CS 521 Lecture II

DREXEL UNIVERSITY DEPT. OF COMPUTER SCIENCE

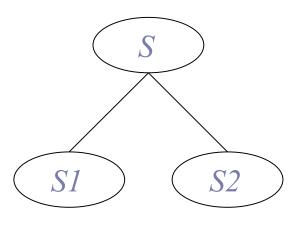
FALL 2011

Today's Lecture



- Continue with Recurrences
 - O Merge Sort
 - Master Theorem
 - Ouicksort
- First Data-Structure:
 - o Heap
 - Its Application in Sorting

Merge-Sort

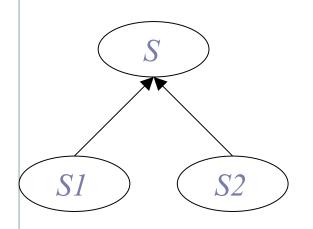


- Problem: Given a list *S* of *n* integers, create a sorted list of elements in *S*.
- Merge-sort Algorithm:

Divide: If *S* has at lest two elements (nothing needs to be done if *S* is empty or has only one element), remove all the elements from *S* and put them into two sequences, *S1* and *S2*, each containing about half of the elements of *S*.

- **Recursion:** Sort sequences S_1 and S_2 .
- **Conquer:** Put back the elements into *S* by merging the sorted sequences *S1* and *S2* into a unique sorted sequence.

Merging Two Sorted Sequences



- Problem: Given two sequences S_1 and S_2 of sizes n_1 and n_2 , create a (union) sorted list S (of size $n=n_1+n_2$).
- Algorithm $Merge(S_1, S_2, S)$:
 - $top(S_i)$ = first element in S_i , for i in $\{1,2\}$.
 - While S_1 is not empty and S_2 is not empty do

```
if top(S_1) < top(S_2) then
move top(S_1) at the end of S
advance top(S_1)
```

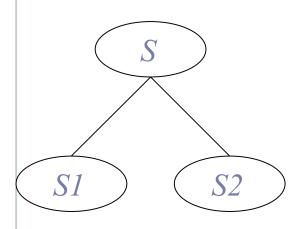
else

move $top(S_2)$ at the end of S advance $top(S_2)$

While S_1 is not empty **do** move the remaining of S_1 to S

While S_2 is not empty do move the remaining of S_2 to S

Recurrence for Merge Sort:



• Recurrence Relation:

$$T(n)=2T(n/2)+n$$

 $T(1)=1$

• Solution by unfolding:

$$T(n) = 2(2T(n/4) + (n/2)) + n$$

$$= 4T(n/2) + 2n$$

$$= 4(2T(n/8) + (n/4)) + 2n$$

$$= 8T(n/8) + 3n =$$
...
$$= 2^{i}T(n/2^{i}) + i.n$$

$$= ...$$

The expansion stops for i=log n $T(n) = 2^{log n} + n log n$ Total Number of moves:

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

Iterative recurrences

$$T(n) = 4T(n/2) + n$$

$$= n + 4(n/2 + 4T(n/4))$$

$$= n + 2n + 16T(n/4)$$

$$= n + 2n + 16[n/4 + 4T(n/8)]$$

$$= n + 2n + 4n + 4T(n/8)$$

$$= n + 2n + 4n + \dots$$

$$= n \sum_{i=0}^{\log n-1} 2^i + 4^{\log n}T(1)$$

$$= \Theta(n^2) + \Theta(n^2)$$

Disadvantage:

- Tedious
- o Error-Prone
- Use to generate initial guess, and then prove by induction.

Master Theorem



• Let a and b be constants, and let f(n) be a nonnegative function defined on integral powers of b. Let T(n) be defined on the integral powers of b as

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ aT\left(\frac{n}{b}\right) + f(n) & \text{if } n = b^k \end{cases}$$

- Then we have:
 •If $f(n) = O(n^{\log_b a \varepsilon})$ for some constant $\varepsilon > 0$, then $T(n) = O(n^{\log_b a})$
- •If $f(n) = \Theta(n^{\log_b a})$ for some constant $\varepsilon > 0$, then $T(n) = O(n^{\log_b a} \log n)$
- •If $f(n) = \Omega(n^{\log_b a + \varepsilon})$ for some constant $\varepsilon > 0$, and if $n \ge b \Rightarrow af(n/b) \le cf(n)$

