Algorithms CSE 2415

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Introduction

Algorithm is a sequence of steps / step-by-step procedure to solve a problem.

Properties of Algorithm:

- Specific input
- Specific output
- Definiteness
- Finiteness
- Effectiveness

Time and Space Complexity

Examples:

```
Algorithm:
                                     space complexity
                                                                       Space complexity
                                     cost
                                               repeat
                                                                       cost repeat total
                                                            total
int i, j;
                                                                       i=4
                                                                                      4
                                      1
                                                  1
                                                              1
                                                                                1
for(i = 0; i < n; i++){
                                                                       j= 4
                                    1+1+1
                                              1+(n+1)+n
                                                            2n+2
                                                                                1
                                                                                      4
    for(j = 0; j < n; j++)
                                           (1+(n+1)+n)+n
                                                                       n=4
                                    1+1+1
                                                           2n^2+2n
                                                                                1
                                                                                      4
        printf(" \%d ", i+j);
                                      1
                                                n^2
                                                             n^2
}
                                     F(n) = 3n^2 + 3n + 1
                                                                       S(n) = 12
                                                                       SC -> O(1)
                                     TC \rightarrow O(n^2)
```

```
Algorithm:
int i, j, n, A[i][j], B[i][j], C[i][j];
for(i = 0; i < n; i++){
    for(j = 0; j < n; j++)
        C[i][j] = A[i][j] + B[i][j];
}
space complexity
                                     Space complexity
cost
          repeat
                        total
                                      cost repeat total
 1
             1
                          1
                                      i=4
                                               1
                                                       4
                                                       4
1+1+1
         1+(n+1)+n
                         2n+2
                                               1
                                      j=4
1+1+1
        n+n(n+1)+n^2
                        2n^2+2n
                                      n=4
                                               1
                                                       4
            n^2
                         n^2
                                     A[][]
                                                      4n^2
 1
                                              4*n*n
                                     B[][]
                                              4*n*n
                                                      4n^2
                                     C[][]
                                                      4n^2
                                              4*n*n
F(n) = 3n^2 + 4n + 3
                                     S(n) = 12n^2 + 12
TC \rightarrow O(n^2)
                                      SC \rightarrow O(n^2)
Algorithm:
int i, n;
for(i = 0; i < n; i++)
    printf(" \%d ", 2*i);
space complexity
                                     Space complexity
                                       cost repeat total
cost
            repeat
                         total
 1
              1
                           1
                                       i=4
                                                1
                                                        4
1+1+1 1+(n/2)+1+(n/2)
                                                1
                                                        4
                          n+2
                                       n=4
 1
             n/2
                          n/2
```

F(n) = (3n/2) + 3

 $TC \rightarrow O(n)$

S(n) = 8

SC -> O(1)

```
Algorithm:
int p=0, i, n;
for(i = 1; i <= p; i++)
    p+=1;
Step analysis:
        р
0
       0+1
    0+1+2
1
2 0+1+2+3
3 0+1+2+3+4
k = 0+1+2+3+4+...+k = k*(k+1)/2
assume , p>n where step number is k and p = k*(k+1)/2
       k(k+1)/2 = n
     \Rightarrow k(k+1) = 2n
     => k^2 + k - 2n = 0
     \Rightarrow k^2 = 2n [ removed k as k^2 > k ]
     => k = sqrt(2n)
     => k = sqrt(2) * sqrt(n)
thus O(sqrt(n))
int i, n;
for(i =1; i<n; i*=2)
    printf("%d", i);
Step Analysis:
steps
            2*i
1
             1
 2
             2
 k
            2^k
                         [actual value is 2<sup>(k-1)</sup>]
```

```
assume, code stopped at step k where , i 2^k
     2^k > n
   =>2^k = n
   =>\log 2^k = \log n
   =>k = log n [log 2 ^ x = x , here base is always 2 causo fo binary ]
   \rightarrow O(\log n)
 int i, n;
 for(i =n; i > 1; i/=2)
     printf("%d", i);
 Step Analysis:
 steps
              i/2
  1
              n
              n/2
             n/2^k
  n/2^k < 1
=>2^k < n
=>2^k = n
=>\log 2^k = \log n
\Rightarrow k = log n
 int p=0, i;
 for(i =0; i<n; i*=2)
                         ----> O(log n)
     p++;
 for(i=p; i>1; i/=2) '---> p = log n
printf("%d", i); '---> O(log log n)
```

