

Continuation

Unit 1

Recurrence relation for factorial

Algorithm F(n)

F(n)

{

if $n == 0$

return 1

else

return $n * F(n-1)$

* Recurrence Relation for above algorithm is

$$T(n) = \begin{cases} 1 & n = 0 \\ n * T(n-1) & n > 1 \end{cases}$$

$$T(n) = n * T(n-1)$$

~~T(n)~~

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

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

$$T(n) = n * T(n-1)$$

~~Substituting equation (1) in equation (2)~~

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

~~Substituting equation (2) in equation (3)~~

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

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

* Process continues until k^{th} step

$$T(n) = n * (n-k)$$

$$T(n) = n * T(n-1) \quad \text{--- (1)}$$

$$T(n-1) = (n-1) * T(n-2) \quad \text{--- (2)}$$

$$T(n-2) = (n-2) * (n-1) * T(n-3) \quad \text{--- (3)}$$

* Substitute eq (2) in eq (1)

$$T(n) = n * (n-1) * T(n-2) \quad \text{--- (4)}$$

* Substitute ~~eq (2) in eq (4)~~ equation (3) in equation (4)

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

⋮

k^{th} step [continue till k^{th} step]

$$T(n) = k * (n-k) * (n-k) * (n-k) * T(n-k) \quad \text{--- (5)}$$

∴ whenever $n=1$ || $n=0$ program terminates

$$\therefore n-k=1$$

$$\boxed{k = n-1}$$

* Substitute k in equation (5)

$$T(n) = n-1 * n-(n-1) * n-(n-1) * n-(n-1) * T(n-(n-1))$$

$$= n-1 * \cancel{n-n+1} * \cancel{n-n+1} * \cancel{n-n+1} * T(1)$$

$$= n-1 * T(1)$$

$$= (n-1) * 1$$

$$= n-1$$

$$T(n) = n \quad \therefore O(n)$$

Master theorem

$$T(n) = aT(n/b) + O(n^k)$$

$$a > 1, b > 1, k \geq 0$$

1) If $a = 1$ $T(n) = O(n^{k+1})$

2) If $a > 1$ $T(n) = O(n^k \cdot a^{n/b})$

3) If $a < 1$ $T(n) = O(n^k)$

Problems

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

$$T(n) = 1T(n-1) + 1 \cdot n^0$$

$$a = 1, b = 1, k = 0$$

case 1:

$$T(n) = O(n^{0+1})$$

$$T(n) = \underline{\underline{O(n)}}$$

2) Algorithm Test(n)

{

if (n > 0)

{

printf("%d ", n);

Test(n-1);

Test(n-2);

}

}

Recurrence relation for

is

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

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

$$\therefore T(n) = 2T(n-1) + 1 \cdot n^0$$

$$a = 2, b = 1, k = 0$$

case 2:

$$T(n) = O(n^0 \cdot 2^{n/1}) = \underline{\underline{O(2^n)}}$$

Problem 3

```
void Test(n)
```

```
{
```

```
    if (n == 0)
```

```
    {
```

```
        for (i = 0; i < n; i++)
```

```
        {
```

```
            Statement;
```

```
        }
```

```
        Test(n-1);
```

```
    }
```

```
}
```

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

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

~~Recurrence Relation~~

~~Recurrence Relation~~

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

$$T(n) = T(n-1) + O(n^1)$$

$$a=1 \quad b=1 \quad k=1$$

Case 1:

$$T(n) = O(n^{1+1})$$

$$T(n) = O(n^2)$$

Problem 4: Dividing Function

~~Test~~

Algorithm BinRec(n)

// Input: A positive decimal integer n

// Output: TL number of binary digits in n's binary representation

If $n == 1$

return 1

else

return BinRec($n/2$) + 1

* Recurrence Relation:

$$T(n) = \begin{cases} T(n/2) + 1 & n > 1 \\ 1 & n = 1 \end{cases}$$

$$T(n) = T(n/2) + 1 \quad \text{--- (1)}$$

$$T(n/2) = T(n/2^2) + 1 \quad \text{--- (2)}$$

$$T(n/2^2) = T(n/2^3) + 1 \quad \text{--- (3)}$$

* Substitute eq (2) in eq (1)

$$\begin{aligned} T(n) &= T(n/2) + 1 \\ &= T(n/2^2) + 1 + 1 \\ &= T(n/2^3) + 1 + 1 + 1 \end{aligned}$$

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

Understanding the Problem

- * First thing you need to do before designing an algorithm is to understand completely the problem given.
- * Read the problem's description carefully and ask questions if you have any doubts about the problem.
- * An input to an algorithm specifies an instance of the problem the algorithm solves.
- * Verify what should be the input and what should be the output that you are expecting.
- * A correct algorithm is not one that works most of the time, but one that works correctly for all legitimate inputs.

Ascertaining the capabilities of a computational Design

- * Decide the capabilities of the computational device the algorithm is intended for.
 1. Sequential algorithm
 2. Parallel algorithm
- * Instructions are executed one after the other, one operation at a time. Sequential algorithm.
- * Instructions are executed concurrently or parallelly, called as parallel algorithm.

choosing b/w exact and approximate problem solving

* next principal decision is to choose b/w solving the problem

- exactly or
- ~~and~~ solving it approximately.

