

# Analysis and Design of Algorithms

## Divide and Conquer

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### 1 Maximum Subarray Problem

- Input: Array  $A[1..n]$  of integers (not necessarily positive).
- Output: Indices  $i, j$  such that the sum of the elements in  $A[i..j]$  is maximum possible.

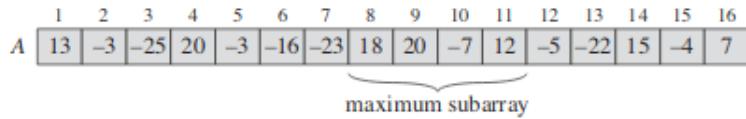


Figure 1: Taken from the book Cormen, Introduction to Algorithms

Brute force solution: Try all possible subsequences. As there are  $\binom{n}{2}$  possibilities, the algorithm is  $\Theta(n^2)$ .

Can we do better?

Note that, given an array  $A[low..high]$ , a maximum subarray has three possibilities. It is

- Entirely in  $A[low..mid]$
- Entirely in  $A[mid + 1..high]$
- With a part in  $A[low..mid]$  and the other part in  $A[mid + 1..high]$

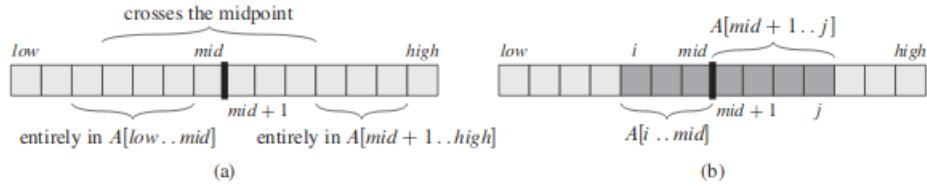


Figure 2: Taken from the book Cormen, Introduction to Algorithms

This observation allows us to design a divide and conquer algorithm for the problem.

- **Divide** We split  $A$  into two subarrays of size  $(high - low + 1)/2$
- **Conquer:** In each subarray, we find the corresponding maximum subarray.
- **Combine** We find which of the two subarrays from the recursive calls has the maximum sum.

Then we need to compare this sum with a third candidate, which is the maximum subarray with a part in  $A[low..mid]$  and the other part in  $A[mid + 1..high]$

Next, we will see how to find this mentioned candidate in the "Combine" operation.

*Input:* An array  $A$  of integers, and three indices  $low \leq mid < high$ .

*Output:* Three integers  $i, j, \sum_{k=i}^j A[k]$  such that  $\sum_{k=i}^j A[k]$  is maximum for all  $i, j$  satisfying  $i \leq mid < j$

```
FIND-MAX-CROSSING-SUBARRAY( $A, low, mid, high$ )
1  left-sum =  $-\infty$ 
2  sum = 0
3  for  $i = mid$  downto  $low$ 
4      sum = sum +  $A[i]$ 
5      if sum > left-sum
6          left-sum = sum
7          max-left =  $i$ 
8  right-sum =  $-\infty$ 
9  sum = 0
10 for  $j = mid + 1$  to  $high$ 
11     sum = sum +  $A[j]$ 
12     if sum > right-sum
13         right-sum = sum
14         max-right =  $j$ 
15 return (max-left, max-right, left-sum + right-sum)
```

Figure 3: Taken from the book Cormen, Introduction to Algorithms

Execution time:

$$T(n) = c_1 + (low - mid + 1)c_2 + (high - mid)c_3 = c_1 + c_4(low - mid + 1) = c_1 + 4n = \Theta(n)$$

Now we will analyze the main algorithm.

*Input:* An array of integers  $A[low..high]$ .

*Output:* Three integers  $i, j, \sum_{k=i}^j A[k]$  such that  $\sum_{k=i}^j A[k]$  is maximum for all  $i, j$

```

FIND-MAXIMUM-SUBARRAY( $A, low, high$ )
1  if  $high == low$ 
2    return ( $low, high, A[low]$ )                                // base case: only one element
3  else  $mid = \lfloor (low + high)/2 \rfloor$ 
4    ( $left-low, left-high, left-sum$ ) =
      FIND-MAXIMUM-SUBARRAY( $A, low, mid$ )
5    ( $right-low, right-high, right-sum$ ) =
      FIND-MAXIMUM-SUBARRAY( $A, mid + 1, high$ )
6    ( $cross-low, cross-high, cross-sum$ ) =
      FIND-MAX-CROSSING-SUBARRAY( $A, low, mid, high$ )
7    if  $left-sum \geq right-sum$  and  $left-sum \geq cross-sum$ 
8      return ( $left-low, left-high, left-sum$ )
9    elseif  $right-sum \geq left-sum$  and  $right-sum \geq cross-sum$ 
10   return ( $right-low, right-high, right-sum$ )
11   else return ( $cross-low, cross-high, cross-sum$ )