Time Complexity Cheat Sheet

```
Constant TC
             - 0(1)
                             -> for( i=0; i<k; i++ )
              - 0(sqrt(n))
                             -> for( i=0; p<n; i++ )
Linear TC
              -0(n)
                             -> for( i=1; i<n; i++ )
                             -> for( i=0; i<n; i+=2)
                             -> for( i=n; i>n; i-=5)
Logarithm TC - O(log n)
                             -> for( i=0; i<n; i*=2)
                             -> for( i=n; i>1; i/=2)
              - O(log_a n)
                             -> for( i=1; i<n; i*=a)
Polynomial TC - O(n^2)
                             -> for(--) for(--)
              -0(n^3)
                             -> for(--) for(--) for(--)
Exponential TC- O(2^n),O(n^n)\rightarrow Any recursive function
```

Recursion:

Substitution method:

```
-> It solves all types of recursive problems
-> It always gives right answers
-> Takes more time then Master method

Master method:
-> It solves Specific types of recursive problems
-> Format: T(n) = aT(n/b) + f(n); where a >= 1 and b > 1

Different cases:
    1. f(n) < n log_b a;
        T(n) = (theta) (n log_b a)

2. f(n) = n log_b a;
        T(n) = (theta) (n log_b a . log n)

3. f(n) > n log_b a;
        T(n) = (theta) (f(n))
```

Examples:

 EQ^n :

$$T(n) = \begin{cases} t(n-1) + n &, n > 1\\ 1 &, n \le 1 \end{cases}$$

 Sol^n : We have,

$$T(n) = t(n-1) + n$$
 —(1)

Replace n by n-1 in equation (1):

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

Putting equation (2) in equation (1) we get,

$$T(n) = t(n-2) + (n-1) + n \quad -(3)$$

Similarly, replacing n by n-2 in equation (1) we get,

$$T(n-2) = t(n-3) + (n-2) \quad -(4)$$

Putting equation (4) in equation (3) we get.

$$T(n) = t(n-3) + (n-2) + (n-1) + n \quad -(5)$$

So, the general equation may be,

$$T(n) = t(1) + 2 + 3 + \dots + (n-2) + (n-1) + n$$
 (6)

Given T(1) = 1 we get from equation (6),

$$T(n) = t(1) + 2 + 3 + \dots + (n-2) + (n-1) + n$$

$$= 1 + 2 + 3 + \dots + (n-2) + (n-1) + n$$

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

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

Here, n^2 is the time complexity. Thus, the answer is $O(n^2)$.

 EQ^n :

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

 Sol^n : We have,

$$T(n) = t\left(\frac{n}{2}\right) + c \quad -(1) \tag{7}$$

Replace n by n/2 in equation (1):

$$T\left(\frac{n}{2}\right) = t\left(\frac{n}{4}\right) + c \quad -(2)$$

Putting equation (2) in equation (1) we get,

$$T(n) = t\left(\frac{n}{4}\right) + c + c \quad -(3)$$

Similarly, replacing n by n/4 in equation (1) we get,

$$T\left(\frac{n}{4}\right) = t\left(\frac{n}{8}\right) + c \quad -(4)$$

Putting equation (4) in equation (3) we get,

$$T(n) = t\left(\frac{n}{8}\right) + c + c + c \quad -(5)$$

So, the general equation may be,

$$T(n) = t\left(\frac{n}{2^k}\right) + kc \quad -(6)$$

Assume that in the k-th step, $\frac{n}{2^k} = 1$

$$\Rightarrow \frac{n}{2^k} = 1$$

$$\Rightarrow n = 2^k$$

$$\Rightarrow \log n = \log 2^k$$

$$\Rightarrow k = \log n$$

Using the value of k in equation (1), we get,

$$T(n) = t(1) + kc = 1 + (\log n) \cdot c \tag{13}$$

Here, t(1) is the base case complexity. Thus, the answer is $O(\log n)$.

