

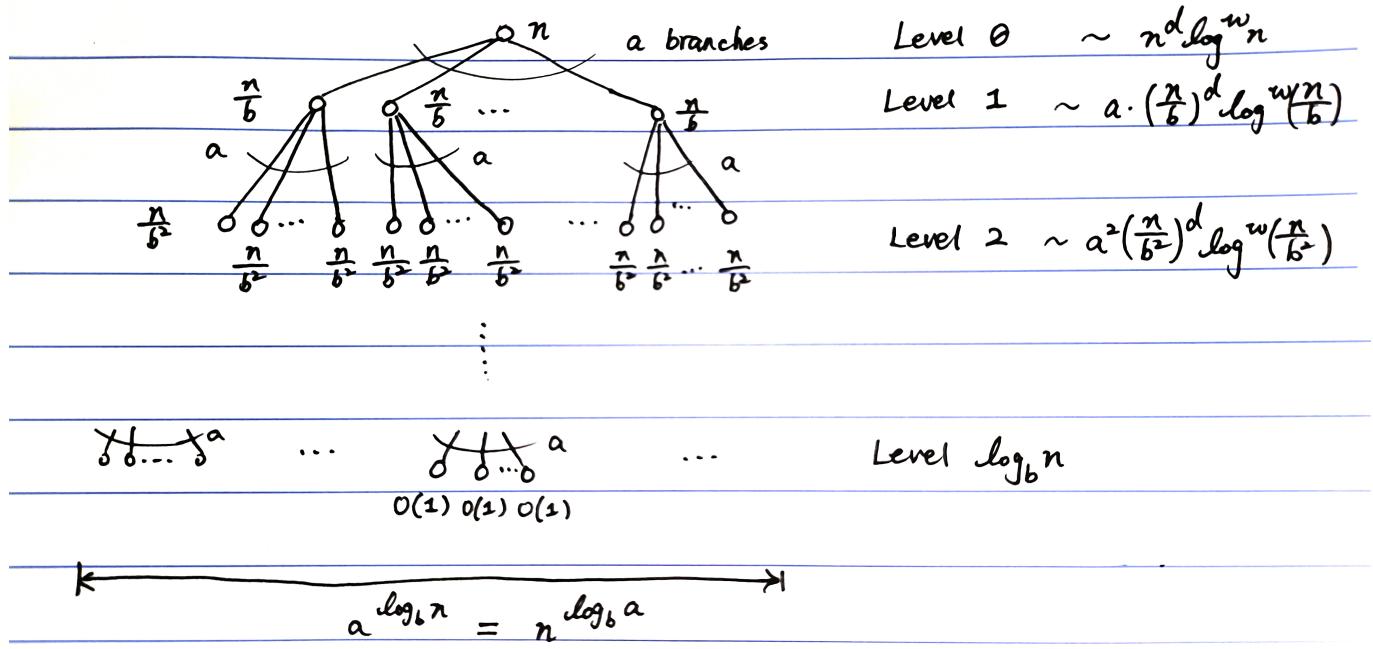
# Algorithm Homework 01

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## 1 Problem 01

*Proof.* We can plot the process of dividing and conquering as follows.



From the recursion tree above, we know for all  $a^k$  subproblems in level  $k$ , the time consumption is at most  $a^k \left(\frac{n}{b^k}\right)^d \log^w \left(\frac{n}{b^k}\right) = a^k \left(\frac{n}{b^k}\right)^d (\log^w n - k \log^w b)$ .

