

# Priority Queues, Heaps, and Heapsort

CSE 2320 – Algorithms and Data Structures  
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# Priority Queues

- So far we have seen sorting methods that works in **batch mode**:
  - They are given all the items at once
  - They sort the items.
  - Done!
- Another case of interest is **online** methods, that deal with data that change.
- Goal: support (efficiently):
  - Insertion of a new element.
  - Deletion of the max element.
  - Initialization (organizing an initial set of data).
- The abstract data type that supports these operations is called **priority queue**.

# Priority Queues - Applications

- Scheduling:
  - Flight take-offs and landings.
  - Programs getting executed on a computer.
  - Real-time requests for information on a database system.
  - Computer simulations and games, to schedule a sequence of events.
- Waiting lists:
  - Students getting admitted to college.
  - Patients getting admitted to a hospital.
- Lots more...

# Priority Queues and Sorting

- Priority queues support:
  - Insertion of a new element.
  - Deletion of the max element.
  - Initialization (organizing an initial set of data).
- These operations support applications that batch methods, like quicksort, mergesort, do not support.
- However, these operations can also support sorting:
- Given items to sort:
  - Initialize a priority queue that contains those items.
  - Initialize result to empty list.
  - While the priority queue is not empty:
    - Remove max element from queue, add it to beginning of result.
- We will see an implementation (heapsort) of this algorithm that takes  $\Theta(N \lg N)$  time.

# Naïve Implementation Using Arrays

- Initialization:
  - Given N data, just store them on an array.
  - Time:  $\Theta(???)$
- Insertion of a new item:
  - (Assumption: the array has enough memory.)
  - Store the item at the end of the array.
  - Time:  $\Theta(???)$
- Deletion of max element:
  - Scan the array to find max item.
  - Delete that item.
  - Time:  $\Theta(???)$

# Naïve Implementation Using Arrays

- Initialization:
  - Given  $N$  data, just store them on an array.
  - Time:  $\Theta(N)$ , good!
- Insertion of a new item:
  - (Assumption: the array has enough memory.)
  - Store the item at the end of the array.
  - Time:  $\Theta(1)$ , good!
- Deletion of max element:
  - Scan the array to find max item.
  - Delete that item.
  - Time:  $\Theta(N)$ , bad!

# Naïve Implementation Using Lists

- Initialization:
  - Given  $N$  data, just store them on an list.
  - Time:  $\Theta(N)$ , good!
- Insertion of a new item:
  - Store the item at the beginning (or end) of the list.
  - Time:  $\Theta(1)$ , good!
- Deletion of max element:
  - Scan the list to find max item.
  - Delete that item.
  - Time:  $\Theta(N)$ , bad!

# Using Ordered Arrays/Lists

- Initialization:
  - Given  $N$  data, sort them.
  - Time:  $\Theta(???)$
- Insertion of a new item:
  - (Assumption: if using an array, it must have enough memory.)
  - Insert the item at the right place, to keep array/list sorted.
  - Time:  $\Theta(???)$
- Deletion of max element:
  - Delete the last item.
  - Time:  $\Theta(???)$



# Using Ordered Arrays/Lists

- Initialization:
  - Given  $N$  data, sort them.
  - Time:  $O(N \lg N)$ . OK!
- Insertion of a new item:
  - (Assumption: if using an array, it must have enough memory.)
  - Insert the item at the right place, to keep array/list sorted.
  - Time:  $O(N)$ . Bad!
- Deletion of max element:
  - Delete the last item.
  - Time:  $\Theta(1)$ . Good!

# Using Heaps (New Data Type)

- Initialization:
  - Given  $N$  data, **heapify** them (we will see how in a few slides).
  - Time:  $\Theta(N)$ . Good!
- Insertion of a new item:
  - Insert the item at the right place, to maintain the **heap property**. (details in a few slides).
  - Time:  $O(\lg N)$ . Good!
- Deletion of max element:
  - Delete the first item.
  - Rearrange other items, to maintain the **heap property**. (details in a few slides).
  - Time:  $O(\lg N)$ . Good!

# Definition of Heaps

- We have two equivalent representations of heaps:
  - As binary trees.
  - As arrays.
- Thus, we have two logically equivalent definitions:
- A binary tree is a heap if, for every node  $N$  in that tree, the key of  $N$  is larger than or equal to the keys of the children of  $N$ , if any.
- An array  $A$  (**with 1 as the first index**) is a heap if, for every position  $N$  of  $A$ :
  - If  $A[2N]$  is not out of bounds, then  $A[N] \geq A[2N]$ .
  - If  $A[2N + 1]$  is not out of bounds, then  $A[N] \geq A[2*N + 1]$ .

# Representing a Heap

- Consider this array:

position	1	2	3	4	5	6	7	8	9	10	11	12
value	X	T	O	G	S	M	N	A	E	R	A	I

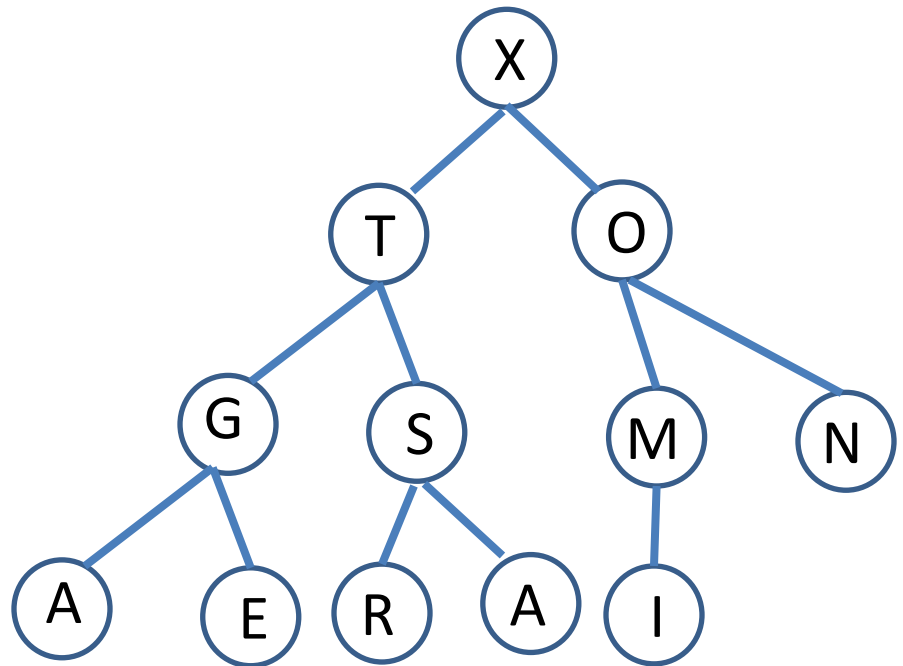
- We can draw the array as a tree.
  - The children of  $A[N]$  are  $A[2N]$  and  $A[2N+1]$ .