for some positive constant $c \ge 0$, then $T(n) = \Theta(f(n))$

Example



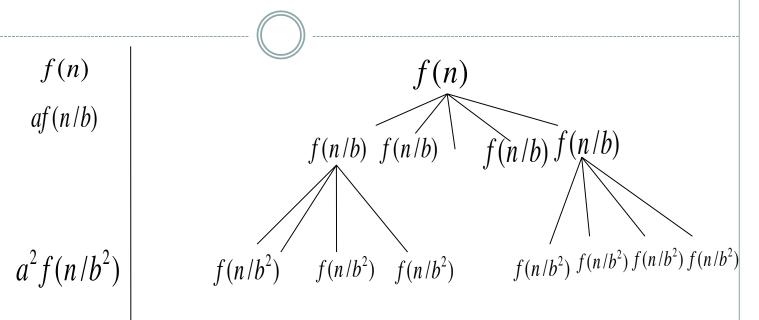
- Consider the recurrence T(n) = T(n/2) + 1 (binary search) Then a = 1, b = 2 and $f(n) = 1 = n^{\log_2 1}$, so by case 2 of Master Theorem $T(n) = \Theta(n^{\log_2 1} \log n) = \Theta(\log n)$.
- Consider the recurrence T(n) = 2T(n/2) + n (merge sort) Then a = 2, b = 2 and $f(n) = n = n^{\log_2 2}$, so by case 2 of Master Theorem $T(n) = \Theta(n \log n)$.
- Consider the recurrence $T(n) = T(n/4) + n^{1/2}$ Then a = 1, b = 4 and $f(n) = n^{1/2} = \Omega(n^{\log_2 1})$, and $af(n/b) = (n/4)^{1/2} = n^{1/2}/2 = 0.5 f(n)$. So by case 3 of Master Theorem $T(n) = \Theta(n^{1/2})$.

Build recursive tree

The tree:

$$f(n)$$
 $af(n/b)$

$$a^2 f(n/b^2)$$



Last row: $\Theta(a^{\log_b n}) = \Theta(n^{\log_b a})$ elemenst, each one $\Theta(1)$.

Total:
$$\Theta(n^{\log_b a}) + \sum_{i=1}^{\log_b n-1} a^i f(n/b^i)$$

Which term dominates?

Back to Algorithms

- Quick Sort
 - Sort in place
 - Very practical
 - Divide-and-conquer
- Algorithm
 - Divide into two arrays around the first element
 - Recursively sort each array
 - Merge/combine-trivial

Partition Routine

```
Partition(A,p,r)

x=A(r)

i=p-1

for j=p to r-1

if A(j) <= x then

i++

exchange(A(i),A(j))

exchange(A(i+1),A(r))

return(i+1)
```

	$\leq = \chi$	$>_{\mathcal{X}}$??		
ß)	i	j	•	r	

Quick Sort

```
Quicksort(A,p,r)
while (p < r)
q = partition(A,p,r)
Quicksort(A,p,q-1)
Quicksort(A,q+1,r)
end
```

- To simplify, assume distinct elements: $T(n) = 2T(n/2) + \Theta(n) = \Theta(n \log n)$
 - Lucky always an even element: $T(n) = 2T(0) + T(n-1) + \Theta(n) = \Theta(n^2)$
 - O Unlucky:
- How to avoid bad case?
 - Partition around middle element (does not work!)
 - Idea: Partition around a random element!

Quicksort (cont'd.)

- Partition around a <u>Randomly</u> chosen element and let T(n) be the <u>expected</u> time to sort.
- Consider the case where the partition is (*k*,*n*-*k*-1). In this case, the expected time to terminate is:

$$T(k) + T(n-k-1) + \Theta(n)$$

• Condition on **k** being a specific value, note that any value of **k** from **o** to **n-1** is equally likely:

$$T(n) = \sum_{k} \Pr[(k, n - k - 1) \text{ split}] T(n \mid (k, n - k - 1) \text{ split})$$

$$= \frac{1}{n} \sum_{k} [T(k) + T(n - k - 1) + \Theta(n)]$$

$$= \frac{2}{n} \sum_{k=1}^{n-1} [T(k) + \Theta(n)]$$

Solving the recurrence

Next:

We try to prove that $T(n) \le an \log n + b$

First, Choose *b* large enough to satisfy $T(1) \le a \log 1 + b = b$

Inductive step:

$$T(n) = \frac{2}{n} \sum_{k=1}^{n-1} T(k) + \Theta(n) \le \frac{2}{n} \sum_{k=1}^{n-1} ak \log k + b + \Theta(n)$$

$$= \frac{2a}{n} \sum_{k=1}^{n-1} k \log k + \frac{2}{n} nb + \Theta(n)$$
Need to prove this is $\le \frac{1}{2} n^2 \log n - \frac{1}{8} n^2$

$$\le \frac{2a}{n} \left(\frac{1}{2} n^2 \log n - \frac{1}{8} n^2 \right) + 2b + \Theta(n)$$

$$= an \log n + b + \left(\Theta(n) + b - \frac{an}{4} \right)$$

Technical Lemma

• We need to show $n^2 \log n$ bound is true.

