

## 1

### 1.1

The idea is to scan the array from left to right and update the maximum revenue meanwhile. See algorithm below for details.

---

**Algorithm 1** Max Revenue-1,1( $\mathbf{a}$ )

---

```
1: procedure MAX REVENUE-1,1( $\mathbf{a}$ ):  $\triangleright \mathbf{a} = \{a_1, a_2, a_3, \dots, a_n\}$ 
2:   result  $\leftarrow$  0, sum  $\leftarrow$  0,  $i \leftarrow$  1  $\triangleright$  sum is our current optimal to maintain
3:   while  $i < n + 1$  do
4:     sum  $\leftarrow$  sum +  $a_i$ 
5:     if sum > result then
6:       result  $\leftarrow$  sum  $\triangleright$  Update the optimal subsequence
7:     if sum < 0 then
8:       sum  $\leftarrow$  0  $\triangleright$  Discard the bad subsequence we do not want
9:      $i \leftarrow i + 1$ 
10:  return result
```

---

Since our algorithm only requires us to scan the array once, clearly the time complexity is  $O(n)$ .

## 1.2

The idea is still scanning the array from left to right once, but we will use an array  $\mathbf{s}[n]$  to store the maximum revenue of a subsequence ending at position  $i, i \in [n]$ .

First we initialize  $s_1, s_2, s_3, \dots, s_L$ . Then for each element  $a_i, i \in [n]$ , we need to update  $R - L + 1$  elements in  $\mathbf{s}[n]$ , which are  $s_{i+L}, s_{i+L+1}, \dots, s_{i+R}$ . The initialize rule and update rule are given in the algorithm below.

---

**Algorithm 2** Max Revenue( $L, R, \mathbf{a}$ )

---

```
1: procedure MAX REVENUE( $L, R, \mathbf{a}$ ):▷  $\mathbf{a} = \{a_1, a_2, a_3, \dots, a_n\}$ 
2:   result  $\leftarrow 0, i \leftarrow 1, s_i \leftarrow 0, \forall i \in [n]$ ▷  $\mathbf{s}$  is described above
3:   while  $i < L + 1$  do
4:     if  $a_i > 0$  then
5:        $s_i \leftarrow a_i$ ▷ Initialization for the first  $L$  elements
6:     if  $a_i > \text{result}$  then
7:       result  $\leftarrow a_i$ ▷ Update the result
8:      $i \leftarrow i + 1$ 
9:    $i \leftarrow 1$ 
10:  while  $i + L < n + 1$  do
11:    step  $\leftarrow L$ 
12:    while  $i + \text{step} < n + 1$  do
13:      if  $s_i + a[i + \text{step}] > s[i + \text{step}]$  then
14:         $s[i + \text{step}] = s_i + a[i + \text{step}]$ ▷ update the optimal result at  $i$ 
15:      if  $s[i + \text{step}] > \text{result}$  then
16:        result  $\leftarrow s[i + \text{step}]$ 
17:      step  $\leftarrow \text{step} + 1$ 
18:     $i \leftarrow i + 1$ 
19:  return result
```

---

We need to do  $R - L + 1$  updates at each round and there are  $n$  rounds in total, so the time complexity is  $O((R - L + 1)n)$ . So with the difference between  $L$  and  $R$  approaching  $n$ , our algorithm actually becomes  $O(n^2)$ .

## 1.3

We still need to scan the array from left to right once, but instead of updating  $R - L + 1$  elements at each iteration, we use a better strategy.

For each  $a_i$ , we look for the largest  $s_j, i - R \leq j \leq i - L$  and use this largest result to update  $s_i$ . Then by our algorithm in class, we can find all the largest  $s_j$  for our current  $a_i$  with only  $O(n)$  time. A detailed algorithm is given below.

---

### Algorithm 3 Max Revenue( $L, R, \mathbf{a}$ )

---

```

1: procedure MAX REVENUE( $L, R, \mathbf{a}$ ): ▷  $\mathbf{a} = \{a_1, a_2, a_3, \dots, a_n\}$ 
2:   result ← 0,  $i \leftarrow 1$   $s_i \leftarrow 0, \forall i \in [n]$ 
3:   while  $i < L + 1$  do
4:     if  $a_i > 0$  then
5:        $s_i \leftarrow a_i$  ▷ Initialization is the same as algorithm 2
6:     if  $a_i > \text{result}$  then
7:       result ←  $a_i$  ▷ Update the result
8:      $i \leftarrow i + 1$ 
9:    $i \leftarrow L + 1$ 
10:  while  $i < n + 1$  do
11:    max ← k-Largest( $\mathbf{s}, i - R, i - L$ ) ▷ Algorithm in class to find max in  $O(1)$  time
12:    if max +  $a_i > s_i$  then
13:       $s_i \leftarrow \text{max} + a_i$ 
14:    if  $s_i > \text{result}$  then
15:      result ←  $s_i$ 
16:     $i \leftarrow i + 1$ 
17:  return result

```

---

The reason why this algorithm is faster is that by using the k-Largest algorithm, we do not have to update  $R - L + 1$  times each round. We only need to look up the largest result before a certain element and the look up process is  $O(1)$  for each element. As a result, the total running time of our algorithm can be reduced to  $O(n)$ .