Let  $q = \frac{a}{b^d}$ . Thus, the total time consumption is

$$\begin{aligned}
 T(n) &= \sum_{k=0}^{\log_b n} a^k \left(\frac{n}{b^k}\right)^d (\log^w n - k \log^w b) \\
 &= n^d \log^w n \sum_{k=0}^{\log_b n} \left(\frac{a}{b^d}\right)^k - n^d \log^w b \sum_{k=0}^{\log_b n} k \left(\frac{a}{b^d}\right)^k \\
 &= n^d \left( \log^w n \sum_{k=0}^{\log_b n} q^k - \log^w b \sum_{k=0}^{\log_b n} k q^k \right)
 \end{aligned}$$

**CASE 1.** When  $q = \frac{a}{b^d} < 1$ , i.e.  $a < b^d$ . Let  $K = \log_b n$ .

$$\begin{aligned} T(n) &= n^d \left( \log^w n \frac{1 - q^{K+1}}{1 - q} - \log^w b \frac{Kq^{K+2} - (K+1)q^{K+1} + q}{(q-1)^2} \right) \\ &= n^d (O(\log^w n) - O(1)) \\ &= O(n^d \log^w n) \end{aligned}$$

**CASE 2.** When  $q = \frac{a}{b^d} = 1$ , i.e.  $a = b^d$ ,

$$\begin{aligned} T(n) &= n^d \left( \log^w n \log_b n - O((\log_b n)^2) \right) = n^d \left( O(\log^{w+1} n) - O(\log_b n)^2 \right) \\ &= n^d (O(\log^{w+1} n)) \\ &= O(n^d \log^{w+1} n) \end{aligned}$$

**CASE 3.** When  $q = \frac{a}{b^d} > 1$ , i.e.  $a > b^d$ . Let  $K = \log_b n$ .

$$\begin{aligned} T(n) &= n^d \left( \log^w n \frac{1 - q^{K+1}}{1 - q} - \log^w b \frac{Kq^{K+2} - (K+1)q^{K+1} + q}{(q-1)^2} \right) \\ &= n^d (O(q^K \log^w n) - O(Kq^K)) \\ &= O\left(n^d \left(\frac{a}{b^d}\right)^{\log_b n} (\log^w n - \log_b n)\right) \\ &= O\left(n^d \frac{a^{\log_b n}}{n^d}\right) \\ &= O(n^{\log_b a}) \end{aligned}$$

$$\text{Thus, } T(n) = \begin{cases} O(n^d \log^w n) & \text{if } a < b^d. \\ O(n^d \log^{w+1} n) & \text{if } a = b^d. \\ O(n^{\log_b a}) & \text{if } a > b^d. \end{cases}$$

■

## 2 Problem 02

*Solution.* Similar to the process of **Merge Sort**, the process of **One Third Merge Sort Algorithm** is as follows.

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**Algorithm 1:** One-Third Merge Sort Algorithm

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```

Function One-Third Merge Sort ( $a[1 : n]$ )
     $mid \leftarrow \lceil n/3 \rceil;$ 
     $left \leftarrow \text{OneThirdMergeSort}(a[1 : mid]);$ 
     $right \leftarrow \text{OneThirdMergeSort}(a[mid + 1 : n]);$ 
    Return: Merge( $left, right$ )
end