```

Figure 4: Taken from the book Cormen, Introduction to Algorithms

Execution time of FIND-MAXIMUM-SUBARRAY. When  $n = 1$ , lines 1 and 2 are executed: time  $c_1 + c_2$ . When  $n > 1$ , we have  $T(n) = T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + c_3 + c_7 + c_8 + c_9 + c_{10} + c_{11} + k_1 n$

$$T(n) = \begin{cases} c & n = 1 \\ T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + k_1 n + k_2 & \text{otherwise} \end{cases}$$

So  $T(n) = \Theta(n \log n)$ . (To see this, apply the master theorem, since when  $n$  is a power of 2, we have  $T(n) = 2T(n/2) + kn$ , we have  $a = b = 2, k = 1$ , second case of the master theorem)

## 2 Multiplication of Natural Numbers

- Input: Two natural numbers  $a$  and  $b$  with  $n$  digits each.
- Output: The product  $a \cdot b$

The usual algorithm:

$$\begin{array}{r}
 \begin{array}{r} 9999 \\ 7777 \\ \hline 69993 \\ 69993 \\ 69993 \\ \hline 69993 \\ \hline 77762223 \end{array} \quad \begin{array}{l} A \\ B \\ C \\ D \\ E \\ F \\ G \end{array} \end{array}$$

In the previous figure,  $A \cdot B = C + D + E + F = G$ . The execution time is proportional to  $4 + 4 + 4 + 4 = 16 = 4^2$ .

This is a simple algorithm that can be described as follows:

**Require:** Two integers represented by  $a[1..n], b[1..n]$

**Ensure:** The product  $a \cdot b$

BASIC-MULTIPLICATION( $a, b, n$ )	<i>cost</i>	<i>times</i>
1: total = 0	$c_1$	1
2: <b>for</b> $j = 1$ <b>to</b> $n$	$c_2$	$n + 1$
3:   sum = 0	$c_3$	$n$
4: <b>for</b> $i = 1$ <b>to</b> $n$	$c_4$	$(n + 1) \cdot n$
5:     sum = sum $\cdot 10 + b[j] \cdot a[i]$	$c_5$	$n \cdot n$
6:   total = total $\cdot 10 +$ sum	$c_6$	$n$
7: <b>return</b> total	$c_7$	1

It's clear that the execution time is  $\Theta(n^2)$ . Now we'll see how to improve this algorithm.

## 2.1 A Divide and Conquer Algorithm

Note that, given two natural numbers  $a$  and  $b$  with  $n$  digits, we can express them as

$$\begin{aligned}
 a &= a_1 \cdot 10^m + a_2, \\
 b &= b_1 \cdot 10^m + b_2.
 \end{aligned}$$

Where  $m = \lceil n/2 \rceil$ .

For example,  $3213209842 = 32132 \cdot 10^5 + 09842$  or  $953421412 = 9534 \cdot 10^5 + 21412$ .

Therefore

$$\begin{aligned}
 a \cdot b &= (a_1 \cdot 10^m + a_2) \cdot (b_1 \cdot 10^m + b_2) \\
 &= (a_1 b_1) \cdot 10^{2m} + (a_1 b_2 + a_2 b_1) \cdot 10^m + (a_2 b_2)
 \end{aligned}$$

This way, we can divide our original problem into four subproblems: multiplying  $a_1$  by  $b_1$ , multiplying  $a_1$  by  $b_2$ , multiplying  $a_2$  by  $b_1$ , and multiplying  $a_2$  by  $b_2$ .

We obtain the following divide and conquer algorithm:

**Require:** Two integers  $a$  and  $b$  with  $n$  digits, where  $n$  is a power of two, and both  $a$  and  $b$  do not contain zeros.