 EQ^n :

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

 Sol^n :

We have,

$$T(n) = 2t\left(\frac{n}{2}\right) + n \quad -(1) \tag{14}$$

Replace n by n/2 in equation (1):

$$T\left(\frac{n}{2}\right) = 2t\left(\frac{n}{4}\right) + \frac{n}{2} \quad -(2) \tag{15}$$

Putting equation (2) in equation (1) we get,

$$T(n) = 2\left(2t\left(\frac{n}{4}\right) + \frac{n}{2}\right) + n\tag{16}$$

$$T(n) = 4t\left(\frac{n}{4}\right) + 2n \quad -(3)$$

Similarly, replacing n by n/4 in equation (1) we get,

$$T\left(\frac{n}{4}\right) = 2t\left(\frac{n}{8}\right) + \frac{n}{4} \quad -(4)$$

Putting equation (4) in equation (3) we get,

$$T(n) = 4(2t\left(\frac{n}{8}\right) + \frac{n}{4}) + 2n\tag{19}$$

$$T(n) = 8t\left(\frac{n}{8}\right) + 3n \quad -(5)$$

So, the general equation may be,

$$T(n) = 2^k t\left(\frac{n}{2^k}\right) + kn \quad -(6)$$

Given T(1) = 1,

$$T(n) = 2^k T\left(\frac{n}{2^k}\right) + kn$$
 we get , $n = 2^k$ —(a)
$$\log n = \log 2^k$$
 or, $k = \log n$ —(b)

Since $n = 2^k$, we have $k = \log n$

thus,
$$T(n) = nT\left(\frac{n}{n}\right) + n\log n$$
 [using equations (a) and (b)]
$$T(n) = n + n\log n$$
 thus, we get $TC = O(n\log n)$

Master Method Examples:

 EQ^n :

$$T(n) = \begin{cases} 2t\left(\frac{n}{2}\right) + n &, n > 1\\ 1 &, n = 1 \end{cases}$$

 Sol^n :

$$a = 2, \quad b = 2, \quad f(n) = n$$

 $n^{\log_b a} = n^{\log_2 2} = n^1 = n$

$$f(n) = n^{\log_b a}$$

$$TC = \theta(n^{\log_2 2} \cdot \log n)$$
$$= \theta(n \cdot \log n)$$

 EQ^n :

$$T(n) = \begin{cases} T\left(\frac{n}{2}\right) + c &, n > 1\\ 1 &, n \le 1 \end{cases}$$

 Sol^n :

$$a = 1, \quad b = 2, \quad f(n) = c$$

 $n^{\log_b a} = n^{\log_2 1} = n^0 = 1$

$$f(n) = n^{\log_b a}$$

$$TC = \theta(n^{\log_2 1} \cdot \log n)$$
$$= \theta(1 \cdot \log n)$$
$$= \theta(\log n)$$

 EQ^n :

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

 Sol^n :

$$a = 1, \quad b = 1, \quad f(n) = n$$

Since, b is not greater then a, we cant use master method to find the complexity.

Divide And Conquer Method:

3 step to solve level of problem:

1. Divide:

Divide the problem into problems into a number of subproblems that are similar to the problem.

2. Conquer:

Conquer the subproblems by solving them recursively. If the subproblem's size is small enough, just solve it in straight forward manner.

3. Combine:

Combine the solutions of the subproblems to get the solution of the original problem.

Example:

For MergeSort it works as follows:

- 1. Divide the n-element array into two n/2 elements to Sort.
- 2. Sort the two subsequences recursively using MergeSort.
- 3. Merge the two sorted subsequences.

Time Complexity function Comparison:

Big O cheat sheet

 ${\it TC\ class}$:

$$1 < \log n < \sqrt{n} < n < n \log n < n^2 < n^3 < \ldots < 2^n < 3^n < \ldots < n^n$$