Function Merge( $left, right$ )
     $i \leftarrow 1, j \leftarrow 1, k \leftarrow 1, n \leftarrow \text{Length}(left), m \leftarrow \text{Length}(right);$ 
    while ( $i \leq n$  and  $j \leq m$ ) do
        if  $left[i] \leq right[j]$  then
             $a[k] \leftarrow left[i];$ 
             $i \leftarrow i + 1;$ 
        else
             $a[k] \leftarrow right[j];$ 
             $j \leftarrow j + 1;$ 
        end
         $k \leftarrow k + 1;$ 
    end
    while  $i \leq n$  do
         $a[k] \leftarrow left[i];$ 
         $i \leftarrow i + 1, k \leftarrow k + 1;$ 
    end
    while  $j \leq m$  do
         $a[k] \leftarrow right[j];$ 
         $j \leftarrow j + 1, k \leftarrow k + 1;$ 
    end
    Return:  $a[1 : n + m]$ 
end

```

---

Now we analyze the time complexity of **One-Third Merge Sort Algorithm**. Let the time consumption be  $T(n)$ .

Obvious the time of **Merge** is  $O(n)$ . Then we have

$$\begin{aligned}
T(n) &= T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + O(n) \\
&\leq T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + cn \\
&= T\left(\frac{4n}{3^2}\right) + 2T\left(\frac{2n}{3^2}\right) + T\left(\frac{n}{3^2}\right) + c\frac{n}{3} + c\frac{2n}{3} + cn \\
&= T\left(\frac{4n}{3^2}\right) + 2T\left(\frac{2n}{3^2}\right) + T\left(\frac{n}{3^2}\right) + 2cn
\end{aligned}$$

$$\begin{aligned}
&= T\left(\frac{8n}{3^3}\right) + 3T\left(\frac{4n}{3^3}\right) + 3T\left(\frac{2n}{3^3}\right) + T\left(\frac{n}{3^3}\right) + 3cn \\
&= \dots \\
&= \sum_{i=0}^k \binom{i}{N} T\left(\frac{2^i n}{3^k}\right) + kcn.
\end{aligned}$$

Suppose  $N = \lceil \log_{2/3} n \rceil$ . Then we have  $k \leq N$ ,  $\frac{2^k n}{3^N} \sim O(1)$ . Obvious  $T(1) = O(1)$ .

$$\begin{aligned}
T(n) &\leq \sum_{i=0}^N \binom{i}{N} T(1) + Ncn = O(2^N) + O(Nn) \\
&= O(n) + O(n \log_{2/3} n) = O(n) + O(n \log n) \\
&= O(n \log n).
\end{aligned}$$

Thus, the time complexity of **One-Third Merge Sort Algorithm** is  $O(n \log n)$ . ■

### 3 Problem 03

#### 3.1 $d = 1$

*Solution.* Based on sort algorithms, we design a  $O(n \log n)$  algorithm as follows.

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##### Algorithm 2: Unidimensional Pair Counting Algorithm

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```

Function Uni-Dim Pair Count ( $A, B$ )
   $S \leftarrow sort(A \cup B);$ 
    //  $sort(\cdot)$  returns the array sorted from the largest to the smallest.

   $count \leftarrow 0, b \leftarrow |B|;$ 
  for  $i = 1 \rightarrow n$  do
    | if  $S[i] \in A$  then  $count \leftarrow count + b$  else  $b \leftarrow b - 1$ ;
    end
  Return:  $count$ 
end

```

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The time complexity of  $sort$  is  $O(n \log n)$  while the remaining part of the algorithm takes  $O(n)$  time. Thus, the time complexity of the algorithm above is  $O(n \log n)$ . ■

### 3.2 $d = 2$

*Solution.* Inspired by the algorithm counting inverted-ordered pairs we discussed in class, a feasible algorithm to count two-dimensional  $(\mathbf{a}, \mathbf{b}) \in (A, B)$  s.t.  $a$  is greater than  $b$  is as follows.

Let  $\mathbf{a} = (a_1, a_2)$  for  $\mathbf{a} \in A$ ,  $\mathbf{b} = (b_1, b_2)$  for  $\mathbf{b} \in B$ . Use  $\mathbf{a} > \mathbf{b}$  to denote  $\mathbf{a}$  is greater than  $\mathbf{b}$ .

Each time, we can find a suitable  $mid$  to divide arrays  $A, B$  into four parts,  $A_{left}, A_{right}, B_{left}, B_{right}$ , where  $A_{left} = \{\mathbf{a} \in A \mid a_1 < mid\}$ ,  $A_{right} = \{\mathbf{a} \in A \mid a_1 \geq mid\}$ ,  $B_{left} = \{\mathbf{b} \in B \mid b_1 < mid\}$ ,  $B_{right} = \{\mathbf{b} \in B \mid b_1 \geq mid\}$ .

