Lab01-Algorithm Analysis

CS214-Algorithm and Complexity, Xiaofeng Gao, Spring 2019.

- * If there is any problem, please contact TA Mingran Peng. Also please use English in homework.

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- 1. Read Algorithm 1 and Algorithm 2 carefully.

```
Algorithm 1: SelectionSort
  Input: An array A[1, \dots, n]
  Output: A[1, \cdots, n] sorted
               nonincreasingly
i \leftarrow 1;
2 for i \leftarrow 1 to n-1 do
      max \leftarrow A[i]; pos \leftarrow i;
3
       for j \leftarrow i + 1 to n do
4
           if A[j] > max then
5
               max \leftarrow A[j];
6
               pos \leftarrow j;
7
      swap A[pos] and A[i];
```

```
Algorithm 2: CocktailSort
   Input: An array A[1, \dots, n]
   Output: A[1, \dots, n] sorted
               nonincreasingly
i \leftarrow 1; j \leftarrow n; sorted \leftarrow false;
2 while not sorted do
       sorted \leftarrow true;
       for k \leftarrow i to j-1 do
4
           if A[k] < A[k+1] then
5
               swap A[k] and A[k+1];
 6
               sorted \leftarrow false;
 7
       j \leftarrow j - 1;
8
       for k \leftarrow j downto i + 1 do
9
           if A[k-1] < A[k] then
10
               swap A[k-1] and A[k];
11
               sorted \leftarrow false;
12
       i \leftarrow i + 1;
13
```

Fill in the blank and explain your answers. You need to answer when the best case and the worst case happen. (Hint: if it's both O(g) and $\Omega(g)$, just answer $\Theta(g)$)

Algorithm	Time Complexity	Space Complexity
$\overline{InsertionSort}$	$O(n^2), \Omega(n)$	$\Theta(1)$
CocktailSort	$O(n^2), \Omega(n)$	$\Theta(1)$
Selection Sort	$\Theta(n^2)$	$\Theta(1)$

Solution.

- Analysis of Cocktail Sort:
 - (a) An illustration of Cocktail Sort

Before the analysis of time complexity and space complexity, we firstly show how *Cocktail Sort* works. Each iteration of *Cocktail Sort* is broken up into following **two stages***. †

 $^{^{\}dagger}$ Rahul Agrawal (2019) Cocktail Sort. [Online] Available from: https://www.geeksforgeeks.org/cocktail-sort [Accessed 06/03/19].

- The first stage loops through the array from left position i to position j-1. During the inner **for** loop, adjacent items are compared and if value on the left is smaller than the value on the right, values are swapped.

At the end of first inner for loop, the smallest number will reside at the position j.

- The second stage loops through the array in the opposite direction - starting from the position j-1, and moving back to the position i+1. Similarly, adjacent items are compared and are swapped if meeting the requirement.

At the end of second inner for loop, the largest number will reside at the position i.

Here is an illustration of *Cocktail Sort* as depicted in Figure 1.

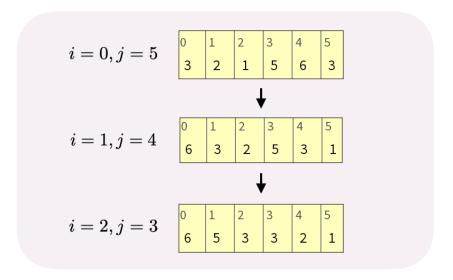


Figure 1: An illustration of Cocktail Sort

(b) Time Complexity

In the analysis of time complexity of *Cocktail Sort*, we use the number of iterations as a good indicative of the running time.

- The best case is that array A is already sorted nonincreasingly.

In the first **while** loop, there is no index k which satisfies A[k] < A[k+1]. Thus sorted is set to be **true** in step 3 and remains it value. There are totally 2n-3 iterations of the inner **for** loop in the first **while** iteration. In the second **while** loop, the sorted value is **true** which contradicts with the loop condition and the loop ends.

We can conclude that the time complexity of *Cocktail Sort* is $\Omega(n)$

The worst case is that array A is in an increasing order.

Firstly, we can easily derive that there are ((j-i)+(j-i-1)) inner iterations in each **while** loop.

So that the total iterations can be calculated as follows:

$$\sum_{i=1,j=n}^{A \text{ is sorted}} (2j-2i-1)$$

After each **while** loop, we decrease the value of j and increase the value of i both by 1. Thus we can get the equation between i and j:

$$j = n + 1 - i$$

For the worst case, the while loop stops until i reaches $\lfloor \frac{n}{2} \rfloor$. So we can derive the total iterations:

$$\sum_{i=1}^{i=\lfloor \frac{n}{2} \rfloor} (2n+1-4i) = \sum_{i=1}^{i=\lfloor \frac{n}{2} \rfloor} (n+1-2i) + \sum_{i=1}^{i=\lfloor \frac{n}{2} \rfloor} (n-2i)$$

$$\cdot = \sum_{i=1}^{n} (n-i)$$

$$= \frac{n(n-1)}{2}$$

So that the time complexity of *Cocktail Sort* is $O(n^2)$.

(c) Space Complexity

In Algorithm $Cocktail\ Sort$, only constant number of auxiliary memory cells are used to hold the value of sorted, i, j, k, sorted.