# Representing a Heap

- Consider this array:

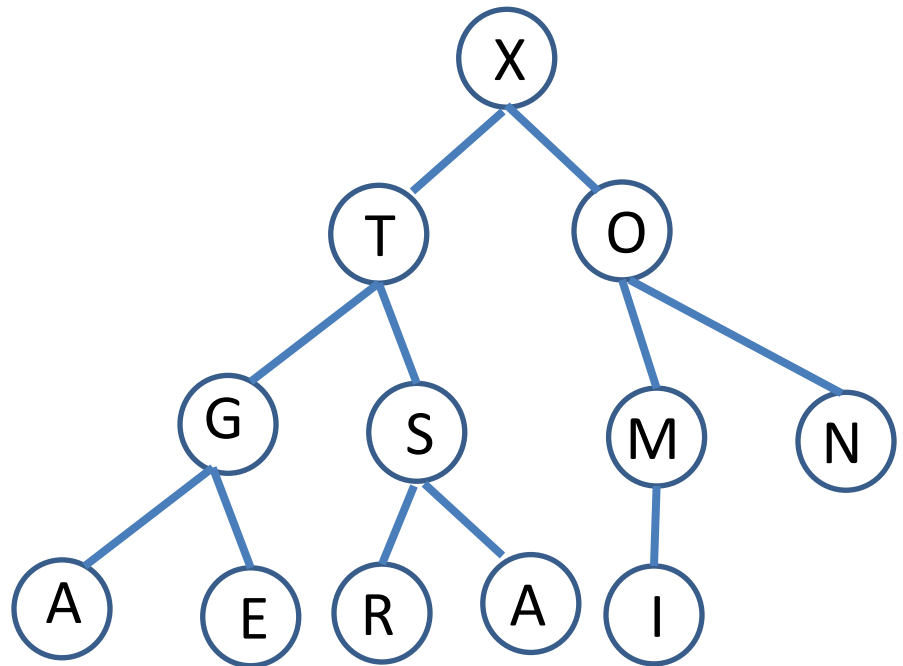
position	1	2	3	4	5	6	7	8	9	10	11	12
value	X	T	O	G	S	M	N	A	E	R	A	I

- We can draw the array as a tree.
  - The children of  $A[N]$  are  $A[2N]$  and  $A[2N+1]$ .
  - This example shows that the tree and array representations are equivalent.



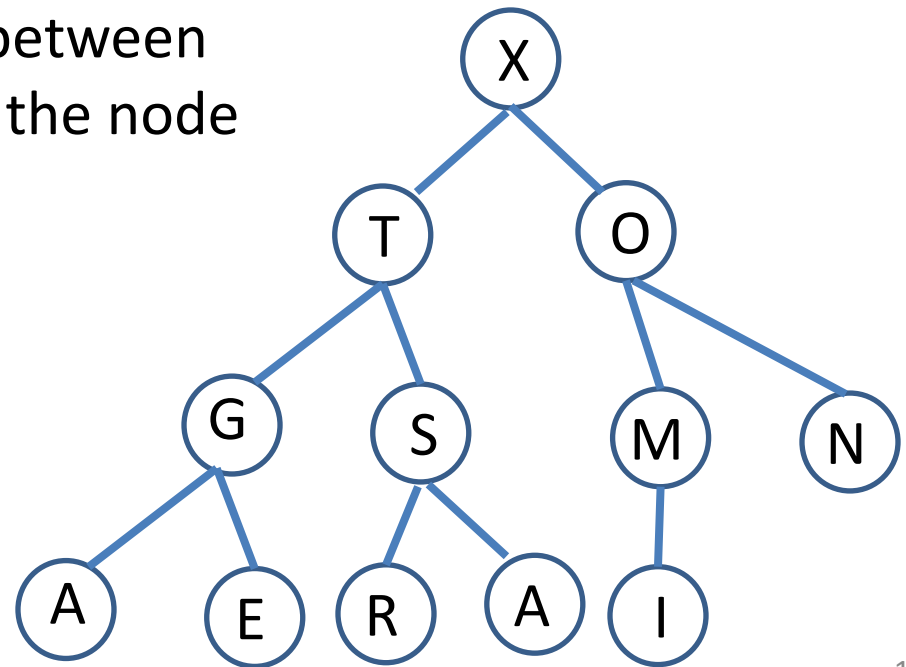
# Representing a Heap

- A binary tree representing a heap should be **complete**.
- All levels are full, except possibly for the last level.
- At the last level:
  - Nodes are placed on the left.
  - Empty positions are placed on the right.