**Ensure:** The product  $a \cdot b$

MULTIPLICATION-DC ( $a, b, n$ )	cost	times
1: <b>if</b> $n = 1$	$\Theta(1)$	1
2: <b>return</b> $a \cdot b$	$\Theta(n)$	1
3: $a_1 = \lfloor a/10^{n/2} \rfloor$	$\Theta(n)$	1
4: $a_2 = a \bmod 10^{n/2}$	$\Theta(n)$	1
5: $b_1 = \lfloor b/10^{n/2} \rfloor$	$\Theta(n)$	1
6: $b_2 = b \bmod 10^{n/2}$	$\Theta(n)$	1
7: $p = \text{MULTIPLICATION-DC}(a_1, b_1, n/2)$	$T(n/2)$	1
8: $q = \text{MULTIPLICATION-DC}(a_1, b_2, n/2)$	$T(n/2)$	1
9: $r = \text{MULTIPLICATION-DC}(a_2, b_1, n/2)$	$T(n/2)$	1
10: $s = \text{MULTIPLICATION-DC}(a_2, b_2, n/2)$	$T(n/2)$	1
11: <b>return</b> $p \cdot 10^n + (q + r) \cdot 10^{n/2} + s$	$\Theta(n)$	1

Execution time:

$$T(n) = \begin{cases} \Theta(n) & n = 1 \\ 4T(n/2) + \Theta(n) & \text{otherwise} \end{cases}$$

By the Master Theorem,  $\lg 4/\lg 2 = 2 > 1$ . Thus  $T(n) = \Theta(n^2)$ .

We see that MULTIPLICATION-DC does not improve BASIC-MULTIPLICATION.

## 2.2 Karatsuba's Algorithm

This algorithm will improve the previous recurrence by making only 3 recursive calls. Therefore, we'll have an execution time of  $\Theta(n^{\lg 3/\lg 2}) = \Theta(n^{1.59})$ .

The main observation is as follows.

Given two natural numbers  $a$  and  $b$  with  $n$  digits, we express them as in the previous subsection:

$$a = a_1 \cdot 10^m + a_2,$$

$$b = b_1 \cdot 10^m + b_2.$$

Where  $m = \lceil n/2 \rceil$ .

We have

$$\begin{aligned} a \cdot b &= (a_1 \cdot 10^m + a_2) \cdot (b_1 \cdot 10^m + b_2) \\ &= (a_1 b_1) \cdot 10^{2m} + (a_1 b_2 + a_2 b_1) \cdot 10^m + (a_2 b_2) \\ &= (a_1 b_1) \cdot 10^{2m} + ((a_1 + a_2)(b_1 + b_2) - a_1 b_1 - a_2 b_2) \cdot 10^m + (a_2 b_2) \end{aligned}$$

This way, we only need to calculate three products:  $a_1 b_1$ ,  $a_2 b_2$ , and  $(a_1 + a_2)(b_1 + b_2)$ . The product  $a_1 b_2 + a_2 b_1$  can be calculated as  $(a_1 + a_2)(b_1 + b_2) - a_1 b_1 - a_2 b_2$ .

**Require:** Two integers  $a$  and  $b$  with  $n$  digits, where  $n$  is a power of 2 and both  $a$  and  $b$  do not contain zeros

**Ensure:** The product  $a \cdot b$

KARATSUBA ( $a, b$ )	<i>cost</i>	<i>times</i>
1: <b>if</b> $n \leq 1$	$\Theta(1)$	1
2: <b>return</b> $a \cdot b$	$\Theta(n)$	1
3: $a_1 = \lfloor a/10^{n/2} \rfloor$	$\Theta(n)$	1
4: $a_2 = a \bmod 10^{n/2}$	$\Theta(n)$	1
5: $b_1 = \lfloor b/10^{n/2} \rfloor$	$\Theta(n)$	1
6: $b_2 = b \bmod 10^{n/2}$	$\Theta(n)$	1
7: $p = \text{KARATSUBA}(a_1, b_1)$	$T(n/2)$	1
8: $q = \text{KARATSUBA}(a_1 + a_2, b_1 + b_2)$	$T(n/2)$	1
9: $s = \text{KARATSUBA}(a_2, b_2)$	$T(n/2)$	1
10: <b>return</b> $p \cdot 10^n + (q - p - s) \cdot 10^{n/2} + s$	$\Theta(n)$	1

Execution time:

$$T(n) = \begin{cases} \Theta(1) & n = 1 \\ 3T(n/2) + \Theta(n) & \text{otherwise} \end{cases}$$

By the Master Theorem,  $\lg 3 / \lg 2 > 1$ . Thus  $T(n) = \Theta(n^{\lg 3 / \lg 2}) = \Theta(n^{1.59\dots})$ .