$$\sum_{k=1}^{n-1} k \log k = \sum_{k=1}^{\left[\frac{n}{2}-1\right]} k \log k + \sum_{\left[\frac{n}{2}\right]}^{n-1} k \log k$$

$$\leq 2\log n \left(\sum_{k=1}^{n-1} k - \sum_{k=1}^{\left[\frac{n}{2}-1\right]} k \right)$$

$$\leq \log n \frac{n(n-1)}{2} - \frac{n}{2} \left(\frac{n}{2} - 1 \right)$$

$$\leq \frac{1}{2} n^2 \log n - \frac{n^2}{8}$$

Heaps, Priority Queues and Heap Sort

Priority Queue

- Handles a collection of items, called keys.
- There exists a way to compare keys to each other. This is called an order relation.
- The result of these comparisons determines the priority of the keys.
- Operations supported:
 - o insert a key
 - *Remove* the largest key

Applications

- Scheduling
- Operating systems
- Keeping track of largest *n* elements in a sequence
- Sorting

Methods of a Priority Queue

- *Initialize*: initialize the structure
- *Insert (key)*: insert a new key
- Remove Max: return and remove largest key

PQ-Sort in procedural pseudocode

- (sorting an array with using a priority queue)
 - Initialize
 - o for i = 1 to nInsert (a[i])
 - o for i := n downto 1
 a[i] :=RemoveMax

How to Implement a Priority Queue

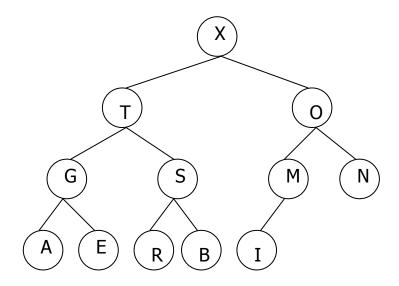
Implementation	Insert	Insert Remove Max		Average	
Unsorted Array or Linked List	O(1)	O(n)	O(1)	O(n)	
Sorted Array or Linked List	O(n)	O(1)	O(1)	O(n)	
Heap	<i>O(</i> log <i>n)</i>	$O(\log n)$	$O(\log n)$	<i>O(</i> log <i>n)</i>	

Heap

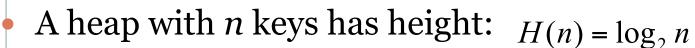


keys, with the following properties:

- o partial order:
 - * key (child) < key(parent)</pre>
- o left-filled levels:
 - * the last level is left-filled
 - * the other levels are full



Logarithmic Height



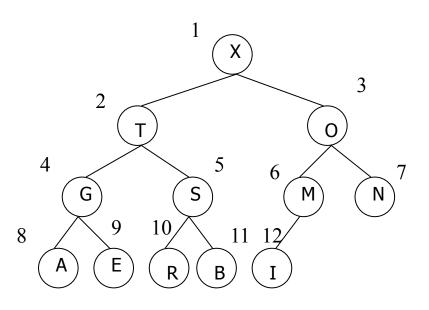
• Proof:

Let n be the number of keys, and H(n) be the height. We have:

$$2^{H(n)-1} < n < 2^{H(n)}$$

Taking logarithm of both sides; the result will follow.

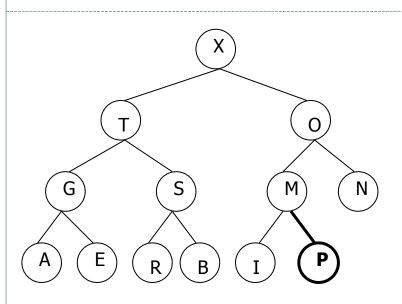
Heap Representations



- left_child(i)= 2i
- right_child(i) = 2i+1
- parent(j) = j div 2

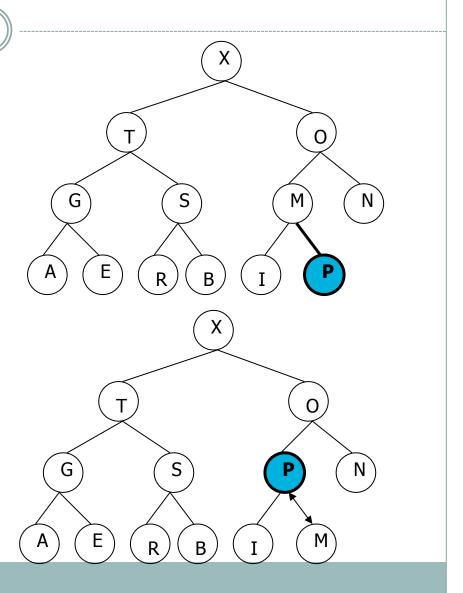
X	Т	О	G	S	M	N	A	Е	R	В	I
1	2	3	4	5	6	7	8	9	10	11	12