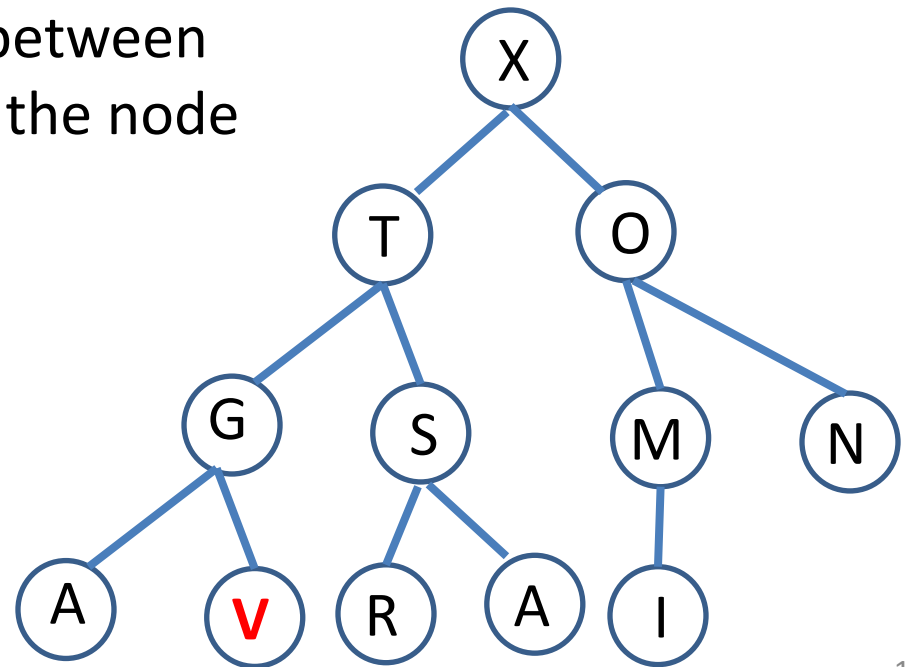
# Increasing a Key

- Also called “increasing the priority” of an item.
- Such an operation can lead to violation of the heap property.
- Easy to fix:
  - Exchange items as needed, between node and parent, starting at the node that changed key.



# Increasing a Key

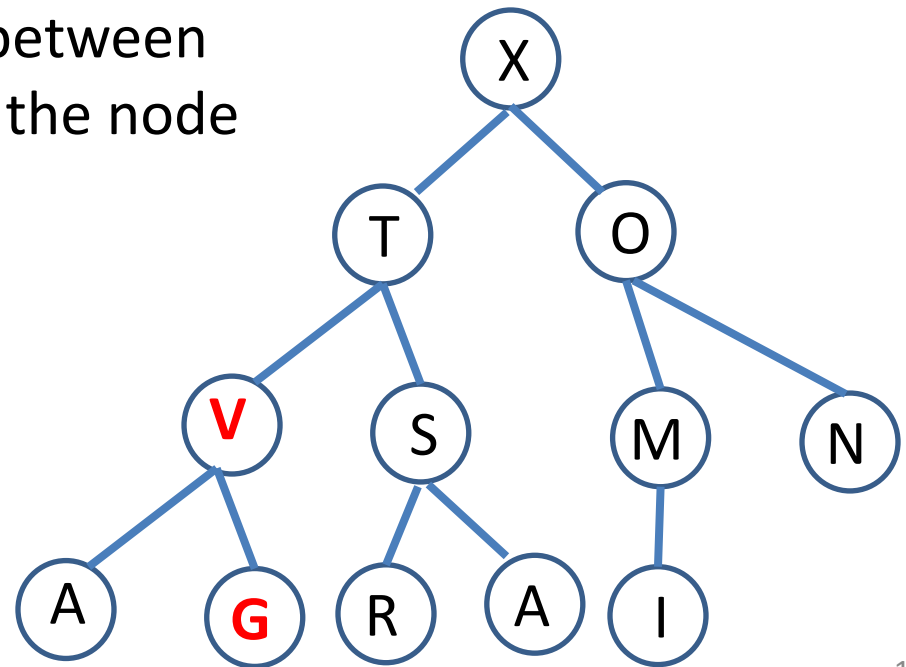
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- Example:
  - An E changes to a V.





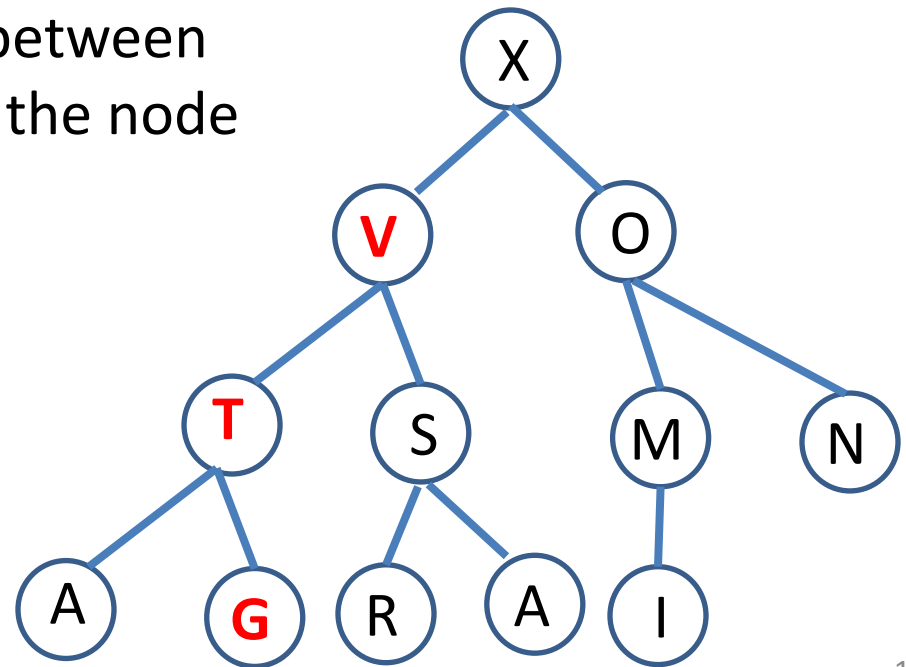
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- Example:
  - An E changes to a V.
  - Exchange V and G. Done?