Let  $N_{l,l}, N_{r,r}, N_{r,l}, N_{l,r}$  be the number of pairs between  $A_{left}$  and  $B_{left}$ , between  $A_{right}$  and  $B_{right}$ , between  $A_{right}$  and  $B_{left}$ , and between  $A_{left}$  and  $B_{right}$  respectively. Obvious we have  $N_{r,l} = 0$  since  $\forall a \in A_{right}, \forall b \in B_{left}, a_1 < mid \leq b_1$ . Meanwhile, since  $\forall a \in A_{left}, \forall b \in B_{right}, a_1 \geq mid > b_1$ , we know  $N_{l,r}$  is also the number of pairs between  $A' = \{a_2 \mid \mathbf{a} = (a_1, a_2), \mathbf{a} \in A_{left}\}$  and  $B' = \{b_2 \mid \mathbf{b} = (b_1, b_2), \mathbf{b} \in B_{right}\}$ , i.e.  $N_{l,r} = Uni\ Dim\ Pair\ Count(A', B')$ .

Thus, we can divide the problem into three subproblems since the total number of pairs between  $A$  and  $B$  is  $N = N_{l,l} + N_{l,r} + N_{r,l} + N_{r,r} = N_{l,l} + N_{r,r} + Uni\ Dim\ Pair\ Count(A', B')$ .

The algorithm is as follows.

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**Algorithm 3:** Two-Dimensional Pair Counting Algorithm

---

```

Function Two-Dim Pair Count ( $A, B$ )
  if  $A$  is empty or  $B$  is empty then   Return: 0;
   $S \leftarrow sort(A \cup B);$ 
  // sort(.) returns the array sorted by the first element from the largest to
  // the smallest.
   $mid \leftarrow$  the first element of  $S[Length(S)/2];$ 
   $A_{left} \leftarrow \{\mathbf{a} = (a_1, a_2) \in A \mid a_1 < mid\}, A_{right} \leftarrow \{\mathbf{a} = (a_1, a_2) \in A \mid a_1 \geq mid\};$ 
   $B_{left} \leftarrow \{\mathbf{b} = (b_1, b_2) \in B \mid b_1 < mid\}, B_{right} \leftarrow \{\mathbf{b} = (b_1, b_2) \in B \mid b_1 \geq mid\};$ 

   $count \leftarrow Two\text{-Dim Pair Count}(A_{left}, B_{left}) + Two\text{-Dim Pair Count}(A_{right}, B_{right});$ 
   $A' \leftarrow \{a_2 \mid \mathbf{a} = (a_1, a_2), \mathbf{a} \in A_{left}\};$ 
   $B' \leftarrow \{b_2 \mid \mathbf{b} = (b_1, b_2), \mathbf{b} \in B_{right}\};$ 
   $count \leftarrow count + Uni\text{-Dim Pair Count}(A', B');$ 
  Return:  $count$ 
end

```

---

Now we analyze the time complexity of the algorithm above. Let the time complexity be  $T(n)$ , where  $n = |A| + |B|$ . Obvious  $sort(A \cup B)$  takes  $O(n \log n)$  time. From 3.1 we know  $Uni\text{-Dim Pair Count}(A', B')$  takes  $O(n \log n)$  time.

We have

$$T(n) = 2T\left(\frac{n}{2}\right) + O(n \log n).$$

According to **Problem 01**, set  $a = b = 2, d = w = 1$  and we know  $T(n) = O(n \log^2(n))$ .

(Since  $a = 2 = b^d$ .)

We know  $O(n \log^2 n) = o(n^{1.1})$ . Therefore, the algorithm satisfies all requirements given by the problem.  $\blacksquare$

### 3.3 $d \in \mathbb{N}^+$

*Solution.* From **3.2** we know we can convert the  $d$ -dimensional problem into a  $(d - 1)$ -dimensional one. It is quite natural to design the following algorithm. The proof of the correctness of the algorithm is similar to the one in **3.2**.