Heap Insertion



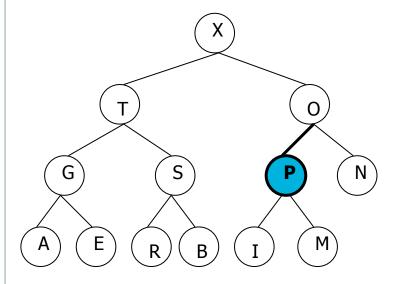
•Add the key in the next available spot in the heap.

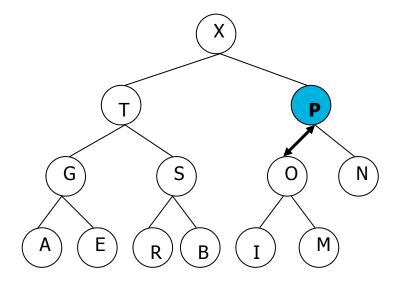
- *Upheap* checks if the new node is greater than its parent. If so, it switches the two.
- •Upheap continues up the tree



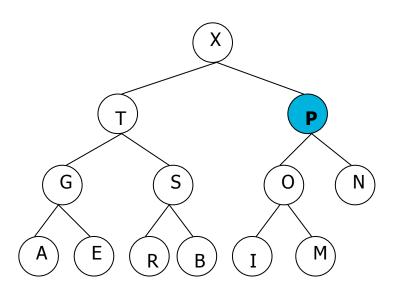
Heap Insertion







Heap Insertion



- *Upheap* terminates when new key is less than the key of its *parent* or the *top of the heap* is reached.
- (total #switches) <= (height of tree-1) = log n

Heapify Algorithm

• Assumes L and R sub-trees of *i* are already Heaps and makes tree rooted at *i* a Heap:

```
Heapify(A,i,n)

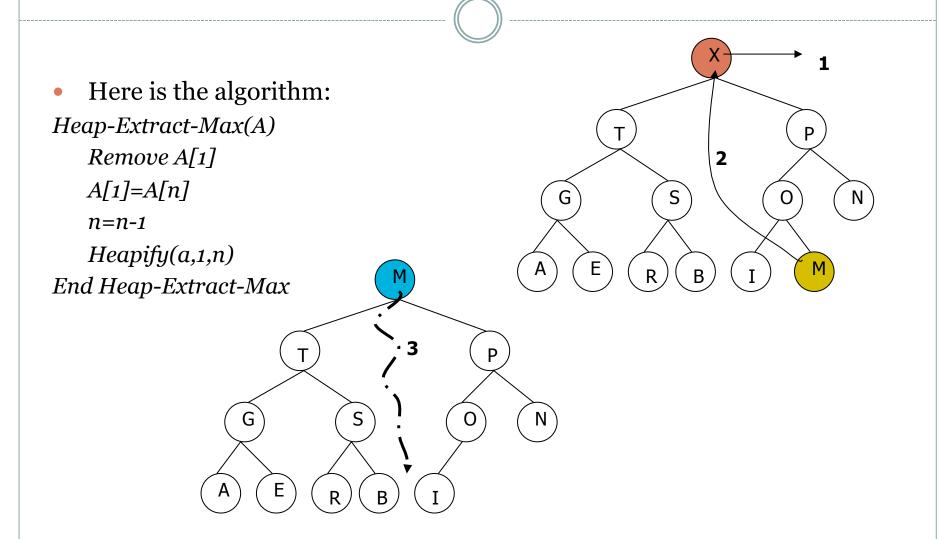
If (2i<=n) & (A[2i]>A[i]) Then
largest=2i

Else largest=i

If (2i+1<=n) & (A[2i+1]>A[largest]) Then
largest=2i+1

If (largest != i) Then
Exchange (A[i],A[largest])
Heapify(A,largest,n)
Endif
End Heapify
```

Extracting the Maximum from a Heap:



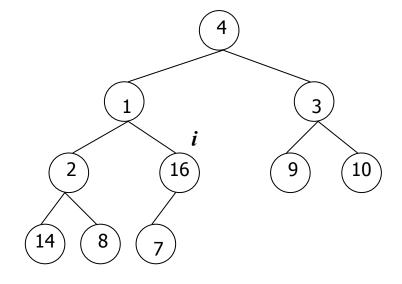
Building a Heap

Builds a heap from an unsorted array:

Build_Heap(A,n)
For i=floor(n/2) down to 1 do
Heapify(A,i,n)

End Build_Heap

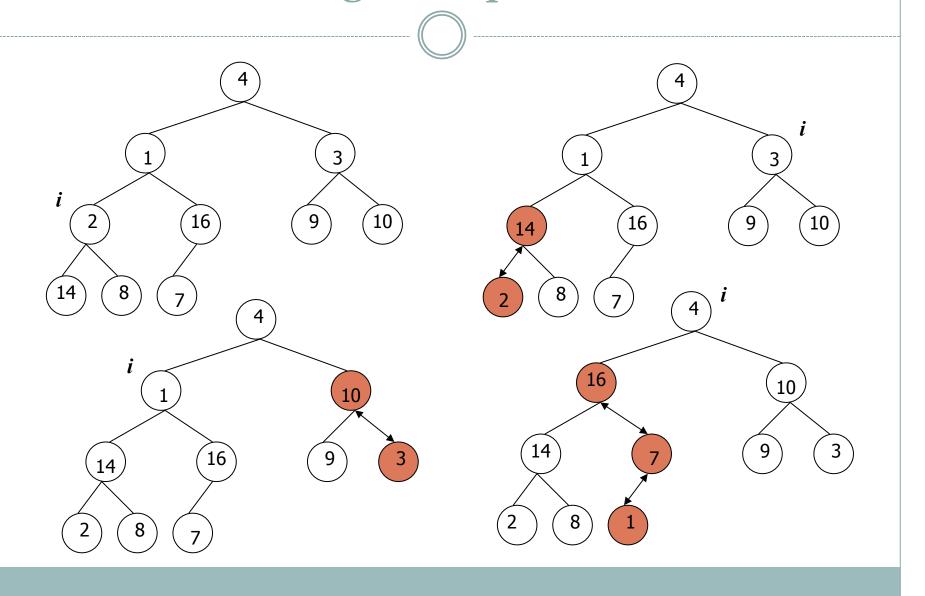
Example:



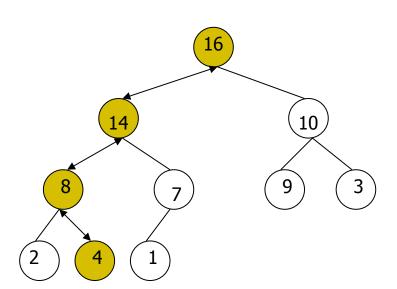
 \boldsymbol{A}

4	1	3	2	16	9	10	14	8	7
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Building a Heap (cont'd.)



Building a Heap (cont'd.)



Running time of Building a Heap

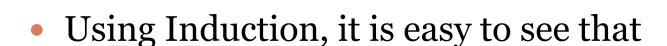


- $O(n \log n)$ is trivial: n calls of Heapify, each of cost $O(\log n)$
- Tighter Bound: *O(n)*
 - The cost of "Heapify" is proportional to the number of levels visited (height of node *i*)
 - Assume $n=2^{k-1}$ (complete binary tree):
 - **For each leaf node, the number of levels visited is 1,**
 - For each node at next level is 2,
 - **x** 3 for next level, etc.

Total # of levels visited =
$$\frac{n+1}{2} \times 1 + \frac{n+1}{4} \times 2 + \frac{n+1}{8} \times 3 + \dots + \frac{n+1}{2^{\log(n+1)}} \times \log(n+1)$$

= $(n+1) \sum_{i=0}^{\log(n+1)} \frac{i}{2^i}$

Running time of Building a Heap (cont'd.)



$$\sum_{i=0}^{\log(n+1)} \frac{i}{2^i} = O(1)$$

Implying:

$$T(n) = \text{Total } \# \text{ of levels visited} = O(n)$$

Heapsort

Heapsort(A,n)

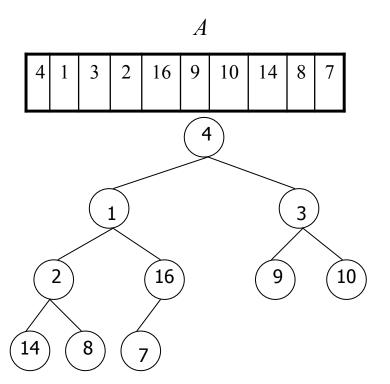
Build-Heap(A,n)

For i=n downto 2 do Exchange A[1] & A[i]

Heapify(A,1,i)