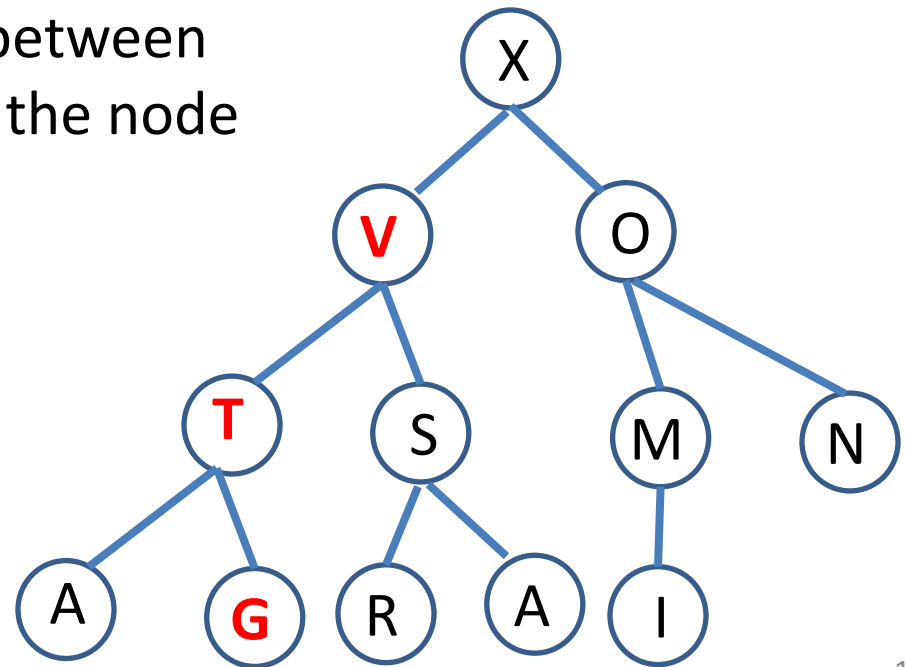
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- Such an operation can lead to violation of the heap property.
- Easy to fix:
  - Exchange items as needed, between node and parent, starting at the node that changed key.
- Example:
  - An E changes to a V.
  - Exchange V and G.
  - Exchange V and T. Done?



# Increasing a Key

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- Such an operation can lead to violation of the heap property.
- Easy to fix:
  - Exchange items as needed, between node and parent, starting at the node that changed key.
- Example:
  - An E changes to a V.
  - Exchange V and G.
  - Exchange V and T. Done.

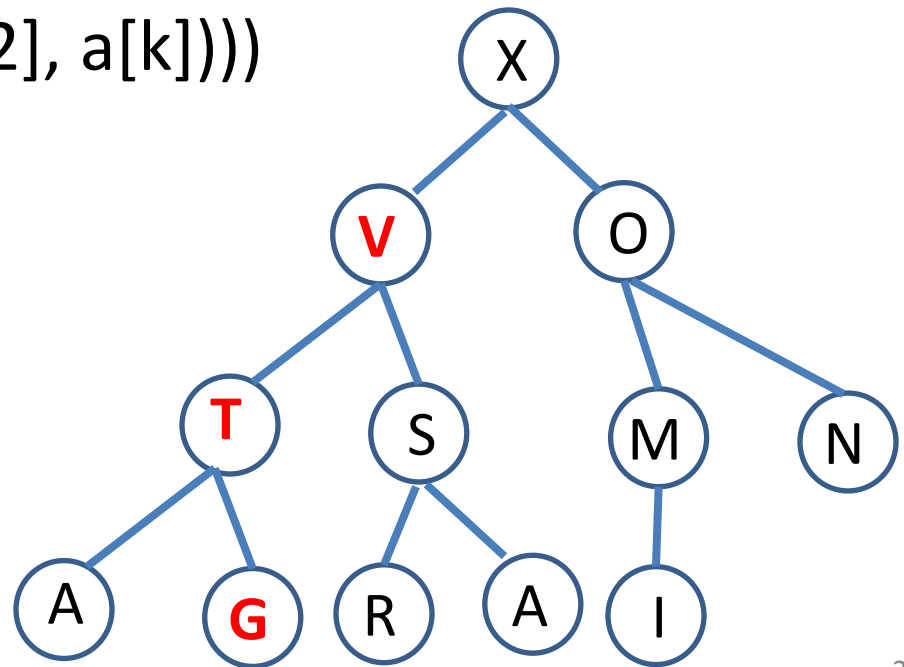


# Increasing a Key

- Implementation:

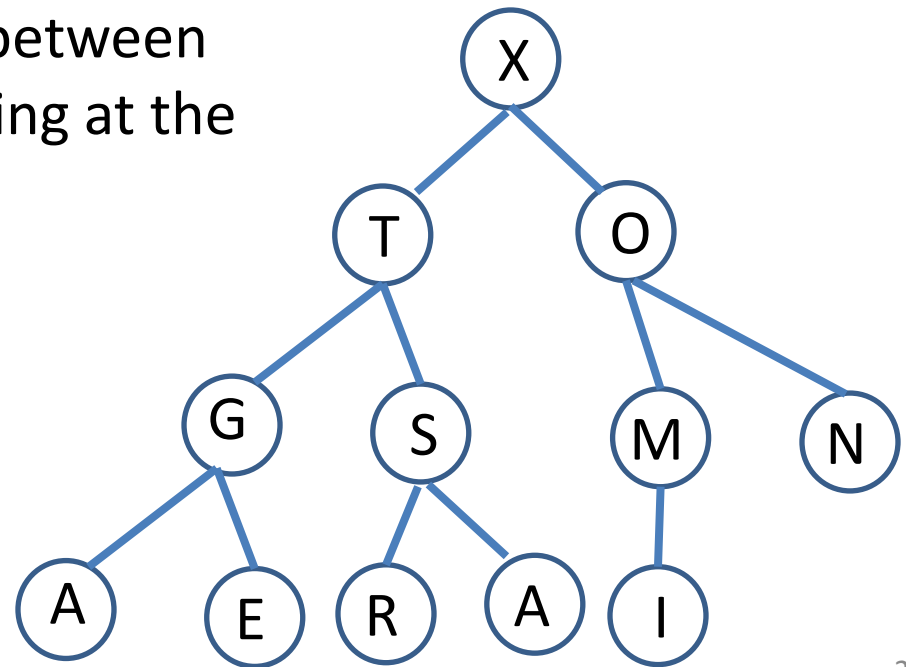
```
fixUp(Item a[], int k)
```

```
{  
  while ((k > 1) && (less(a[k/2], a[k])))  
  {  
    exch(a[k], a[k/2]);  
    k = k/2;  
  }  
}
```



