Data Structures and Algorithms (CS210A)

Lecture 29:

- **Building** a Binary heap on *n* elements in O(*n*) time.
- Applications of Binary heap : sorting
- Binary trees: beyond searching and sorting

Recap from the last lecture

A complete binary tree



Building a Binary heap

Problem: Given *n* elements $\{x_0, ..., x_{n-1}\}$, build a binary heap H storing them.

Trivial solution:

(Building the Binary heap incrementally)

CreateHeap(H);

For(i = 0 to n - 1) Insert(x_i ,H);





 \rightarrow Theorem: Time complexity of building a binary heap incrementally is $O(n \log n)$. 5



Building a Binary heap incrementally





heap property:"Every node stores value smaller than its children"We just need to ensure this property at each node.

Think of alternate approach for building a binary heap



heap property:"Every node stores value smaller than its children"We just need to ensure this property at each node.

A new approach to build binary heap

- 1. Just copy the given n elements $\{x_0, ..., x_{n-1}\}$ into an array **H**.
- 2. The **heap property** holds for all the leaf nodes in the corresponding complete binary tree.
- 3. Leaving all the leaf nodes,

process the elements in the decreasing order of their numbering and set the heap property for each of them.

















Heapify(*i*,H)

Heapify(*i*,H) { $n \leftarrow size(H) -1$;

}



Heapify(*i*,H)

```
Heapify(i,H)
  n \leftarrow \text{size}(H) - 1;
{
    Flag \leftarrow true;
    While ( i \leq \lfloor (n-1)/2 \rfloor and Flag = true )
          min \leftarrow i;
    {
          If H[i]>H[2i+1] min \leftarrow 2i+1;
          If 2i + 2 \le n and H[min] > H[2i + 2] ) min \leftarrow 2i + 2;
          If (min \neq i)
             {
                  H(i) \leftrightarrow H(min);
                    i \leftarrow min; \}
          else
                 Flag ← false;
     }
```

}





Each subtree is also a <u>complete</u> binary tree.

→ A subtree of height h has <u>at least 2^h nodes</u>

Moreover, <u>no</u> two subtrees of height h in the given tree have <u>any element in common</u>₂

Building Binary heap in O(n) time

Lemma: the number of nodes of height h is bounded by $\frac{n}{2h}$.

Hence Time complexity to build the heap =
$$\sum_{h=1}^{\log n} \frac{n}{2h} O(h)$$

= $n C \sum_{i=1}^{\log n} \frac{h}{2h}$
= $O(n)$



Sorting using a Binary heap

Sorting using heap

```
Build heap H on the given n elements;
```

```
While (H is not empty)
```

```
{ x ← Extract-min(H);
```

```
print x;
```

```
}
```

This is **HEAP** SORT algorithm

```
Time complexity : O(n log n)
```

Question:

Which is the best sorting algorithm : (Merge sort, Heap sort, Quick sort)? Answer: Practice programming assignment ③

Binary trees: beyond searching and sorting

- Elegant solution for two interesting problem
- An important lesson:

Lack of **proper understanding** of a problem is a big hurdle to solve the problem

Two interesting problems on sequences

What is a sequence ?

A sequence **S** = $\langle x_0, ..., x_{n-1} \rangle$

- Can be viewed as a mapping from [0, *n*].
- Order <u>does</u> matter.

Multi-increment

Given an initial sequence $S = \langle x_0, ..., x_{n-1} \rangle$ of numbers,

maintain a compact data structure to perform the following operations:

• ReportElement(*i*):

Report the current value of x_i .

Multi-Increment(*i*, *j*, Δ):

Add Δ to each x_k for each $i \leq k \leq j$

Example:

Let the initial sequence be $S = \langle 14, 12, 23, 12, 111, 51, 321, -40 \rangle$ After Multi-Increment(2,6,10), S becomes $\langle 14, 12, 33, 22, 121, 61, 331, -40 \rangle$ After Multi-Increment(0,4,25), S becomes $\langle 39, 37, 58, 47, 146, 61, 331, -40 \rangle$ After Multi-Increment(2,5,31), S becomes $\langle 39, 37, 89, 78, 177, 92, 331, -40 \rangle$

Given an initial sequence $S = \langle x_0, ..., x_{n-1} \rangle$ of numbers,

maintain a compact data structure to perform the following operations:

• ReportElement(*i*):

Report the current value of x_i .

Multi-Increment(*i*, *j*, Δ):

```
Add \Delta to each x_k for each i \leq k \leq j
```

Trivial solution :

```
Store S in an array A[0.. n-1] such that A[i] stores the current value of x_i.

Multi-Increment(i, j, \Delta)

{

For (i \le k \le j) A[k] \leftarrow A[k] + \Delta;

}

O(j - i) = O(n)

ReportElement(i){ return A[i] }

O(1)
```

Given an initial sequence **S** = $\prec x_0, ..., x_{n-1} \succ$ of numbers,

maintain a compact data structure to perform the following operations:

• ReportElement(*i*):

Report the current value of x_i .

Multi-Increment(*i*, *j*, Δ):

Add Δ to each x_k for each $i \leq k \leq j$

Trivial solution :

Store **S** in an array **A**[**0**.. *n*-**1**] such that **A**[*i*] stores **the current value** of *x_i*.

Question: the source of difficulty in breaking the **O** (n) barrier for **Multi-Increment() ? Answer:** we need to **explicitly maintain** in **S**.

Question: who asked/inspired us to maintain S explicitly.

Answer: 1. incomplete understanding of the problem

2. conditioning based on incomplete understanding

Towards efficient solution of Problem 1

Assumption: without loss of generality assume *n* is power of 2.

Explore ways to maintain sequence S implicitly such that

- Multi-Increment(*i*, *j*, △) is efficient
- **Report**(*i*) is efficient too.

Main hurdle: To perform Multi-Increment(*i*, *j*, △) efficiently

Dynamic Range-minima

Given an initial sequence $S = \langle x_0, ..., x_{n-1} \rangle$ of numbers, maintain a compact data structure to perform the following operations efficiently for any $0 \leq i < j < n$.

• ReportMin(*i*, *j*):

Report the minimum element from $\{x_k \mid \text{ for each } i \leq k \leq j\}$

• Update(*i*, a):

a becomes the new value of x_i .

AIM:

- **O**(*n*) size data structure.
- **ReportMin**(*i*, *j*) in **O**(log *n*) time.