Observation: In the above pseudocode, we have assumed for simplicity that both  $a$  and  $b$  have  $n$  digits and that  $n$  is a power of 2. Otherwise, a sufficient trick for a good implementation is to pad with leading zeros if necessary.

### 3 Matrix Multiplication

Problem: Given two matrices  $A = (a_{ij})$  and  $B = (b_{ij})$  of dimensions  $n \times n$ , calculate  $C = A \cdot B$ . Recall that  $C = (c_{ij})$  is defined as

$$c_{ij} = \sum_{k=1}^n a_{ik}b_{kj}.$$

We have the following algorithm to compute  $C$ .

Input: Two matrices  $A$  and  $B$  of dimensions  $n \times n$  Output:  $A \cdot B$

MULTIPLY ( $A, B$ )

- 1: **for**  $i = 1$  to  $n$
- 2:   **for**  $j = 1$  to  $n$
- 3:      $c_{ij} = 0$
- 4:     **for**  $k = 1$  to  $n$
- 5:        $c_{ij} = c_{ij} + a_{ik}b_{kj}$
- 6:   **return**  $C$

A simple analysis of the algorithm tells us that its execution time is  $\Theta(n^3)$ .

#### 3.1 Divide and Conquer Algorithm

For simplicity, let's assume that  $n$  is a power of 2. We can partition each of the matrices into four parts. That is,

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix},$$

$$B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix},$$

$$C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

Therefore,

$$C_{11} = A_{11} \cdot B_{11} + A_{12} \cdot B_{21} \quad (1)$$

$$C_{12} = A_{11} \cdot B_{12} + A_{12} \cdot B_{22} \quad (2)$$

$$C_{21} = A_{21} \cdot B_{11} + A_{22} \cdot B_{21} \quad (3)$$

$$C_{22} = A_{21} \cdot B_{12} + A_{22} \cdot B_{22} \quad (4)$$

From these equations, we can formulate the following divide and conquer algorithm.

Input: Two matrices  $A$  and  $B$  of dimensions  $n \times n$  Output:  $A \cdot B$

MULTIPLY-DC ( $A, B$ )

- 1: **if**  $n = 1$
- 2:    $c_{11} = a_{11} \cdot b_{11}$
- 3: **else**
- 4:   Create auxiliary matrices
- 5:    $C_{11} = \text{MULTIPLY-DC}(A_{11}, B_{11}) + \text{MULTIPLY-DC}(A_{12}, B_{21})$
- 6:    $C_{12} = \text{MULTIPLY-DC}(A_{11}, B_{12}) + \text{MULTIPLY-DC}(A_{12}, B_{22})$
- 7:    $C_{21} = \text{MULTIPLY-DC}(A_{21}, B_{11}) + \text{MULTIPLY-DC}(A_{22}, B_{21})$
- 8:    $C_{22} = \text{MULTIPLY-DC}(A_{21}, B_{12}) + \text{MULTIPLY-DC}(A_{22}, B_{22})$
- 9:   Fill  $C$  from its parts
- 10: **return**  $C$

We have the following recurrence.

$$T(n) = \begin{cases} \Theta(1) & n = 1 \\ 8T(n/2) + \Theta(n^2) & \text{otherwise} \end{cases}$$

The Master Theorem tells us that  $T(n) = \Theta(n^3)$ . It has not improved the basic algorithm.