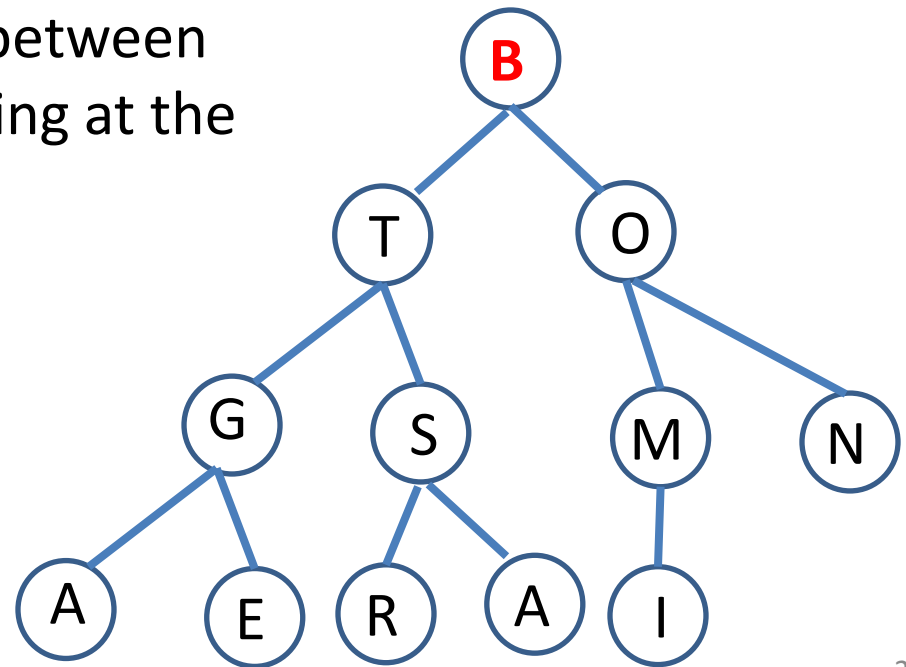
# Decreasing a Key

- Also called “decreasing the priority” of an item.
- Such an operation can lead to violation of the heap property.
- Easy to fix:
  - Exchange items as needed, between node and **largest** child, starting at the node that changed key.



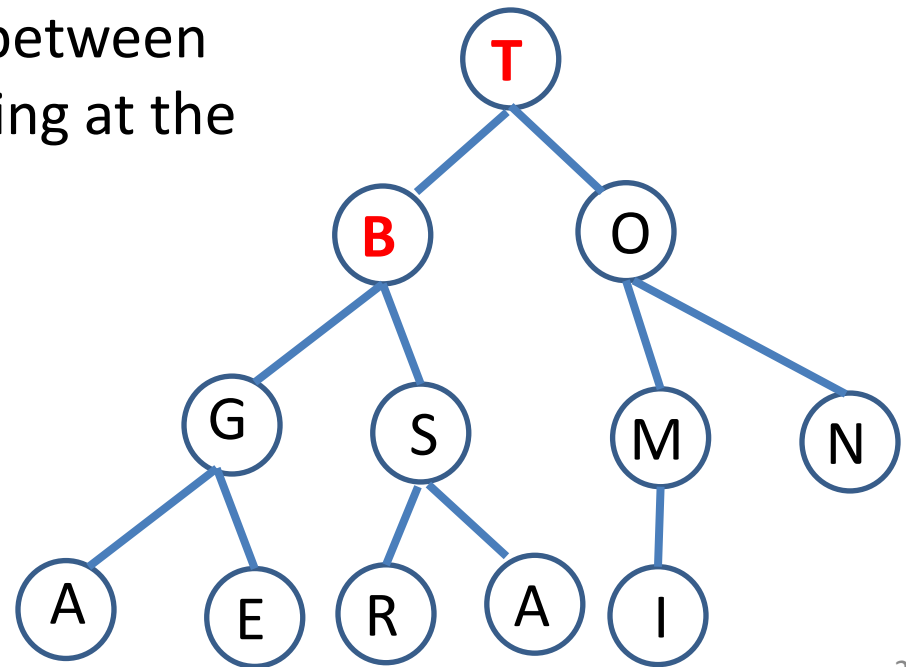
# Decreasing a Key

- Also called “decreasing the priority” of an item.
- Such an operation can lead to violation of the heap property.
- Easy to fix:
  - Exchange items as needed, between node and **largest** child, starting at the node that changed key.
- Example:
  - An X changes to a B.