End For

End Heapsort

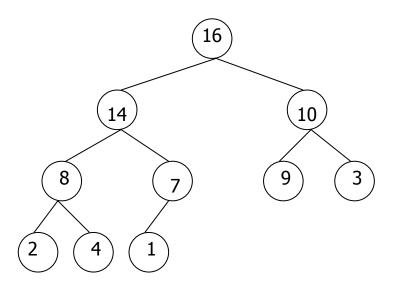


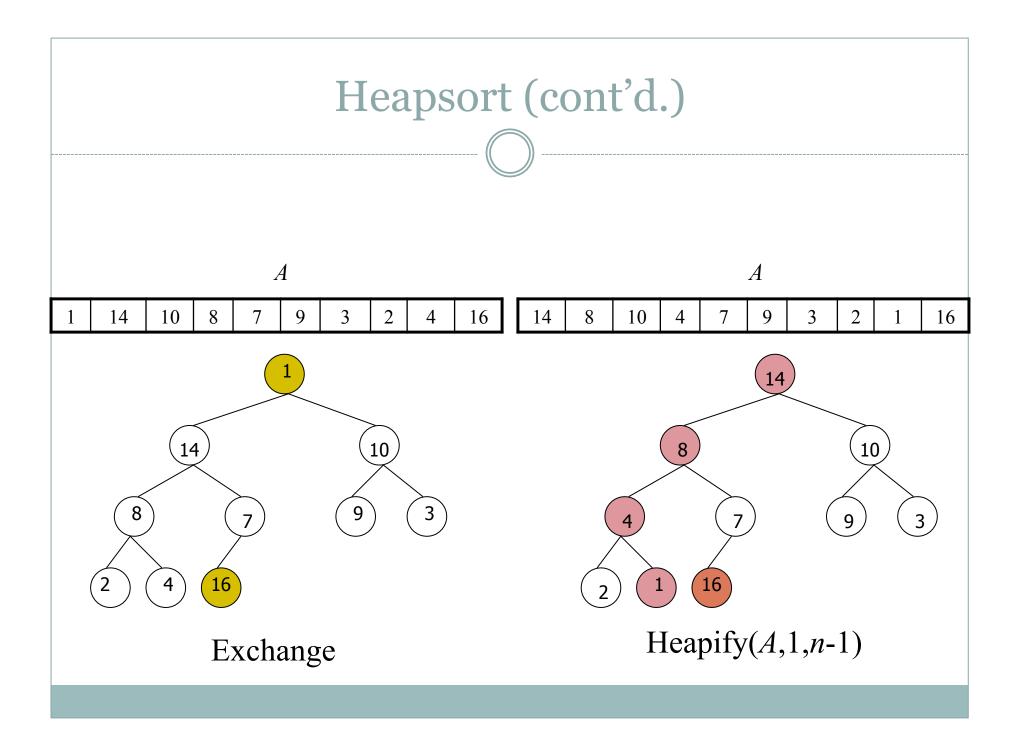
Heapsort (cont'd.)