### 3.2 Strassen's Algorithm

Analogously to the Karatsuba algorithm, Strassen's algorithm attempts to perform only 7 recursive operations. Before the calls, we will create the following matrices:

$$S_1 = B_{12} - B_{22}$$

$$\begin{aligned}
S_2 &= A_{11} + A_{12} \\
S_3 &= A_{21} + A_{22} \\
S_4 &= B_{21} - B_{11} \\
S_5 &= A_{11} + A_{22} \\
S_6 &= B_{11} + B_{22} \\
S_7 &= A_{12} - A_{22} \\
S_8 &= B_{21} + B_{22} \\
S_9 &= A_{11} - A_{21} \\
S_{10} &= B_{11} + B_{12}
\end{aligned}$$

Input: Two matrices  $A$  and  $B$  of dimensions  $n \times n$  Output:  $A \cdot B$

STRASSEN ( $A, B$ )

```

1: if  $n = 1$ 
2:    $c_{11} = a_{11} \cdot b_{11}$ 
3: else
4:   Create auxiliary matrices
5:    $P_1 = \text{MULTIPLY-DC}(A_{11}, S_1)$ 
6:    $P_2 = \text{MULTIPLY-DC}(S_2, B_{22})$ 
7:    $P_3 = \text{MULTIPLY-DC}(S_3, B_{11})$ 
8:    $P_4 = \text{MULTIPLY-DC}(A_{22}, S_4)$ 
9:    $P_5 = \text{MULTIPLY-DC}(S_5, S_6)$ 
10:   $P_6 = \text{MULTIPLY-DC}(S_7, S_8)$ 
11:   $P_7 = \text{MULTIPLY-DC}(S_9, S_{10})$ 
12:   $C_{11} = P_5 + P_4 - P_2 + P_6$ 
13:   $C_{12} = P_1 + P_2$ 
14:   $C_{21} = P_3 + P_4$ 
15:   $C_{22} = P_5 + P_1 - P_3 - P_7$ 
16:  Fill  $C$  from its parts
17: return  $C$ 

```

We have the following recurrence

$$T(n) = \begin{cases} \Theta(1) & n = 1 \\ 7T(n/2) + \Theta(n^2) & \text{otherwise} \end{cases}$$

The Master Theorem tells us that  $T(n) = \Theta(n^{\lg 7})$ .

## 4 Counting Inversions

- Input: Array  $A[1..n]$  of distinct integers
- Output: Number of inversions. Where an *inversion* is a pair  $(i, j)$  such that  $i < j$  and  $A[i] > A[j]$

For example, let  $A = [2, 4, 1, 3, 5]$ . The inversions are  $(1, 3), (2, 3), (2, 4)$

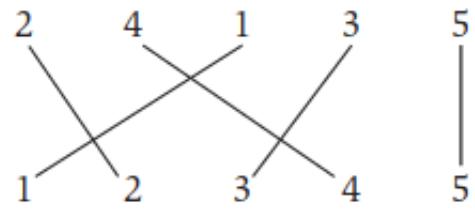


Figure 5: Taken from the book Kleinberg-Tardos, Algorithm Design

Naive Algorithm: Evaluate all possible ordered pairs and check if they are inversions.

Input: An array of distinct integers  $A[1..n]$

Output: The number of inversions in  $A$ .

NAIVE-INVERSIONS( $A, n$ )	$cost$	$times$
1: total = 0	$c_1$	1
2: <b>for</b> $i = 1$ to $n - 1$	$c_2$	$n$
3: <b>for</b> $j = i + 1$ to $n$	$c_3$	$\sum_{i=1}^{n-1} n - i + 1$
4: <b>if</b> $A[i] > A[j]$	$c_4$	$\sum_{i=1}^{n-1} n - i$
5:       total = total + 1	$c_5$	$\sum_{i=1}^{n-1} t_i$
6: <b>return</b> total	$c_6$	1

Where  $t_i$  is the number of times we execute line 5 during the  $i$ -th iteration.

Clearly we have  $0 \leq t_i \leq n - i$ .

This algorithm, both in the worst and best case, consumes  $\Theta(n^2)$  time. (An example of the worst case occurs if the array is sorted in decreasing order). Next, we will see a divide and conquer algorithm for this problem.

## 4.1 Divide and Conquer

We observe that, given the array  $A[1..n]$ , an inversion  $(i, j)$  can be in exactly one of these possibilities:

- $i, j \in \{1, \dots, \lfloor n/2 \rfloor\}$
- $i, j \in \{\lfloor n/2 \rfloor + 1, \dots, n\}$
- $i \in \{1, \dots, \lfloor n/2 \rfloor\}, j \in \{\lfloor n/2 \rfloor + 1, \dots, n\}$

That gives us the idea of a divide and conquer algorithm.

