

# Chapter 1

1. (a)

1	<u>7</u>	<u>2</u>	6	3	5	4
1	2	<u>7</u>	<u>6</u>	3	5	4
1	2	6	<u>7</u>	<u>3</u>	5	4
1	2	<u>6</u>	<u>3</u>	7	5	4
1	2	3	6	<u>7</u>	<u>5</u>	4
1	2	3	<u>6</u>	<u>5</u>	7	4
1	2	3	5	6	<u>7</u>	4
1	2	3	5	<u>6</u>	<u>4</u>	7
1	2	3	<u>5</u>	<u>4</u>	6	7
1	2	3	4	5	6	7

(b)

1	2	<u>8</u>	<u>7</u>	3	4	6	5
1	2	7	<u>8</u>	<u>3</u>	4	6	5
1	2	<u>7</u>	<u>3</u>	8	4	6	5
1	2	3	7	<u>8</u>	<u>4</u>	6	5
1	2	3	<u>7</u>	<u>4</u>	8	6	5
1	2	3	4	7	<u>8</u>	<u>6</u>	5
1	2	3	4	<u>7</u>	<u>6</u>	8	5
1	2	3	4	6	7	<u>8</u>	<u>5</u>
1	2	3	4	6	<u>7</u>	<u>5</u>	8
1	2	3	4	<u>6</u>	<u>5</u>	7	8
1	2	3	4	5	6	7	8

3. (a)

1	<u>7</u>	<u>2</u>	6	3	5	4
1	2	<u>7</u>	6	<u>3</u>	5	4
1	2	3	<u>6</u>	7	5	<u>4</u>
1	2	3	4	<u>7</u>	<u>5</u>	6
1	2	3	4	5	<u>7</u>	<u>6</u>
1	2	3	4	5	6	7

(b)

1	2	<u>8</u>	7	<u>3</u>	4	6	5
1	2	3	<u>7</u>	8	<u>4</u>	6	5
1	2	3	4	<u>8</u>	7	6	<u>5</u>
1	2	3	4	5	<u>7</u>	<u>6</u>	8
1	2	3	4	5	6	7	8

5. Take  $N = 1$  and  $c = 13$ . Then

$$f(n) = 4n^2 + 9n^3 \leq 4n^3 + 9n^3 = 13n^3 = cn^3$$

for all  $n \geq N$ . Therefore, by the definition of  $O$ , we have  $f(n) = O(n^3)$ .

Now take  $N = 1$  and  $c = 1$ . Then

$$f(n) = 4n^2 + 9n^3 \geq n^3 = cn^3$$

for all  $n \geq N$ . Thus, by the definition of  $\Omega$ , we have  $f(n) = \Omega(n^3)$ .

By the definition of  $\Theta$ , since we have  $f(n) = O(n^3)$  and  $f(n) = \Omega(n^3)$ , then  $f(n) = \Theta(n^3)$ .

7. We have

$$\lim_{n \rightarrow \infty} \frac{2017n^4 + 4n^{2017}}{n^{2017}} = \lim_{n \rightarrow \infty} \left( \frac{2017n^4}{n^{2017}} + \frac{4n^{2017}}{n^{2017}} \right) = \lim_{n \rightarrow \infty} \left( \frac{2017}{n^{2013}} + 4 \right) = 4.$$

Therefore, by the limit criterion, we have  $2017n^4 + 4n^{2017} = \Theta(n^{2017})$ .

9. The following algorithm finds the largest number in a list of  $n$  numbers by scanning the list and keeping track of the largest number found so far.