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#### Algorithm 4: Multi-Dimensional Pair Counting Algorithm

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```

Function Multi-Dim Pair Count ( $A, B, d$ )
  //  $A, B$ : two arrays.  $d$ : the number of dimensions of elements in  $A$  and  $B$ .

  if  $A$  is empty or  $B$  is empty then Return: 0;
  if  $d=1$  then Return: Uni-Dim Pair Count( $A, B$ );

   $S \leftarrow sort(A \cup B)$ ;
  //  $sort(\cdot)$  returns the array sorted by the first element from the largest to
  // the smallest.
   $mid \leftarrow$  the first element of  $S[Length(S)/2]$ ;
   $A_{left} \leftarrow \{\mathbf{a} = (a_1, \dots, a_d) \in A \mid a_1 < mid\}, A_{right} \leftarrow \{\mathbf{a} = (a_1, \dots, a_d) \in A \mid a_1 \geq mid\}$ ;
   $B_{left} \leftarrow \{\mathbf{b} = (b_1, \dots, b_d) \in B \mid b_1 < mid\}, B_{right} \leftarrow \{\mathbf{b} = (b_1, \dots, b_d) \in B \mid b_1 \geq mid\}$ ;

   $count \leftarrow Multi-Dim\ Pair\ Count(A_{left}, B_{left}, d) + Multi-Dim\ Pair\ Count(A_{right}, B_{right}, d)$ ;
   $A' \leftarrow \{(a_2, \dots, a_d) \mid \mathbf{a} = (a_1, a_2, \dots, a_d), \mathbf{a} \in A_{left}\}$ ;
   $B' \leftarrow \{(b_2, \dots, b_d) \mid \mathbf{b} = (b_1, b_2, \dots, b_d), \mathbf{b} \in B_{right}\}$ ;
   $count \leftarrow count + Multi-Dim\ Pair\ Count(A', B', d - 1)$ ;
  Return:  $count$ 
end

```

---

Now we analyze the time complexity of the algorithm above. We prove that  $T(n, d) = O(n \log^d n)$  by induction.

**BASE STEP.** When  $d = 1$ , from **3.1** we know  $T(n, d) = n \log n$ .

**INDUCTIVE HYPOTHESIS.** When  $d = D$ ,  $T(n, d) = n \log^d n$ .

**INDUCTIVE STEP.** When  $d = D + 1$ , we have

$$T(n, D + 1) = 2T\left(\frac{n}{2}, D + 1\right) + T(n, D) = 2T\left(\frac{n}{2}, D + 1\right) + O(n \log^D n)$$

From **Problem 01**, set  $a = b = 2, w = D, d = 1$  and we know  $T(n, D + 1) = O(n \log^{D+1} n)$  (since  $a = 2 = b^d$ ), i.e.  $T(n, d) = O(n \log^d n)$  still holds for  $d = D + 1$ .

Thus,  $\forall n, d \in \mathbb{N}^+, T(n, d) = O(n \log^d n)$ .

Therefore, the time complexity of the algorithm is  $O(n \log^d n)$ .  $\blacksquare$

## 4 Problem 04

### 4.1 Median of Medians Is Close to the Actual Median

*Proof.* Since  $x$  is the median of  $\frac{n}{3}$  medians, we know at least exist  $\frac{n}{6}$  numbers smaller than  $x$ .

Meanwhile, by the definition of medians, we know each median is larger than a number in the triple group. Therefore, there exist at least  $\frac{n}{6}$  numbers which are not medians but are smaller than  $x$ .

In conclusion, there are  $\frac{n}{6} + \frac{n}{6} = \frac{n}{3}$  numbers smaller than  $x$ .

Similarly, we know exist  $\frac{n}{3}$  numbers larger than  $x$ . ■

### 4.2 Algorithm Design

*Solution.* The following algorithm requires  $n$  to be the power of 3. This is easy to reach since we can add several numbers smaller than the minimal element in the original  $a$  array into the array and extend the length to a power of 3, which takes only  $O(n)$  time.

After extend the array  $a$ , the algorithm is as follows.

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#### Algorithm 5: Median of Medians Algorithm

---

```

Function Find k-th Largest( $a[1 : n]$ ,  $k$ )
    //  $a$ : the array containing distinct numbers.
    // We require  $n = 3^m$ ,  $m \in \mathbb{N}^+$ .
    //  $k$ : the expected output should be the  $k$ -largest number in  $a$ .