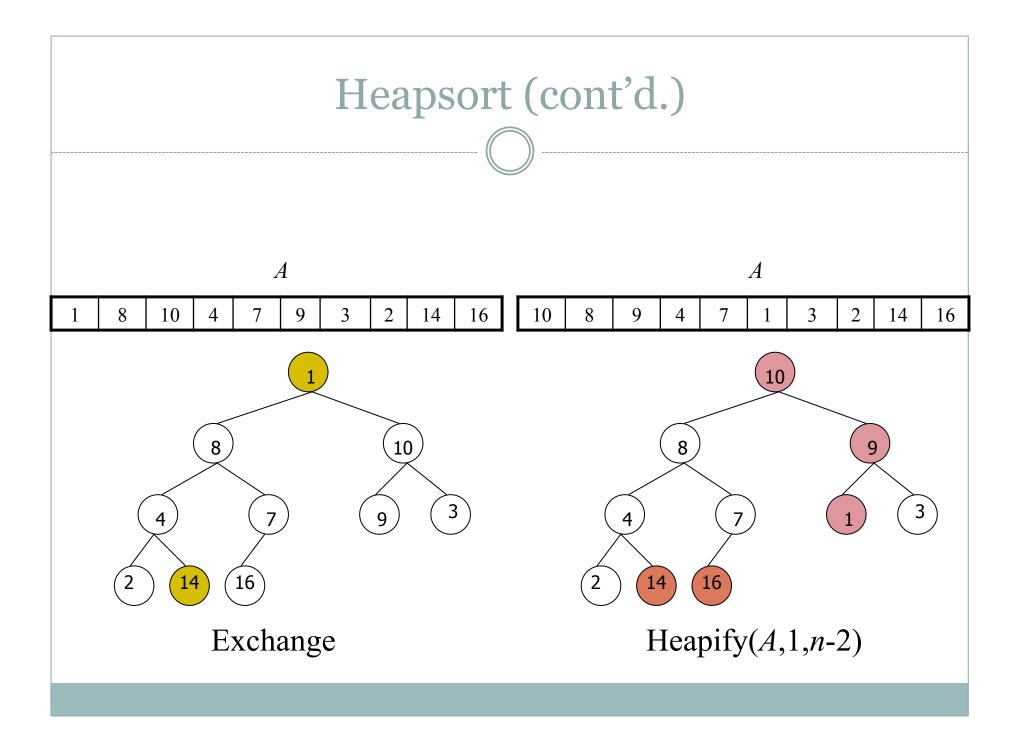


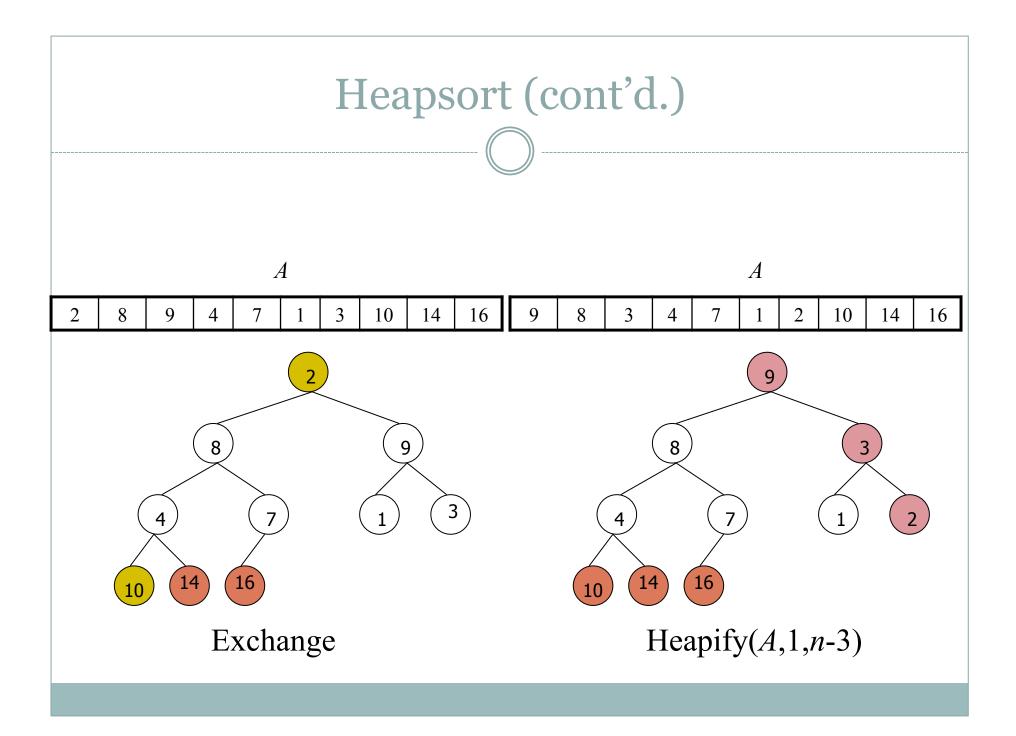
 A

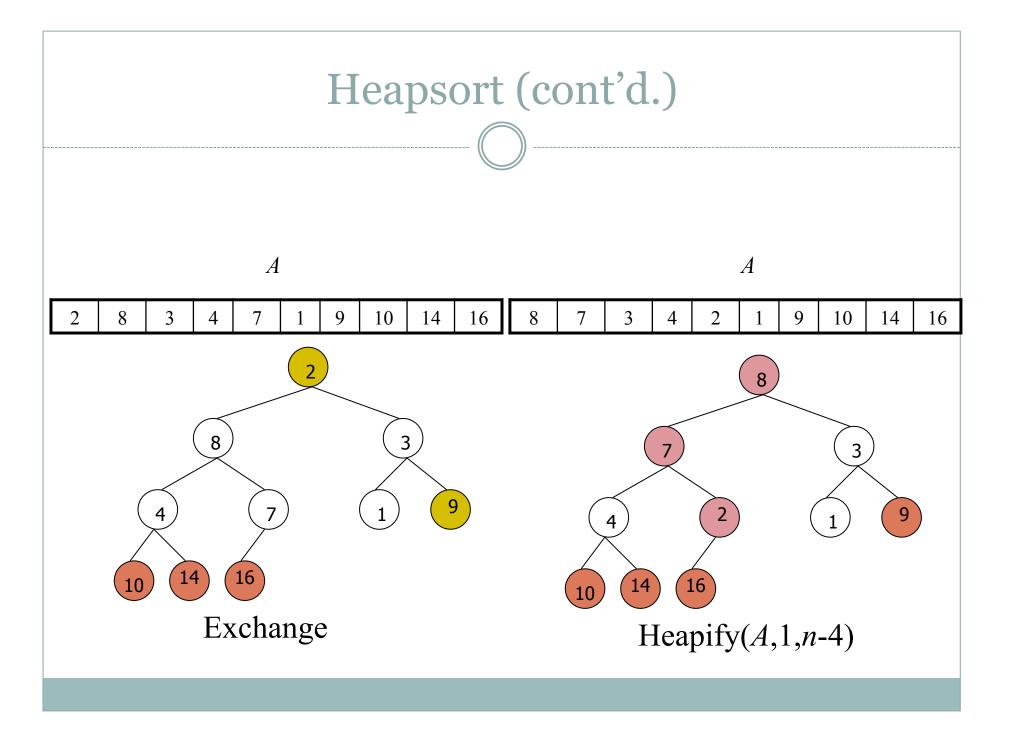
 16
 14
 10
 8
 7
 9
 3
 2
 4
 1

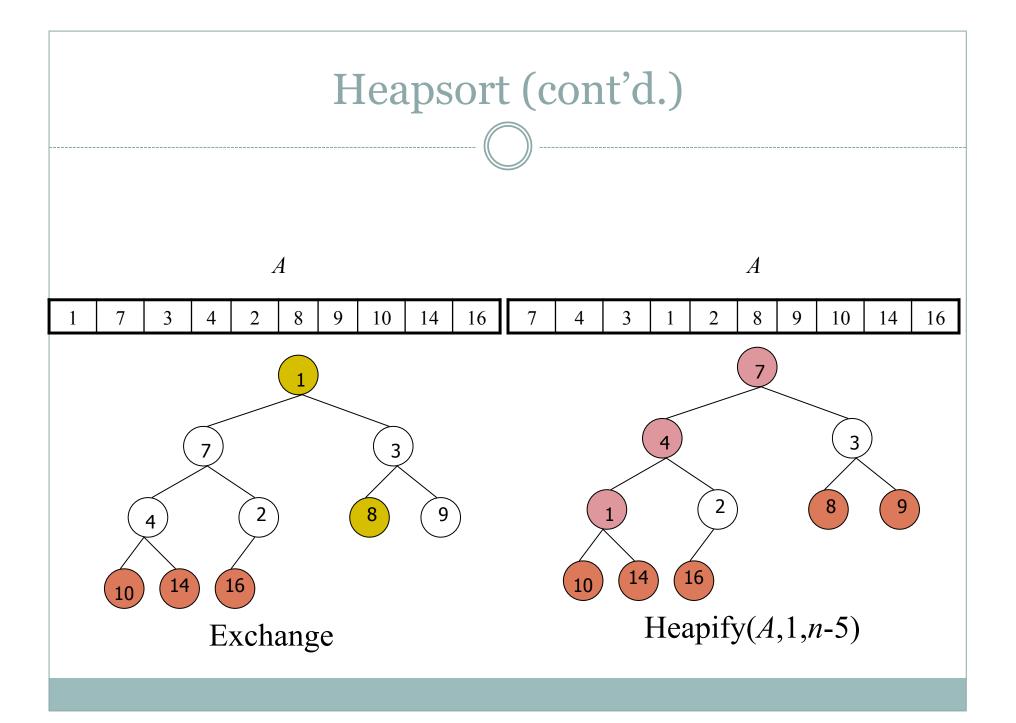










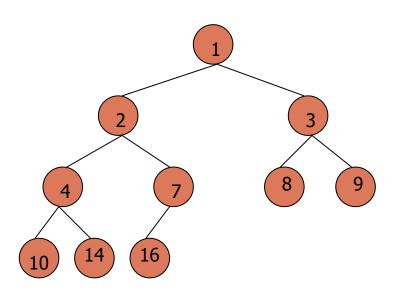


Heapsort (cont'd.)



A

1	2	3	4	7	8	9	10	14	16
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Running Time

Heapsort(A,n)

Build-Heap (A,n)	O(n)
For $i=n$ downto 2 do	n-1 Times
Exchange A[1] & A[i]	O(1)
Heapify $(A, 1, i)$	$O(\log n)$

End Heapsort

• Total Running time: $O(n \log n)$