# CS 218, Spring 2017 Abhishek Kumar SRIVASTAVA

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Solution 1. Given that

$$T(n) = \begin{cases} 2 & n = 2\\ 4 \cdot T(n^{1/2}) + \log^2(n) & n > 2 \end{cases}$$

Consider when n > 2

$$T(n) = 4 \cdot T(n^{1/2}) + \log^2(n)$$

Master theorem cannot be applied on this equation. We need to apply transformation to convert into Master theorem equation.

We can substitute  $n = 2^m$  in the equation. Which will give us

$$T(2^m) = 4 \cdot T(2^{m/2}) + \log^2(2^m)$$
$$= 4 \cdot T(2^{m/2}) + m^2$$

Substituting again in this equation by  $T(2^m) = S(m)$ . We get,

$$S(m) = 4 \cdot S(m/2) + m^2$$

Now we can apply Master theorem on the above equation.

Case 1:  $f(n) \in O(n^{\log_b a - \varepsilon})$  where  $\varepsilon > 0$ .

$$m^2 \epsilon O(m^{\log_2 4 - \varepsilon}) = O(m^{2 - \varepsilon})$$

For any  $\varepsilon > 0$  this is not possible. So case 1 fails.

Case 2:  $f(n) \in \Theta(n^{\log_b a} \cdot \log^k n)$  where  $k \ge 0$ .

$$m^2 \in \Theta(m^{\log_2 4} \cdot \log^k m) = \Theta(m^2 \cdot \log^k m)$$

k=0 satisfy the equation. So case 2 holds. Thats why,

$$S(m) = \Theta(m^{\log_2 4} \cdot \log^{k+1} m)$$
$$= \Theta(m^2 \cdot \log m)$$

By substituting back  $m = \log n$  we get,

$$T(n) = \Theta(\log^2 n \cdot \log \log n)$$

**Solution 2.** Given A[1, ..., n] as a fixed array of distinct integers in which we have to find the position for the elements in array X[1, ..., k].

## **Naive Solution:**

We can search the positions of each element in X[1,...,k] by doing linear search in array A[1,...,n] for that element. Which will the complexity of

$$T(n) = O(n \cdot k)$$

But this is not the lower bound.

# **Binary Decision Tree Solution:**

We can do better than the naive solution to get the lower bound by using binary decision tree model. We first sort the array A[1, ..., n] and then do a binary search for each element in the array X[1, ..., k].

Total cost of doing search this way will be cost of sorting the array and searching each element using binary search. Therefore,

$$T(n) = n \cdot \log n + k \cdot \log n$$

 $(n \cdot \log n)$  since it is the lower bound of sorting  $A[1, \ldots, n]$  using comparison model.  $(k \cdot \log n)$  since  $X[1, \ldots, k]$  contains k elements and doing binary search on the sorted array  $A[1, \ldots, n]$  have the lower bound of  $\log n$ .

$$T(n) = O((n+k)\log n)$$

This gives the lower bound on the time complexity, as a function of n and k, using the binary decision tree model for searching and sorting.

Pseudocode for described algorithm is as following:

Array $A[1,\ldots,n]$	⊳ Fixed Array
Array $X[1,\ldots,k]$	> Array to be searched
Array $I[1,\ldots,k]$	> Array to store position

# function FIND\_POSITION(A, X, I)

SORT(A)  $\triangleright$  Sort the array A in increasing order

for i less than length(X) do  $I[i] = BINARY\_SEARCH(A, X[i])$   $\Rightarrow$  Search X[i] in A and return its position end for

#### end function

**Solution 3.** Implement a queue using two stacks  $S_1$  and  $S_2$ .

#### Part 1:

Queue is data structure which follows **FIFO**(First In First Out) property. On the other hand Stack is data structure which follows **LIFO**(Last In First Out) property. Therefore to implement Queue using Stack we need two of them. We always treat one stack for en-queuing data and other one for de-queuing.

While Enqueue(x) operation we push the element x in one stack for example S1. For Dequeue(x) operation we first check that other stack(S2) is empty or not. If it is not empty we pop the top element from S2 stack. If it is empty we then transfer all the elements from S1 to S2 and then perform the pop operation. By using this method we can achieve **FIFO** property. Psuedocode for the above described method is following:

```
Stack S1
                                                                                    ▶ Input stack
Stack S2
                                                                                  ▶ Output stack
function ENQUEUE(x)
   S1 \cdot push(x)
                                                                      \triangleright Pushing in the stack S1
end function
function DEQUEUE()
                                                                          \triangleright Check if S2 is empty
   if S2.empty() = true then
       TRANSFER_ELEM_S1_TO_S2()
                                                          \triangleright Transfer all elements from S1 to S2
   end if
   return S2.pop()
                                                                     \triangleright Pop the element from S2
end function
function TRANSFER_ELEM_S1_TO_S2()
   while S1.empty() = false do
                                                                        \triangleright While S1 is not empty
       S2.push(S1.pop())
                                                       \triangleright Pop elements from S1 and push in S2
   end while
end function
```

## Part 2: Accounting Method

Since we are using two stack we will have 4 operation to perform. push and pop for both S1 and S2.

## • Proposed charging Scheme:

- For Engueue(x) operation we will charge \$4
  - \* \$1 pays for the pushing the element in S1.
  - \* \$3 is deposited(credit invariant) on the element to pay \$1 when popping the element from S1 later.
  - \* \$1 pays for pushing the element in S2 from remaining \$2.
  - \* \$1 pays for popping the element from S2 from the last remaining \$1.
- For Dequeue() operation we will charge \$0

For any sequence of n For Enqueue(x) and n Dequeue() operations the cost will be

$$T(n) = (4 + \dots n \text{ times } \dots + 4) + (0 + \dots n \text{ times } \dots + 0) = 4n$$

: Each operation costs

$$T(n) = 4n/2n = 2 = O(1)$$