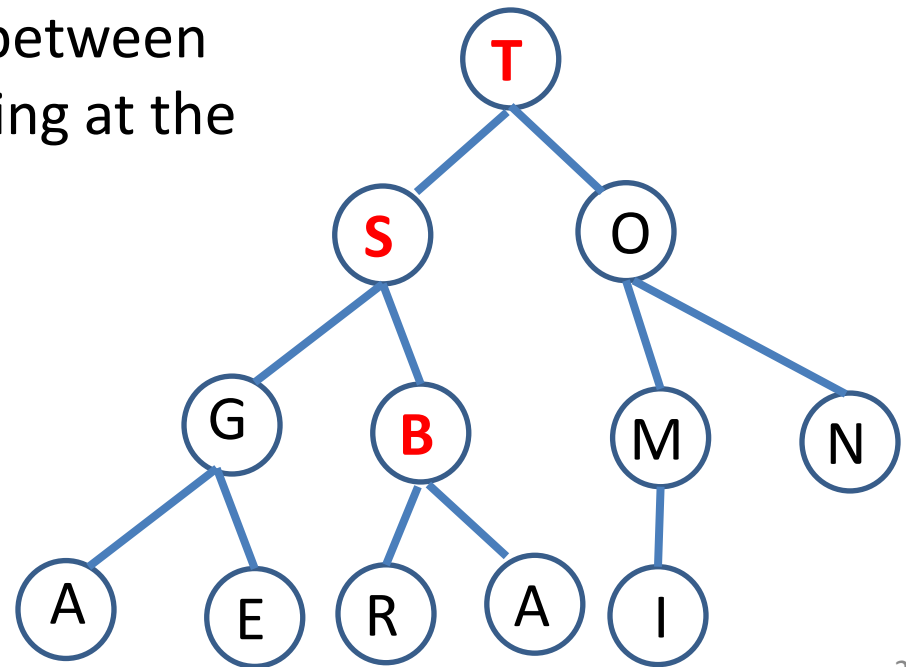
# Decreasing a Key

- Also called “decreasing the priority” of an item.
- Such an operation can lead to violation of the heap property.
- Easy to fix:
  - Exchange items as needed, between node and **largest** child, starting at the node that changed key.
- Example:
  - An X changes to a B.
  - Exchange B and T.



# Decreasing a Key

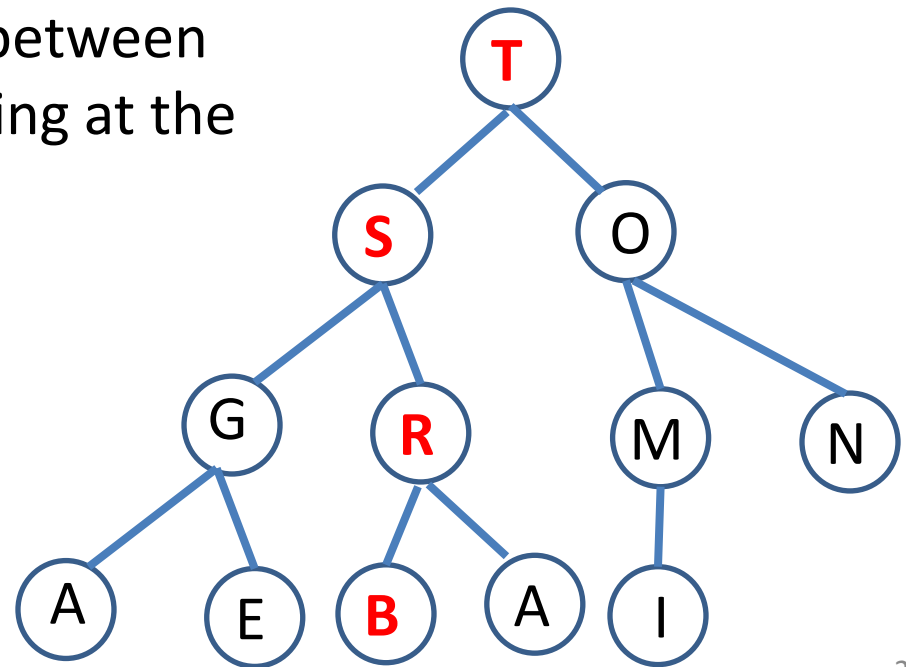
- Also called “decreasing the priority” of an item.
- Such an operation can lead to violation of the heap property.
- Easy to fix:
  - Exchange items as needed, between node and **largest** child, starting at the node that changed key.
- Example:
  - An X changes to a B.
  - Exchange B and T.
  - Exchange B and S.





# Decreasing a Key

- Also called “decreasing the priority” of an item.
- Such an operation can lead to violation of the heap property.
- Easy to fix:
  - Exchange items as needed, between node and **largest** child, starting at the node that changed key.
- Example:
  - An X changes to a B.
  - Exchange B and T.
  - Exchange B and S.
  - Exchange B and R.

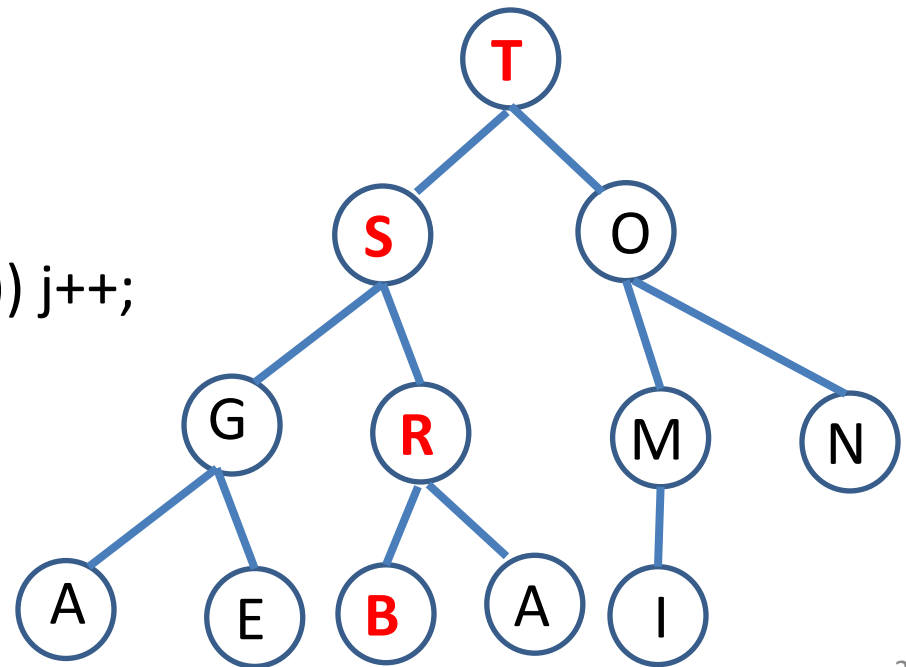


# Decreasing a Key

- Implementation:

```
fixDown(Item a[], int k, int N)
```

```
{  
  int j;  
  while (2*k <= N)  
  {  
    j = 2*k;  
    if ((j < N) && less ((a[j], a[j+1]))) j++;  
    if (!less(a[k], a[j])) break;  
    exch(a[k], a[j]); k = j;  
  }  
}
```



