 1

else return $F(n-1) * n$

Overall Time complexity: $O(n)$

$n = 4$



eg: counting no. of binary digits

2

1

$\text{BinRec}(n)$

if $n == 1$ return 1

else return $\text{BinRec}(\lfloor n/2 \rfloor) + 1$

eg: Tower of Hanoi

$$O(2^n)$$

0 1 2 3 4 5 6
89 45 68 90 29 34 17

10th February, 2023 ⑦

$n = 7$

Brute Force

└ Selection sort
└ Bubble sort

→ Selection sort

algo: for $i \leftarrow 0$ to $n-2$ do
 $\text{min} \leftarrow i$
 for $j \leftarrow i+1$ to $n-1$ do
 if $A[j] < A[\text{min}]$
 $\text{min} \leftarrow j$
 swap $A[i]$ and $A[\text{min}]$

$$\begin{aligned} C(n) &= \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 = \sum_{i=0}^{n-2} (n-1-(i+1)+1) = \sum_{i=0}^{n-2} (n-1-i) \\ &= (n-1) + (n-2) + \dots + (n-1-(n-2)) \\ &= \frac{(n-1)(n)}{2} = \frac{n^2-n}{2} \end{aligned}$$

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{n^2-n}{2(n^2)} &= \lim_{n \rightarrow \infty} \frac{2n-1}{2n} = \lim_{n \rightarrow \infty} \frac{1}{2} - \frac{1}{2n} = \frac{1}{2} \\ &O(n^2) \end{aligned}$$

→ Bubble sort

algo: for $i \leftarrow 0$ to $n-2$ do
 for $j \leftarrow 0$ to $n-2-i$ do
 if $A[j+1] < A[j]$
 swap $A[j]$ and $A[j+1]$

$$\begin{aligned} C(n) &= \sum_{i=0}^{n-2} \sum_{j=0}^{n-2-i} 1 = \sum_{i=0}^{n-2} [n-2-i-0+1] = \sum_{i=0}^{n-2} (n-1-i) \\ &= \frac{(n-1)n}{2} \Rightarrow O(n^2) \end{aligned}$$

→ Sequential search

Algo: sequentialsearch ($A[0..n]$, k)

$A[n] \leftarrow k$

$i \leftarrow 0$

while $A[i] \neq k$ do

$i \leftarrow i + 1$

if $i < n$ return i

else return -1

$O(n)$

→ String Matching

Algo: BruteForceStringMatch ($T[0..n-1]$, $P[0..m-1]$)

for $i \leftarrow 0$ to $n - m$ do

$j \leftarrow 0$

while $j < m$ and $P[j] = T[i+j]$ do

$j \leftarrow j + 1$

if $j == m$ return i

return -1

$O(nm)$

Decrease And Conquer

(Inductive/Incremental approach)

- Decrease by a constant (usually by 1)
 - insertion sort
 - graph traversal (DFS, BFS)
 - topological sorting
 - algs for generating permutations & subsets
- Decrease by a constant factor (usually by half)
 - binary search & bisection method
 - Exponentiation by Squaring
 - Multiplication à la russe
- Variable-size decrease
 - Euclid's algo
 - Selection by partition
 - Nim-like games

→ Insertion Sort

eg1: 6, 4, 1, 8, 5

```

6 | 4 1 8 5
4 6 | 1 8 5
1 4 6 | 8 5
1 4 6 8 | 5
1 4 5 6 8

```

eg2: 89, 45, 68, 90, 29, 34, 17

```

89 | 45 68 90 29 34 17
45 89 | 68 90 29 34 17
45 68 89 | 90 29 34 17
45 68 89 90 | 29 34 17
29 45 68 89 90 | 34 17
29 34 45 68 89 90 | 17
17 29 34 45 68 89 90

