

1: Solving Recurrences

Stated Below is the Master Theorem which will be used exhaustively in the following answers.

$$T(n) = \begin{cases} c & \text{if } n < d, \\ aT(n/b) + f(n) & \text{if } n \geq d, \end{cases}$$

where $a \geq 1, b > 1$, and d are integers and c is a positive constant. Let $\nu = \log_b a$.

Case (i) $f(n)$ is “definitely smaller” than n^ν : If there is a small constant $\epsilon > 0$, such that $f(n) \preceq n^{\nu-\epsilon}$, that is, $f(n) \prec n^\nu$, then $T(n) \sim n^\nu$.

Case (ii) $f(n)$ is “similar in size” to n^ν : If there is a constant $k \geq 0$, such that $f(n) \sim n^\nu (\log n)^k$, then $T(n) \sim n^\nu (\log n)^{k+1}$.

Case (iii) $f(n)$ is “definitely larger” than n^ν : If there are small constants $\epsilon > 0$ and $\delta < 1$, such that $f(n) \succeq n^{\nu+\epsilon}$ and $af(n/b) \leq \delta f(n)$, for $n \geq d$, then $T(n) \sim f(n)$.

(a) $T(n) = 4T(n/4) + n$.

Here $a = 4, b = 4, f(n) = n$.

a, b both are greater than 1,

Hence applying master theorem:

Compare $f(n)$ with $n^{\log_b a}$

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

Condition 2 of Master Method is valid.

Therefore, $T(n) = O(n^{\log_b a} \log n) = O(n \log n)$

(b) $T(n) = 4T(n/4) + 1$.

a, b both are greater than 1

Hence applying master theorem:

Compare $f(n)$ with $n^{\log_b a}$

$n^{\log_b a} = n$

We know that $f(n)=1$ which is in the form $O(n^{\log_4 4 - \epsilon})$ where ϵ is a constant.

Condition 1 for master method holds

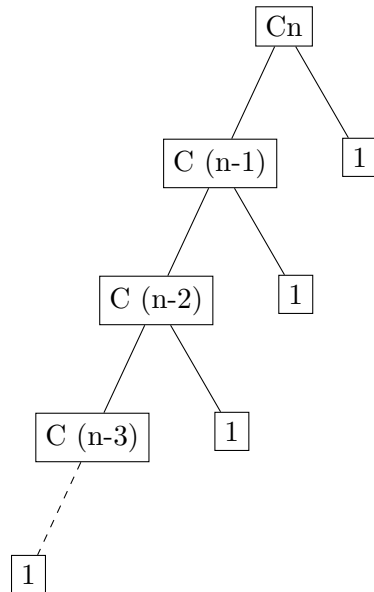
Hence, $T(n) = O(n^{\log_4 4}) = O(n)$

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

Solution using Recursion Tree

Cost of all subproblem is C_n , where n is subproblem.

The tree structure is



Size of sub prob decreases by 1 at each node
 depth at which size of subproblem = 1 is 'n'.
 depth of tree = n

Recurrence = Sum of costs at each level =

$$Cn + C(n-1) + C(n-2) + \dots + 1$$

$$\Rightarrow n \cdot Cn - e, \text{ where } e = \text{sum of constant terms}$$

$$\Rightarrow Cn^2 - e$$

$$\Rightarrow O(n^2)$$

(d) $T(n) = T(n/3) + T(n/2) + \sqrt{n}$.

Guess: $O(n)$

$$T(n) < d(n/3) + d(n/2) + \sqrt{n}$$

$$T(n) < d(5n/6) + \sqrt{n}$$

$$T(n) < Cn + \sqrt{n}$$

$$T(n) < Cn$$

$$T(n) = O(n)$$

(e) $T(n) = T(\sqrt{n}) + 4$.

$$\text{let } m = \log n \Rightarrow n = 2^m$$

Replacing n with m

$$T(2^m) = T(2^{m/2}) + 1$$

$$\text{let } T(2^m) = S(m)$$

$$S(m) = S(m/2) + 1$$

Applying Master,

$$a = 1, b = 2, f(n) = 1$$

$$n^{\log_b a} = n^{\log_2 1} = 1 = f(n)$$

$$\text{Condition 2 holds} \Rightarrow S(m) = O(\log m) = O(\log \log n)$$

(f) Suppose we have $T(n) = 3T(n/2) + g(n)$.