# Insertions and Deletions

- To insert an item to a heap:
  - Insert the item to the end of the heap.
  - Call fix up to restore the heap property.
  - Time =  $O(\log n)$
- The only element we care to delete from a heap is the maximum element.
- This element is always the first element of the heap.
- To delete the maximum element:
  - Exchange the first and last elements of the heap.
  - Delete the last element (which is the maximum element).
  - Call fixDown to restore the heap property.
  - Time =  $O(\log n)$

# Insertions and Deletions

- To insert an item to a heap:
  - Insert the item to the end of the heap.
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- To delete the maximum element:
  - Exchange the first and last elements of the heap.
  - Delete the last element (which is the maximum element).
  - Call fixDown to restore the heap property.
  - Time =  $O(\lg N)$

# Batch Initialization

- Batch initialization of a heap is the process of converting an unsorted array of data into a heap.
- We will see two methods that are pretty easy to implement:
- **Top-down batch initialization.**
  - $O(N \lg N)$  time.
  - $O(N)$  extra space (in addition to the space that the input array already takes).
- **Bottom-up batch initialization.**
  - $O(N)$  time.
  - $O(1)$  extra space (in addition to the space that the input array already takes).

# Top-Down Batch Initialization

```
Heap top_down_heap_init(Item * array, int N)
```

```
    Heap result = newHeap(N).
```

```
    for counter = 0, ..., N-1.
```

```
        heap_insert(array[counter]).
```

```
    return result.
```

- How much time does this take?

# Top-Down Batch Initialization

Heap top\_down\_heap\_init(Item \* array, int N)

Heap result = newHeap(N).

for counter = 0, ..., N-1.

heap\_insert(array[counter]).

return result.

- How much time does this take?
  - We need to do N insertions.
  - Each insertion takes  $O(\lg N)$  time.
  - So, in total, we need  $O(N \lg N)$  time.

# Bottom-Up Batch Initialization

```
struct heap_struct  
{  
    int length;  
    Item * array;  
};
```

```
typedef struct heap_struct * Heap;  
  
Heap bottom_up_heap_init(Item * array, int N)  
    for counter = N/2, ..., 1  
        fixDown(array, counter, N).  
  
Heap result = malloc(sizeof(*result)).  
result.array = array.  
result.N = N.  
return result.
```



# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 7
- `fixDown(counter, N):`

position	1	2	3	4	5	6	*7	8	9	10	11	12	13	<u>14</u>
value	50	40	30	15	60	10	28	45	35	55	95	90	85	60

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 7
- `fixDown(counter, N):`

position	1	2	3	4	5	6	*7	8	9	10	11	12	13	<u>14</u>
value	50	40	30	15	60	10	60	45	35	55	95	90	85	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 6
- `fixDown(counter, N):`

position	1	2	3	4	5	*6	7	8	9	10	11	<u>12</u>	<u>13</u>	14
value	50	40	30	15	60	10	60	45	35	55	95	90	85	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 6
- `fixDown(counter, N):`

position	1	2	3	4	5	*6	7	8	9	10	11	<u>12</u>	<u>13</u>	14
value	50	40	30	15	60	90	60	45	35	55	95	10	85	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 5
- `fixDown(counter, N):`

position	1	2	3	4	*5	6	7	8	9	<u>10</u>	<u>11</u>	12	13	14
value	50	40	30	15	60	90	60	45	35	55	95	10	85	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 5
- `fixDown(counter, N):`

position	1	2	3	4	*5	6	7	8	9	<u>10</u>	<u>11</u>	12	13	14
value	50	40	30	15	95	90	60	45	35	55	60	10	85	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 4
- `fixDown(counter, N):`

position	1	2	3	*4	5	6	7	<u>8</u>	<u>9</u>	10	11	12	13	14
value	50	40	30	15	95	90	60	45	35	55	60	10	85	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 4
- `fixDown(counter, N):`

position	1	2	3	*4	5	6	7	<u>8</u>	<u>9</u>	10	11	12	13	14
value	50	40	30	45	95	90	60	15	35	55	60	10	85	28



# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 3
- `fixDown(counter, N):`

position	1	2	*3	4	5	<u>6</u>	<u>7</u>	8	9	10	11	12	13	14
value	50	40	30	45	95	90	60	15	35	55	60	10	85	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 3
- `fixDown(counter, N):`

position	1	2	*3	4	5	<u>6</u>	<u>7</u>	8	9	10	11	<u>12</u>	<u>13</u>	14
value	50	40	90	45	95	30	60	15	35	55	60	10	85	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 3
- `fixDown(counter, N):`

position	1	2	*3	4	5	<u>6</u>	<u>7</u>	8	9	10	11	<u>12</u>	<u>13</u>	14
value	50	40	90	45	95	85	60	15	35	55	60	10	30	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 2
- `fixDown(counter, N):`

position	1	*2	3	<u>4</u>	<u>5</u>	6	7	8	9	10	11	12	13	14
value	50	40	90	45	95	85	60	15	35	55	60	10	30	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 2
- `fixDown(counter, N):`

position	1	*2	3	<u>4</u>	<u>5</u>	6	7	8	9	<u>10</u>	<u>11</u>	12	13	14
value	50	95	90	45	40	85	60	15	35	55	60	10	30	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 2
- `fixDown(counter, N):`

position	1	*2	3	<u>4</u>	<u>5</u>	6	7	8	9	<u>10</u>	<u>11</u>	12	13	14
value	50	95	90	45	60	85	60	15	35	55	40	10	30	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 1
- `fixDown(counter, N):`

position	*1	<u>2</u>	<u>3</u>	4	5	6	7	8	9	10	11	12	13	14
value	50	95	90	45	60	85	60	15	35	55	40	10	30	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 1
- fixDown(counter, N):

position	*1	<u>2</u>	<u>3</u>	<u>4</u>	<u>5</u>	6	7	8	9	10	11	12	13	14
value	95	50	90	45	60	85	60	15	35	55	40	10	30	28



# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 1
- `fixDown(counter, N):`

position	<u>*1</u>	<u>2</u>	<u>3</u>	<u>4</u>	<u>5</u>	6	7	8	9	<u>10</u>	<u>11</u>	12	13	14
value	95	60	90	45	50	85	60	15	35	55	40	10	30	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 1
- fixDown(counter, N):

position	<u>*1</u>	<u>2</u>	<u>3</u>	<u>4</u>	<u>5</u>	6	7	8	9	<u>10</u>	<u>11</u>	12	13	14
value	95	60	90	45	55	85	60	15	35	50	40	10	30	28

# Visualizing Bottom-Up Initialization

- $N = 14$
- counter = 1
- DONE!!!
- The heap condition is now satisfied.

position	1	2	3	4	5	6	7	8	9	10	11	12	13	14
value	95	60	90	45	55	85	60	15	35	50	40	10	30	28

# Running Time

- How can we analyze the running time?
- To simplify, suppose that  $N = 2^n - 1$ .
- The counter starts at value ???.
- At that point, we call `fixDown` on a heap of size ???.

```
Heap bottom_up_heap_init(Item * array, int N)
for counter = N/2, ..., 1
    fixDown(array, counter, N).
```

```
Heap result = malloc(sizeof(*result)).
result.array = array.
result.N = N.
return result.
```

# Running Time

- How can we analyze the running time?
- To simplify, suppose that  $N = 2^n - 1$ .
- The counter starts at value  $2^{n-1} - 1$ .
- At that point, we call `fixDown` on a heap of size 3 ( $= 2^2 - 1$ ).
- For counter values between  $2^{n-1} - 1$  and  $2^{n-2}$ , we call `fixDown` on a heap of size  $2^2 - 1$ .

```
Heap bottom_up_heap_init(Item * array, int N)
for counter = N/2, ..., 1
    fixDown(array, counter, N).
```

```
Heap result = malloc(sizeof(*result)).
result.array = array.
result.N = N.
return result.
```

# Running Time

- For counter values between  $2^{n-1} - 1$  and  $2^{n-2}$ , we call `fixDown` on a heap of size  $2^2 - 1$ .
- For counter values between  $2^{n-2} - 1$  and  $2^{n-3}$ , we call `fixDown` on a heap of size ???.
- ...
- For counter value  $2^0$  we call `fixDown` on a heap of size ???.

```
Heap bottom_up_heap_init(Item * array, int N)
  for counter = N/2, ..., 1
    fixDown(array, counter, N).
```

```
Heap result = malloc(sizeof(*result)).
result.array = array.
result.N = N.
return result.
```

# Running Time

- For counter values between  $2^{n-1} - 1$  and  $2^{n-2}$ , we call `fixDown` on a heap of size  $2^2 - 1$ .
- For counter values between  $2^{n-2} - 1$  and  $2^{n-3}$ , we call `fixDown` on a heap of size  $7 (= 2^3 - 1)$ .
- ...
- For counter value  $2^0$  we call `fixDown` on a heap of size  $2^n - 1$ .

```
Heap bottom_up_heap_init(Item * array, int N)
  for counter = N/2, ..., 1
    fixDown(array, counter, N).
```

```
Heap result = malloc(sizeof(*result)).
result.array = array.
result.N = N.
return result.
```

# Running Time

Counter: from	Counter: to	Number of Iterations	Heap Size	Time per Iteration	Time for All Iterations
$2^{n-1} - 1$	$2^{n-2}$	$2^{n-2}$	$2^2 - 1$	$O(2)$	$O(2^{n-2} * 2)$
$2^{n-2} - 1$	$2^{n-3}$	$2^{n-3}$	$2^3 - 1$	$O(3)$	$O(2^{n-3} * 3)$
$2^{n-3} - 1$	$2^{n-4}$	$2^{n-4}$	$2^4 - 1$	$O(4)$	$O(2^{n-4} * 4)$
...					
$2^1 - 1 = 1$	$2^0 = 1$	$2^0 = 1$	$2^n - 1$	$O(n)$	$O(2^0 * n)$

- Sum:  $\sum_{k=0}^{n-2} (2^k * (n - k))$
- This is not that trivial to analyze.
- It turns out that  $\sum_{k=0}^{n-2} (2^k * (n - k)) = \Theta(N)$
- Thus, bottom-up batch initialization takes linear time.



# Bottom-Up Versus Top-Down

- Top-down initialization does not touch the input array.
  - Instead, it creates a new heap, where it inserts the data.
  - Thus, it needs  $O(N)$  extra space, in addition to the space already taken by the input array.
- Bottom-up initialization, instead, changes the input array.
  - The heap does not allocate memory for a new array.
  - Instead, the heap uses the input array as its own array.
  - Consequently, it needs  $O(1)$  extra space, in addition to the space already taken by the input array.

# Heapsort

```
void heapsort(Item a[], int N)
    bottom_up_heap_init(a, N).
    for counter = N, ..., 2
        exch(a[1], a[counter]).
        fixDown(a, 1, counter-1).
```