

Innopolis University, S19

Assignment 2. Theory part

Data structures and
algorithms

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1) Applications of Master Theorem

a) $T(n) = 16 T\left(\frac{n}{4}\right) + n$
 $\alpha = 16, \beta = 4, f(n) = n$

$$f(n) = O\left(n^{\log_4 16 - \epsilon}\right) = O\left(n^{\log_4 16 - \epsilon}\right) = O\left(n^{2-\epsilon}\right) \text{ for } \epsilon > 0$$

Then $T(n) = \Theta(n^2)$

b) $T(n) = T\left(\frac{n}{2}\right) + 2^n$
 $\alpha = 1, \beta = 2, f(n) = 2^n$
 $n^{\log_2 \alpha} = n^{\log_2 1} = n^0 = 1$
 $f(n) = \Theta(2^n) = \Omega(n^{0+\epsilon}) \text{ for } \epsilon > 0$
 $f\left(\frac{n}{2}\right) \leq c f(n)$
 $2^{n/2} \leq c 2^n$

$$\frac{1}{2^{n/2}} \leq c$$

since $n > 0, \frac{1}{2^{n/2}} \in (0; 1) \rightarrow \frac{1}{2^{n/2}} < 1 \Rightarrow \exists c < 1$ that satisfies inequality

$$\Downarrow$$
$$T(n) = \Theta(f(n)) = \Theta(2^n)$$

c) $T(n) = 2T\left(\frac{n}{2}\right) + \frac{n}{\log n}$
 $n^{\log_2 \alpha} = n^{\log_2 2} = n$

$$\frac{n/\log n}{n} = \frac{1}{\log n} \neq n^\epsilon \quad \forall \epsilon \in \mathbb{R} \Rightarrow$$

\Rightarrow there is non-polynomial difference

between $f(n)$ and $n^{\log_2 \alpha}$ \Rightarrow The master theorem is not applicable

d) $T(n) = 4T\left(\frac{n}{2}\right) + n^2$

$$\alpha = 4 \quad b = 2 \quad f(n) = n^2$$

$$n^{\log_b \alpha} = n^2$$

$$f(n) = n^2 = \Theta(n^2) \Rightarrow T(n) = \Theta(n^2 \log n)$$

2) Solving a problem (dynamic prog.)

1) Let $\text{dist}[i]$ be a minimum distance from station 1 to station i :

$1 \leq i \leq n$, $\text{dist}[n]$ is the total cost

2) The step:

$$\text{dist}[i] = \begin{cases} 0, & i=1 \\ \min(\{d[j] + f_{ji} \mid \forall j : 1 \leq j < i\}), & i \neq 1 \end{cases}$$

3) The proof of correctness:

Strong induction:

a) base case: $\text{dist}[1] = 0$ - true, because you do not have to move from station 1 in order to arrive to station 1

b) Suppose that $\text{dist}[1], \text{dist}[2], \dots, \text{dist}[k]$ are minimum distances from station 1 to

stations $1, 2, \dots, K$ correspondingly

c) Prove that $\text{dist}[K+1] = \min(\{\text{dist}[i] + f_{i,K+1} \mid \forall i: 1 \leq i \leq K\})$ is the minimum distance from station 1 to $K+1$.
Since we supposed that $\text{dist}[1], \dots, \text{dist}[K]$ are lengths of shortest distances to stations $1..K$, then $\min(\text{dist}[1] + f_{1,K+1}, \dots)$ is indeed the shortest path

4) As was shown in step 2 and proved at step 3.a, the base case is calculating a distance to the source station. Distance is always 1, therefore, computation complexity $O(1)$

5) int $\text{distance}[1..N_STATIONS]$

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1.  $\text{distance}[1] = 0$  // 1  
2. for  $i = 2$  to  $N\_STATIONS$  do // n  
3.   int  $\text{min\_dist} = \text{INFINITY}$  //  $n-1$   
4.   for  $j = i-1$  down to 1 do //  $\sum_{i=2}^n i$   
5.      $\text{min\_dist} = \min(\text{min\_dist}, \text{distance}[j] + f_{j,i})$  //  $\sum_{i=2}^n (i-1)$   
6.   end  
7.    $\text{distance}[i] = \text{min\_dist}$  //  $n-1$   
end
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$$T(n) = C_1 + n \cdot C_2 + (n-1)C_3 + C_4 \cdot \sum_{i=2}^n i + C_5 \sum_{i=2}^n (i-1) + C_7 (n-1)$$

$$T(n) = C_1 + n \cdot C_2 + (n-1)C_3 + C_4 \frac{(n+2)(n-1)}{2} + C_5 \frac{n(n-1)}{2} + C_7 (n-1)$$

After combining like terms it becomes obvious that $T(n)$ is a 2nd degree polynomial $\Rightarrow T(n) = O(n^2)$