Analyze the behavior when (a) $g(n) = n^2$ (b) $g(n) = n$, and (c) $g(n) = n^{\log_2 3}$.

For 3 cases:

Size of subproblem at every level = $n/2^i$, where 'i' denotes level.

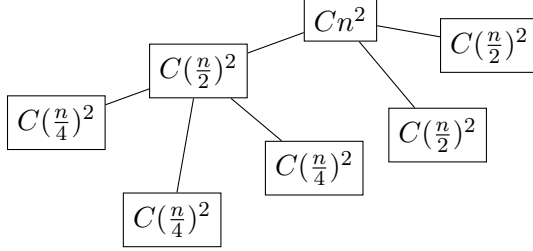
When $\frac{n}{2^i} = 1$, lets find leaf node

$$\implies i = \log_2 n$$

levels = $\log_2 n + 1$

nodes at depth i = 3^i

(a) when $g(n) = n^2 \implies T(n) = 3T(n/2) + n^2$



Cost of a subproblem at each level = $(\frac{n}{2^i})^2$

Cost = each level = $3^i (\frac{n}{2^i})^2 = (\frac{3}{4})^i * n^2$

Cost = leaf node = $3^i = 3^{\log_2 n} = n^{\log_2 3}$, where leaf = T(1) to the cost. So cost at level $\log n$ is $\theta(n^{\log_2 3})$.

$$T(n) = cn^2 + \frac{3}{4}cn^2 + \frac{9}{16}cn^2 + \dots + (\frac{3}{4})^{\log_2(n-1)}cn^2 + \theta(n^{\log_2 3})$$

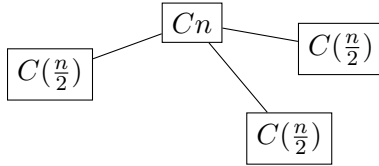
Infinite GP case:

$$T(n) < \frac{cn^2}{1-\frac{3}{4}} + \theta(n^{\log_2 3})$$

$$\implies O(n^2).$$

Condition 3 of the Master Theorem:

(b) when $g(n) = n \implies T(n) = 3T(n/2) + n$



Cost of a subproblem = $(\frac{n}{2^i})$

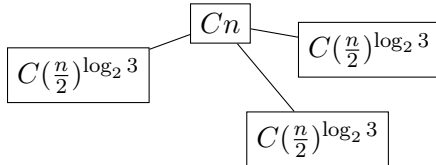
Cost = each level = $3^i (\frac{n}{2^i}) = (\frac{3}{2})^{i*n}$

Cost = leaf node = $3^i = 3^{\log_2 n} = n^{\log_2 3}$, where leaf = T(1) to the cost. So cost at level $\log n$ is $\theta(n^{\log_2 3})$.

$$\implies T(n) = O(n^{\log_2 3})$$

Condition 1 of the Master Theorem:

(c) when $g(n) = n^{\log_2 3} \implies T(n) = 3T(n/2) + n^{\log_2 3}$



Cost = leaf node = $3^i = 3^{\log_2 n} = n^{\log_2 3}$, where leaf = T(1) to the cost. So cost at level $\log n$ is $\theta(n^{\log_2 3})$.

$$T(n) = cn^{\log_2 3} + \frac{3}{2}cn^{\log_2 3} + \frac{9}{4}cn^{\log_2 3} + \dots + (\frac{3}{2})^{\log_2 n - 1}cn^{\log_2 3} + \theta(n^{\log_2 3})$$

$$T(n) = (n^{\log_2 3}) \sum_{j=0}^{\log n - 1} \left(\frac{3}{2}\right)^{\log_2 3}$$

$$T(n) = \mathcal{O}(n^{2(\log_2 3)} \log n)$$

Condition 2 of the Master Theorem.

2: Sorting "nearby" numbers

Algorithm: The given case can be solved using the concept of Counting Sort. The first step is to scan through the array to determine the minimum and the maximum element. Let B be a new array of size M created that has each value as a constant 0. The array is scanned again and for each number in A[i], we increment $B[\min_i\{A[i]\} + A[i]]$. A new array C[0..n-1] is created. In the final step we scan through the array B[] and for each number B[i], we can place B[i] values of i into the coming B[i] null slots of C[].