    // Generate medians.
    Medians  $\leftarrow \emptyset$ ;
    for  $i = 0 \rightarrow \frac{n}{3} - 1$  do
         $b \leftarrow \text{sort}(a[3i + 1 : 3i + 3])$ ;
        Add element  $b[2]$  into Medians;
    end
    mid  $\leftarrow \text{Find } k\text{-th Largest}(\text{Medians}, \lceil \frac{k}{6} \rceil)$ ;

    Left  $\leftarrow \emptyset$ , Right  $\leftarrow \emptyset$ ;
    for  $i = 0 \rightarrow n$  do
        if  $a[i] \geq \text{mid}$  then Add element  $a[i]$  into Right;
        else Add element  $a[i]$  into Left;
    end
    if  $|\text{Right}| = k$  then Return: mid;
    if  $|\text{Right}| > k$  then Return: Find k-th Largest(Right, k);
    Return: Find k-th Largest(Left, k - |Right|)
end

```

---

Now we analyze the time complexity of *Median of Medians* Algorithm.

Let the time complexity be  $T(n)$ .

Obviously median generation takes  $O\left(\frac{n}{3} \log 3\right) = O(n)$  time and dividing  $a$  takes  $O(n)$  time. From **4.1** we know  $|Left| \geq \frac{n}{3}, |Right| \geq \frac{n}{3}$ , i.e.  $|Left| \leq \frac{2n}{3}, |Right| \leq \frac{2n}{3}$ . Then we have

$$\begin{aligned} T(n) &\leq T\left(\frac{n}{3}\right) + T(\max|Left|, |Right|) + O(n) \\ &\leq T\left(\frac{n}{3}\right) + T\left(\frac{2n}{3}\right) + O(n) \end{aligned}$$

We find the equation above is exactly the form of time complexity in **Problem 02**.

Thus, by **Problem 02**, we know  $T(n) = O(n \log n)$ . ■

### 4.3 Alter the Size of Group (The Problem In the Margin)

*Solution.* It's easy to improve the algorithm in **4.2** to a version with alterable group size.

For even size, we select the  $(size/2+1)$ -largest number as the median; For odd size, we select the  $\lceil size/2 \rceil$ -largest number as the median. Obviously there exist at least  $\frac{n}{size} (\lceil size/2 \rceil - 1)$  numbers smaller than the median of medians, which is similar to **4.1**.

The altered algorithm is as follows.

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#### Algorithm 6: Median of Medians Algorithm

---

```

Function Find k-th Largest( $a[1 : n]$ ,  $k$ ,  $size$ )
  //  $a$ : the array containing distinct numbers.
  // We require  $n = size^m$ ,  $m \in \mathbb{N}^+$ .
  //  $k$ : the expected output should be the  $k$ -largest number in  $a$ .
  //  $size$ : the size of median group. We require  $size \geq 3$ .

  // Generate medians.
  Medians  $\leftarrow \emptyset$ ;
  for  $i = 0 \rightarrow \frac{n}{size} - 1$  do
     $b \leftarrow sort(a[size \cdot i + 1 : size \cdot (i + 1) - 1])$ ;
    Add element  $b[\lceil \frac{size}{2} \rceil]$  into Medians;
  end
  mid  $\leftarrow$  Find k-th Largest(Medians,  $\lceil \frac{k}{2 \cdot size} \rceil$ , size);