**Input:** a non-empty list  $a_0, a_1, \dots, a_{n-1}$  of numbers

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 $x \leftarrow a_0$ 
for  $i = 1$  to  $n - 1$  do
  if  $a_i > x$  then
     $x \leftarrow a_i$ 
  end if
end for
return  $x$ 

```

Independently of the input, the for-loop runs  $n - 1$  times, each time performing exactly one comparison. Therefore this algorithm takes  $\Theta(n)$  time in the worst case.

11. Here is the trick. To simplify the discussion, suppose that  $n$  is even. Let  $A[1..n]$  be the array. Split the array into  $\frac{n}{2}$  pairs

$$A[1], A[2] \quad | \quad A[3], A[4] \quad | \quad \dots \quad | \quad A[n-1], A[n].$$

find the minimum and the maximum of each pair by doing one comparison within each pair. Therefore, we have made  $\frac{n}{2}$  comparisons so far.

You end up with  $\frac{n}{2}$  *small numbers* (one for each pair). The minimum of these  $\frac{n}{2}$  small numbers is the minimum you are looking for. Since you have  $\frac{n}{2}$  numbers, you can find this minimum with  $\frac{n}{2}$  comparisons. Hence, in total, we have made  $\frac{n}{2} + \frac{n}{2} = n$  comparisons so far.

You do the same thing for the maximum. That is, you find the maximum of the  $\frac{n}{2}$  *large numbers* and you are done. This adds an extra  $\frac{n}{2}$  comparisons for a total of  $\frac{3}{2}n$  comparisons.

**Input:** An array  $A[1..n]$  of  $n$  numbers.

Initialize an array  $A_{small}[1..n/2]$

Initialize an array  $A_{big}[1..n/2]$

**for**  $i = 1$  **to**  $n/2$  **do**

**if**  $A[2i - 1] < A[2i]$  **then**

$A_{small}[i] \leftarrow A[2i - 1]$

$A_{big}[i] \leftarrow A[2i]$

**else**

$A_{small}[i] \leftarrow A[2i]$

$A_{big}[i] \leftarrow A[2i - 1]$

**end if**

**end for**

$x_{small} \leftarrow A_{small}[1]$

**for**  $i = 2$  **to**  $n/2$  **do**

**if**  $A_{small}[i] < x_{small}$  **then**

$x_{small} \leftarrow A_{small}[i]$

**end if**

**end for**

$x_{big} \leftarrow A_{big}[1]$

**for**  $i = 2$  **to**  $n/2$  **do**

**if**  $A_{big}[i] > x_{big}$  **then**

$x_{big} \leftarrow A_{big}[i]$

**end if**

**end for**

**return**  $x_{small}, x_{big}$

13. (a) **Input:** An array  $A[1..n]$  of  $n$  distinct integers in the interval  $[1, 2019n]$ .

    Initialize an array  $B[1..(2019n)]$  with 0's everywhere.

**for**  $i = 1$  **to**  $n$  **do**

$B[A[i]] \leftarrow 1$

**end for**

$i \leftarrow 1$

**for**  $j = 1$  **to**  $2019n$  **do**

**if**  $B[j] = 1$  **then**

$A[i] \leftarrow j$

$i \leftarrow i + 1$

**end if**

**end for**

This is called *bucket sort*.

It takes  $\Theta(2019n) = \Theta(n)$  time to initialize  $B$ . Then we scan  $A$  once (and modify  $B$ ). This takes  $\Theta(n)$  time. Then we scan  $B$  once (and fill  $A$ ) this takes  $\Theta(2019n) = \Theta(n)$  time. In total, it takes  $\Theta(n)$  time.

- (b) **Input:** An array  $A[1..n]$  of  $n$  distinct integers in the interval  $[1, kn]$ .

Initialize an array  $B[1..(kn)]$  with 0's everywhere.

**for**  $i = 1$  **to**  $n$  **do**

$B[A[i]] \leftarrow 1$

**end for**

$i \leftarrow 1$

**for**  $j = 1$  **to**  $kn$  **do**

**if**  $B[j] = 1$  **then**

$A[i] \leftarrow j$

$i \leftarrow i + 1$

**end if**

**end for**

The number  $k$  is a **fixed** integer. Whatever is the input,  $k$  stays the same. Therefore, it takes  $\Theta(kn) = \Theta(n)$  time to initialize  $B$ . Then we scan  $A$  once (and modify  $B$ ). This takes  $\Theta(n)$  time. Then we scan  $B$  once (and fill  $A$ ) this takes  $\Theta(kn) = \Theta(n)$  time. In total, it takes  $\Theta(n)$  time.

- (c) The same technique works with a tiny modification. Do you see how?