So we can conclude that the space complexity is $\Theta(1)$ for all cases.

• Analysis of Selection Sort:

(a) Time complexity

In the analysis of time complexity of *selection sort*, we use the number of iterations as a good indicative of the running time.

The inner for loop is executed repeatedly for the following values of n:

$$n-1, n-2, n-3, \ldots, 1.$$

Thus, the total number of Step 6 is executed is

$$\sum_{i=1}^{n-1} (n-i).$$

Since

$$\sum_{i=1}^{n-1} (n-i) = \frac{n(n-1)}{2},$$

we conclude that Step 6 is executed $\Theta(n^2)$ times. Note that the number of iterations is independent on the input array A.

We conclude that *Selection Sort* is $\Theta(n^2)$ for both the best case and the worst case.

(b) Space Complexity

In Algorithm Selection Sort, only constant number of memory cells are used to hold the value of max, pos, i, j.

So we can conclude that the space complexity is $\Theta(1)$ for all cases.

2. Let us assume that you have learned two type of data structures: **Stack** and **Queue**. **Stack** has two operations: *push* and *pop*, while **Queue** also has two operations: *enqueue* and *dequeue*.

Now you have two **Stacks**, how can you use them to simulate a **Queue**?

- (a) Briefly eplain your approach. (Pusudo code is not needed.)
- (b) Give out the time complexity of *enque* and *deque* operations of the simulated **Queue**. Use **potential function** for amortized analysis.

Solution.

(a)

enqueue(x):

(1) push x to **Stack1**.

dequeue():

- (1) If both **Stacks** are empty, then raise an **Queue-empty-exception**.
- (2) If **Stack2** is empty, push everything from **Stack1** to **Stack2** While **Stack1** is not empty.
 - (3) Pop the top of **Stack2** and return it.

Here is an illustration of *enqueue* and *dequeue* operation in Figure 2.

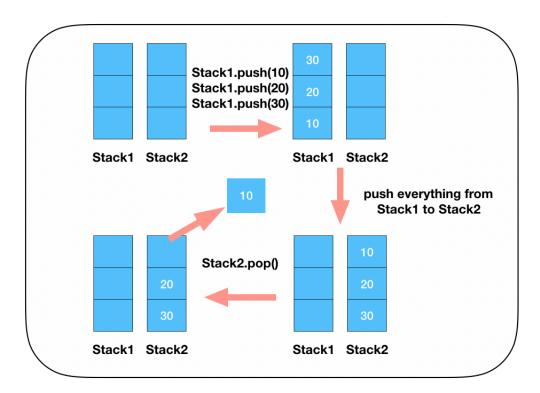


Figure 2: How enqueue and dequeue works on a simple example.

(b)

Considering we have n enqueue and dequeue operations. And we want to use **potential** function method for amortized analysis.

Firstly, we consider the cost C_{enq} for enqueue and C_{deq} for dequeue.

 $C_{enq} = 1$ and C_{deq} is dependent on whether **Stack2** is empty.

$$C_{deq} = \begin{cases} 1 & \textbf{Stack2} \text{ is not empty} \\ 2size(\textbf{Stack1}) + 1 & \textbf{Stack2} \text{ is empty} \end{cases}$$

When we want to do the *dequeue* operation and **Stack2** is empty, we need to first *pop* all elements from **Stack1**, push all elements to **Stack2** and pop the top of **Stack2**. The total cost $C_{deq} = 2size(\mathbf{Stack1}) + 1$. Another two cases are naive and both cost only 1.

We define a potential function $\Phi(S): S \to R$, where S is the state of **Stack1** and **Stack2**.

$$\Phi(S_i) = 2size(\mathbf{Stack1})$$

Correctness:

$$\Phi(S_i) = 2size(\mathbf{Stack1}) \ge 0 = \Phi(S_0)$$

Then amortized cost setting:

$$\hat{C}_i = C_i + \Phi(S_i) - \Phi(S_{i-1})$$

.

The amortized cost of *enqueue*:

$$\hat{C}_{enq} = 1 + \Phi(S_i) - \Phi(S_{i-1})$$

$$= 1 + 2(size(\mathbf{Stack1}) + 1) - 2size(\mathbf{Stack1})$$

$$= 3$$

The amortized cost of dequeue when **Stack2** is not empty:

$$\hat{C}_{enq} = 1 + \Phi(S_i) - \Phi(S_{i-1})$$

$$= 1 + 2size(\mathbf{Stack1}) - 2size(\mathbf{Stack1})$$

$$= 1$$

The amortized cost of dequeue when **Stack2** is empty:

$$\hat{C}_{enq} = 2size(\mathbf{Stack1}) + 1 + \Phi(S_i) - \Phi(S_{i-1})$$

$$= 2size(\mathbf{Stack1}) + 1 + 0 - 2size(\mathbf{Stack1})$$

$$= 1$$

Then the amortized cost of *enqueue* and *dequeue* are 3 and 1 respectively. Finally, we can derive that

$$\sum_{i=1}^{n} C_i \le \sum_{i=1}^{n} \hat{C}_i \le \sum_{i=1}^{n} 3 = 3n.$$

The amortized cost of each operation can be $\frac{T(n)}{n} = O(1)$.

Remark: You need to include your .pdf and .tex files in your uploaded .rar or .zip file.