  Left  $\leftarrow \emptyset$ , Right  $\leftarrow \emptyset$ ;
  for  $i = 0 \rightarrow n$  do
    if  $a[i] \geq mid$  then Add element  $a[i]$  into Right;
    else Add element  $a[i]$  into Left;
  end
  if  $|Right| = k$  then Return: mid;
  if  $|Right| > k$  then Return: Find k-th Largest(Right,  $k$ , size);
  Return: Find k-th Largest(Left,  $k - |Right|$ , size)
end

```

---

Obvious median generation takes  $(\frac{n}{size} \cdot size \log size)$  operations and dividing  $a$  takes  $n$  operations. Also we know  $|Left| \geq \frac{n}{size} (\lceil \frac{size}{2} \rceil - 1)$ ,  $|Right| \geq \frac{n}{size} (\lceil \frac{size}{2} \rceil - 1)$ ,  $|Left| \leq n - \frac{n}{size} (\lceil \frac{size}{2} \rceil - 1)$ ,  $|Right| \leq n - \frac{n}{size} (\lceil \frac{size}{2} \rceil - 1)$ . Thus, we have

$$T(n) \leq T\left(\frac{n}{size}\right) + T\left(n - \frac{n}{size} \left(\lceil \frac{size}{2} \rceil - 1\right)\right) + O(n)$$

Considering

$$\begin{aligned} \frac{n}{size} + n - \frac{n}{size} \left(\lceil \frac{size}{2} \rceil - 1\right) + \frac{n}{size} &\leq n \iff \frac{2n}{size} \leq \frac{n}{size} \left(\lceil \frac{size}{2} \rceil - 1\right) \\ &\iff \left\lceil \frac{size}{2} \right\rceil \geq 3 \\ &\iff size \geq 5, \end{aligned}$$

we analyze time complexity by different  $size$ .

**CASE 01.** When  $size \geq 5$ . we prove that  $T(n) = O(n)$  by induction.

**BASE STEP.** When  $n = 1$ , Obvious  $T(n) = O(n)$  (the time median generation need.)

**INDUCTIVE HYPOTHESIS.** When  $n \leq N, T(n) = O(n)$ .

**INDUCTIVE STEP.** When  $n = N + 1$ ,

$$\begin{aligned} T(N+1) &\leq T\left(\frac{N+1}{size}\right) + T\left(N+1 - \frac{N+1}{size} \left(\lceil \frac{size}{2} \rceil - 1\right)\right) \\ &\quad + N+1 + \frac{N+1}{size} \cdot size \log size \\ &\leq c \cdot \frac{N+1}{size} + c \cdot \left(N+1 - \frac{N+1}{size} \left(\lceil \frac{size}{2} \rceil - 1\right)\right) \\ &\quad + N+1 + \frac{N+1}{size} \cdot size \log size \\ &\leq c \cdot (N+1) \end{aligned}$$

when  $c$  is large enough.

Thus, when  $size \geq 5$ ,  $T(n) = O(n)$ .

**CASE 02.** When  $size = 4$ ,

$$T(n) = T\left(\frac{n}{4}\right) + T\left(\frac{3n}{4}\right) + O(n)$$

By analyses similar to **Problem 02**, we get  $T(n) = O(n \log n)$ .

**CASE 03.** When  $size = 3$ , from **4.2**, we know  $T(n) = O(n \log n)$ .

$$\text{Thus, } T(n) = \begin{cases} O(n), & size \geq 5 \\ O(n \log n), & size = 3, 4 \end{cases}$$

Obvious the best choice of  $size$  is 5. ■

## 5 Rating and Feedback

The completion of this homework takes me seven days, about 40 hours in total. Writing formal proof and solution by *latex* is the most time-consuming part.

The ratings of each problem is as follows.

Problem	Rating
1	3
2	2
3.1	2
3.2	3
3.3	3
4.1	1
4.2	3
4 problem in margin	4

Table 1: Ratings.

Most problems are completed on my own, except ***the one in the margin of the Problem 04.*** The solution of this additional problem is completed with discussions with **Yilin Sun.**