- (d) If we do not have any upper on the numbers to be sorted, we do not know what should be the size of  $B$ . So this technique does not work. Is there another technique we could use to sort  $A$  in  $O(n)$  time? By the end of the semester, we will see that the answer is no.

15.

$$\begin{aligned}
 T(n) &= \sum_{j=1}^{n/2} \sum_{i=1}^j 1 \\
 &= \sum_{j=1}^{n/2} j \\
 &= \frac{n/2(n/2 + 1)}{2} \\
 &= \frac{n^2}{8} + \frac{n}{4} \\
 &= \Theta(n^2)
 \end{aligned}$$

17. How many times do we execute the inner loop? How many times can we do " $\lfloor j/2 \rfloor$ " starting from  $n$ ?

$$1 + \log_2(n)$$

So we get

$$\begin{aligned}
T(n) &= \sum_{i=1}^n (1 + \log_2(n)) \\
&= n(1 + \log_2(n)) \\
&= n + n \log_2(n) \\
&= \Theta(n \log_2(n))
\end{aligned}$$

19. We proceed by induction.

The definition of  $F_n$  is recursive and there are *two* base cases in that definition. Therefore, there are *two* base cases to consider in our proof by induction: we have  $F_6 = 8 \geq 8 = 2^{6/2}$  and  $F_7 = 13 \geq 8\sqrt{2} = 2^{7/2}$ .

For the inductive step, let any natural number  $n \geq 8$  be given and assume that  $F_{n-2} \geq 2^{(n-2)/2}$  and  $F_{n-1} \geq 2^{(n-1)/2}$ . We then have

$$F_n = F_{n-2} + F_{n-1} \geq 2^{(n-2)/2} + 2^{(n-1)/2} \geq 2 \times 2^{(n-2)/2} = 2^{(n-2)/2+1} = 2^{n/2},$$

which completes the proof.

The inequality is false for  $n < 6$  since

$$\begin{aligned}
F_0 &= 0 < 1 = 2^{0/2}, \\
F_1 &= 1 < \sqrt{2} = 2^{1/2}, \\
F_2 &= 1 < 2 = 2^{2/2}, \\
F_3 &= 2 < 2\sqrt{2} = 2^{3/2}, \\
F_4 &= 3 < 4 = 2^{4/2}, \\
F_5 &= 5 < 4\sqrt{2} = 2^{5/2}.
\end{aligned}$$

21. We proceed by induction. For the base cases, we have  $0 \leq 0 = 2F_0$  and  $1 \leq 2 = 2F_1$ .

On top of that, we prove the inequality for  $n = 2$ :

$$2 \leq 2 = 2F_2.$$

For the inductive step, let any natural number  $n \geq 3$  be given and assume that  $n - 2 \leq 2F_{n-2}$  and  $n - 1 \leq 2F_{n-1}$ . We then have

$$2F_n = 2(F_{n-2} + F_{n-1}) = 2F_{n-2} + 2F_{n-1} \geq (n - 2) + (n - 1) = 2n - 3 \geq n.$$

This last inequality is true for all  $n \geq 3$ . This is why we had to prove the case  $n = 2$  separately.

23. Let  $T(n)$  be the number of bit-operations performed by  $fib(n)$ . We will show that there is a constant  $c \geq 1$  such that  $T(n) \leq 2cnF_n$  by induction.

For the base cases, since  $fib(0)$  and  $fib(1)$  only return a number, they do not execute any bit-operation. So  $T(0) = 0 \leq 0 = 2c \cdot 0 \cdot F_0$  and  $T(1) = 0 \leq 2c = 2c \cdot 1 \cdot F_1$  for any constant  $c \geq 1$ .

For the inductive step, let any natural number  $n \geq 2$  be given and assume that  $T(n-2) \leq 2c(n-2)F_{n-2}$  and  $T(n-1) \leq 2c(n-1)F_{n-1}$ . Note that  $F_n < 2^n$  (refer to Exercise #22) implies that  $F_n$  has at most  $O(\log_2(F_n)) = O(\log_2(2^n)) = O(n)$  bits. Hence the sum in the  $fib$  algorithm makes  $O(n)$  bit-operations.

Therefore, from the algorithm, there is a constant  $c \geq 1$  such that

$$\begin{aligned}
 T(n) &\leq T(n-2) + T(n-1) + cn \\
 &\leq 2c(n-2)F_{n-2} + 2c(n-1)F_{n-1} + cn \\
 &= 2cn(F_{n-2} + F_{n-1}) - 4cF_{n-2} - 2cF_{n-1} + cn \\
 &\leq 2cnF_n - 2cF_{n-2} - 2cF_{n-1} + cn \\
 &= 2cnF_n - 2c(F_{n-2} + F_{n-1}) + cn \\
 &= 2cnF_n - 2cF_n + cn \\
 &= 2cnF_n - c(2F_n - n), \\
 &\leq 2cnF_n
 \end{aligned}$$

by Exercise #21.

25. We prove that  $2^n \neq \Omega(3^n)$ , which proves that  $2^n \neq \Theta(3^n)$ . The proof is by contradiction using the definition of  $\Omega$ . Suppose there are positive constants  $N$  and  $c$  such that

$$2^n \geq c3^n$$

for all  $n \geq N$ . Then we get

$$(3/2)^n \leq 1/c \tag{1}$$

for all  $n \geq N$ . As  $n$  increases,  $(3/2)^n$  gets arbitrarily large. Therefore, since  $c$  is a constant, (1) cannot be true for all  $n \geq N$ . This is a contradiction.