```

Algo: InsertionSort($A[0..n-1]$)//ip: $A[0..n-1]$ of n orderable elements//op: $A[0..n-1]$ sorted in non decreasing orderfor $i \leftarrow 1$ to $n-1$ do $v \leftarrow A[i]$ $j \leftarrow i-1$ while ($j \geq 0$ and $A[j] > v$) do $A[j+1] \leftarrow A[j]$ $j \leftarrow j-1$ $A[j+1] = v$

→ Time Efficiency

$$C_{\text{worst}}(n) = \frac{n(n-1)}{2} \in \Theta(n^2)$$

$$C_{\text{avg}}(n) \approx \frac{n^2}{4} \in \Theta(n^2)$$

$$C_{\text{best}}(n) = n-1 \in \Theta(n)$$

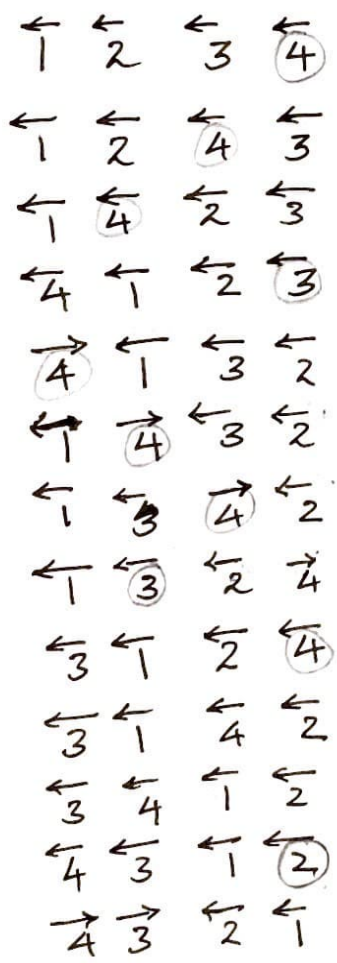
→ Space efficiency: in-place

→ Generating Permutations

Algo: Johnson Trotter(n)

// ip: 1 positive integer n
// op: 1 list of all permutations of $\{1, \dots, n\}$
initialize the first permutation with $\leftarrow 1 \leftarrow 2 \dots \leftarrow n$
while the last permutation has mobile element do
 find its largest mobile element k
 swap k with the adjacent element k's arrow points to
 reverse the direction of all elements that are larger than k
 add the new permutation to the list

* if a component k points to a smaller number adjacent to it, then k is mobile component.



→ Fake-Coin Problem

$$W(n) = W(\lfloor n/2 \rfloor) + 1 \quad \text{for } n > 1, W(1) = 0$$

$$W(n) = \log_2 n$$

→ Russian Peasant Multiplication

n & $m \rightarrow +ve \mathbb{I}$

$$\text{if } n \text{ (even)} \Rightarrow n \cdot m = (n/2) \cdot 2m$$

$$\text{if } n \text{ (odd)} \Rightarrow n \cdot m = ((n-1)/2) \cdot 2m + m$$

eg: $n \quad m \quad 50 \times 65 = 3250$

$$50 \quad 65$$

$$25 \quad 130$$

$$12 \quad 260 (+130)$$

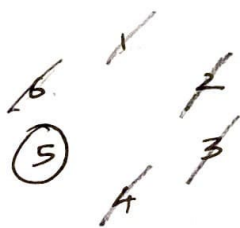
$$6 \quad 520$$

$$3 \quad 1040$$

$$1 \quad 2080 (+1040)$$

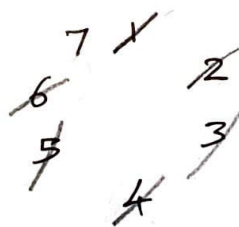
$$\text{Ans} = 2080 + 1040 + 130 = 3250$$

→ Josephus Problem



$$J(6) = 5$$

$$\textcircled{1} 10 \rightarrow \Rightarrow 101 = \textcircled{5}$$



$$J(7) = 7$$

$$111 \rightarrow \Rightarrow 111$$