* we have two kinds of algorithm.

• ~~exactly~~

1. exact algorithm

2. approximation algorithm.

* choose the appropriate data structure for a given algorithm.

Algorithm Design Technique

* An algorithm design technique is a general approach to solving problems algorithmically that is applicable to a variety of problems from different areas of computing.

* There are various Algorithm Design Techniques are available such as

1. Brute force
2. divide and conquer
3. decrease and conquer
4. Dynamic programming
5. Greedy Method etc.

* Choose the appropriate Algorithm Design Technique for the given problem.

Designing an algorithm and data structure

- * Some design technique can be ~~used~~ simply ~~map~~ inapplicable to \forall problem in question
- * Several techniques need to be combined and there are algorithm that are hard to pinpoint \forall applications of \forall ~~the~~ known design technique.
- * Pay close attention to choosing data structures appropriate for \forall operations performed by \forall algorithm.
- * There are 3 ways to design an algorithm
 1. Writing natural language
 2. Writing Flow chart
 3. Writing pseudocode
- * ~~Pseudocode~~ pseudocode is usually more precise than natural language.

Proving an Algorithm's correctness

- * Once an algorithm has been specified, you have to prove its correctness.
- * You have to prove that \forall algorithm produces a required results for every legitimate input in a finite amount of time.

* A common technique for proving correctness is to use mathematical induction because an algorithm's iterations provide a natural sequence of steps needed for such proofs.

Analyzing algorithm

* We usually want our algorithms to possess several qualities.

* After correctness, by far the most important is efficiency.

* There are two kinds of algorithm efficiency

1. Time efficiency → indicating how fast the algorithm runs

2. Space efficiency → indicating how much extra memory it uses

Writing an algorithm

* Most algorithms are destined to be ultimately implemented on a computer.

* programming an algorithm presents both a peril and an opportunity

* Validate program

* Test the computer programs.

Tower of Hanoi

```
void ToH(int n, int A, int B, int C)
```

```
{
```

```
if (n == 1)
```

```
{
```

```
if (n == 1)
```

```
{
```

```
printf("move disk from A to C");
```

```
printf("move disk from A to C", A, C);
```

```
}
```

```
else
```

```
{
```

```
ToH(n-1, A, C, B);
```

```
printf("move disk from A to C", A, C);
```

```
ToH(n-1, B, A, C);
```

```
}
```

```
}
```

Recursive Equation

$$T(n) = \begin{cases} 2T(n-1) + 1 & n > 0 \\ 1 & n = 1 \end{cases}$$

- * Tower of Hanoi is one of the mathematical game where it has 3 rods and n disk
- * The objective of the game is to move the entire disk to another rod with following rules:
 1. Only one disk can be moved at a time.
 2. Disk can be moved if it is in the uppermost disk on a stack.
 3. No disk may be placed on top of a smaller disk.

$$T(n) = 2T(n-1) + 1 \quad \sim (1)$$

$$2T(n-1) = 4T(n-2) + 1 + 1 \quad \sim (2)$$

$$4T(n-2) = 8T(n-3) + 1 + 2 \quad \sim (3)$$

* Substitute (2) in equation (1)

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

$$= 4T(n-2) + 2 + 1$$

$$= 8T(n-3) + 1 + 3$$

$$= 2^2 \{ 2T(n-3) + 1 \} + 3$$

⋮

Continue. till k^{th} step.

$$= 2^k \{ kT(n-k) + k \} + k \quad \sim (4)$$

∴ program terminates when $n-k=1$

$$\therefore \frac{n-k=1}{k=n-1}$$

* Substitute k in equation (4)

$$= 2^{n-1} \{ (n-1) \times T(n-(n-1)) + (n-1) \} + n-1$$

$$= 2^{n-1} \{ (n-1) \times T(1) + n-1 \} + n-1$$

$$= 2^{n-1} \{ n-1 \times 1 + n-1 \} + n-1$$

$$= 2^{n-1} \{ n-1 \times T(1) + n-1 \} + n-1$$

$$= 2^{n-1} \{ n-1 \times 1 + n-1 \} + n-1$$

$$= 2^{n-1} \{ n-1 \times 1 + n-1 \} + n-1$$

$$T(n) = 2^{n-1} \{ n-1 + n-1 + n-1 \} \sim 1 + (n-1)T2 = (n)T$$

$$T(n) = 2^{n-1} \{ 3n-3 \} \sim 1 + 1 + (2-n)T2 = (2-n)T2$$

$$T(n) = 2^{n-1} \{ 3n-3 \}$$

$$T(n) = 2^n \times \frac{1}{2} \times \{ 3n-3 \}$$

$$T(n) = 2^{n-1}$$

$$\underline{\underline{O(2^n)}}$$

By Master Theorem

Order of growth

- * order of growth provides the behavior of an algorithm
- * order of growth allows us to compare the relative performance of alternate algorithms.
- * order of growth means, when input size of a given algorithm increases, the computational time also increases.
- * \therefore Computational time is directly proportional to input size
- * order of growth

~~Order of growth~~

$$1 \ll n! \ll 4^n \ll 2^n \ll n^4 \ll n^3 \ll n^2 \ll n \log n \ll n \ll \log n \ll \log \log n \ll \frac{\log n}{\log \log n} \ll 1$$