Thus we count the exact time value i occurs in the array and store the data in B[i]. We get the sorted array when we scan through B[i] and get the values in order.

Correctness: All the elements of A[] is stored in C[] in sorted order. So accuracy is 100%

Running time: During the scan process, we scan the elements of array with sizes 'n' and 'M' regularly, so it takes $O(n + M)$ times.

3: Selecting in an Union

Algorithm: Check the sum of middle index of array A and B.

if $sum < k$ and *middle element of* $k > \text{middle element of } B$

discard the first half of B and new value of $k = K - (\text{index of } mid - \text{elem of } B) - 1$

else if

mid-elem of B is greater

discard the first half of A and new value of $k = K - (\text{index of } mid - \text{elem of } A) - 1$

repeat till length of A or B is zero.

Correctness: We get the desired kth smallest element using the above steps, thus we have the 100% correctness.

Running time: In every repetition, we are performing computations on middle half of list, and based on if-else we remove half list in one of the array. Thus this dividing process takes $O(\log n)$ time complexity.

4: Closest Pair

(a) Let the instances be 1,3,8 separated by an imaginary axis with instances 9,15,19. Now if we try to find the smallest distance, it comes as 4 between 15 and 19 and 2 between 1 and 3. So the closest distance is considered as 2. But actually points 8 and 9 are separated by just a point distance. So the consideration goes wrong in this case.

(b) Steps for proof:

We will consider a square.

We need to show that any 2 points are at a minimum distance d .

Let us consider a square area.

Let a point be placed on any corner.

The next point cannot be placed anywhere other than the other remaining vertices of the square.

Thus this way we get 4 different points placed on the vertices of the square area.

If we consider any 5th point, the distance of 5th point from other 4 points will be strictly smaller than d .

Thus proved.

(c) Steps involved and time computations:

finding median $x = O(n)$

recurse the two sub probs of size $n/2 = 2T(n/2)$

discard points $= O(n)$

sort y values $= O(n \log n)$

iterate through a list of y variables and computation for each $= O(n)$

Thus total $=$

$$T(n) = 2T(n/2) + O(n \log n).$$

For bound -

$$T(n) = 2T(n/2) + O(n \log n).$$

$$T(n) = \sum_{i=1}^m \log(n/2^{i-1}) + \text{where}(m) = \log n$$

$$T(n) \leq O(2^{\log n}) + O(n(\log n + \log(n/2) + \log(n/4) + \dots + 1))$$

$$T(n) \leq O(n) + O(n \log(\log n))$$

$$T(n) \leq O(n \log^2 n)$$

5: Linear Time Median

(a) Using the recursive formula we know that $T(n) = O(n) + T(n/5) + T(7n/10)$

We assume that $T(n) = (a * n) + T(n/5) + T(7n/10)$

$$C * n \geq T(n/5) + T(7n/10) + a * n$$

$$C * n \geq C * n/5 + C * 7n/10 + a * n$$

$$C > 9 * C/10 + a$$

$$C/10 \geq a$$

$$C \geq 10 * a$$

so $T(n) = O(n)$

Here if we divide the group into 3 then:

$$T(n) = O(n) + T(n/3) + T(2n/3) \text{ so } T(n) > O(n) \dots$$

if groups are divided into more than 5, Value of constant 5 is more, so that is the most optimal solution.

In which case running time is $T_{\text{median}}(n/5) + O(n)$

(b) Recurrence for finding the near median is $T(N) \leq T(N/S) + T(B/N) + O(N)$ as we see,

$T(N/S)$ = cost of finding median of A_{medians} and N is the cost of partitioning algo.

$T(B/N)$ = cost of final recursive call

B = fraction of (BN) i.e. worst case length of either A_l or A_r

$M = \text{median of } A_{\text{medians}}$

M must $= 1/2(N/5)$ except the median itself, not including 5 elements from last median cluster.

$$G - 2 = 1/2(N/5)$$

now at least 3 elements are smaller or equal to M

$$3(G - 2) = 1/2(N/5) \text{ i.e } (3N/10) - 9 \geq N/4$$

This proves the algo.