- **Divide** We divide  $A$  into two subarrays of sizes  $\lfloor n/2 \rfloor, \lceil n/2 \rceil$
- **Conquer:** In each subarray, we find the number of inversions
- **Combine** We add this number with the number of inversions that have one element before or equal to  $\lfloor n/2 \rfloor$  and the other after  $\lfloor n/2 \rfloor$ . This would give us the total number of inversions.

So, informally, we see that  $T(n) = 2T(n/2) + C(n)$ , where  $C(n)$  is the time to combine. If we want the recurrence to be  $\Theta(n \lg n)$  then  $C(n)$  must be  $\Theta(n)$ .

But if we try to count the  $(i, j)$  such that  $1 \leq i \leq \lfloor n/2 \rfloor$  and  $\lfloor n/2 \rfloor + 1 \leq j \leq n$  with brute force, we can spend time proportional to  $n/2 \cdot n/2 = \Theta(n^2)$ . We need a more efficient algorithm.

Note that if the subarrays  $A[1..\lfloor n/2 \rfloor]$  and  $A[\lfloor n/2 \rfloor + 1 \dots n]$  were already sorted, then we would only spend time  $\Theta(n)$  due to the following observation:

If  $(i, j)$  is an inversion with  $i \in \{1, \dots, \lfloor n/2 \rfloor\}$  and  $j \in \{\lfloor n/2 \rfloor + 1, \dots, n\}$   
then  $(i', j)$  is also an inversion for all  $i' > i$  (5)

To prove (5), it suffices to observe that  $A[i] > A[j]$  because  $(i, j)$  is an inversion and that  $A[i'] > A[i]$  because  $A$  is sorted. The following subroutine takes care of this counting. It reminds us of the Merge function of Mergesort. Then, the following subroutine would do what is asked.

Input: An array of distinct integers  $A[1..n]$  and three indices  $p, q, r$  such that  $A[p..q]$  and  $A[q+1..r]$  are sorted.

Output: The number of inversions  $(i, j)$  in  $A$  such that  $i \in \{p, \dots, q\}$ ,  $j \in \{q+1, \dots, r\}$ . Also, it sorts the array  $A[p..r]$ .

CENTRAL-INVERSIONS( $A, p, q, r$ )

```

1:  $n_1 = q - p + 1$ 
2:  $n_2 = r - q$ 
3: Let  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$  be new arrays
4: for  $i = 1$  to  $n_1$ 
5:    $L[i] = A[p + i - 1]$ 
6: for  $j = 1$  to  $n_2$ 
7:    $R[j] = A[q + j]$ 
8:  $L[n_1 + 1] = \infty$ 
9:  $L[n_2 + 1] = \infty$ 
10:  $i = 1$ 
11:  $j = 1$ 
12: total = 0
13: for  $k = p$  to  $r$ 
14:   if  $L[i] > R[j]$ 
15:      $A[k] = R[j]$ 
16:     total = total + ( $n_1 + 1 - i$ )
17:      $j = j + 1$ 
18:   else
19:      $A[k] = L[i]$ 
20:      $i = i + 1$ 
21: return total
```

Note that CENTRAL-INVERSIONS is essentially the Merge subroutine of Mergesort, which consumes time  $\Theta(n)$ .

Finally, with the help of this subroutine, we design our main algorithm.

Input: An array of distinct integers  $A[p..r]$

Output: The number of inversions in  $A$ .

INVERSIONS-DC( $A, p, r$ )	<i>cost</i>	<i>times</i>
1: <b>if</b> ( $p == r$ )	$c_1$	1
2: <b>return</b> 0	$c_2$	0
3: $q = \lfloor \frac{r-p+1}{2} \rfloor$	$c_3$	1
4: $total_1 = \text{INVERSIONS-DC}(A, p, q)$	$T(\lfloor n/2 \rfloor)$	1
5: $total_2 = \text{INVERSIONS-DC}(A, q+1, r)$	$T(\lceil n/2 \rceil)$	1
6: $total_3 = \text{CENTRAL-INVERSIONS}(A, p, q, r)$	$kn$	1
7: <b>return</b> $total_1 + total_2 + total_3$	$c_5$	1

Note that  $T(n) = T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + kn$ . Then, by the Master Theorem,  $T(n) = \Theta(n \lg n)$ .