CS11313 - Spring 2022 Design & Analysis of Algorithms

Heapsort

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Selection Sort?

SELECTION-SORT(array)

```
for i=n-1 \longrightarrow 1:
```

scan this portion of the array *linearly*.

Selection Sort?



HEAP-SORT(array)	rearrange the elements so
<pre>prepare(array)</pre>	— that finding the max is easy!
for i=n-1 → 1:	
<pre>max = FIND-MAX(array, i, 0) insert array[max] into array[i])</pre>	find the maximum element quickly!

Selection Sort?



Roadmap.

- 1. Review Max-Priority Queues and Heaps.
- 2. Learn about Heapsort.

Abstract Data Type (ADT): A specification of the possible operations on a set of values (independent of the implementation).

ADT	Goal	operations
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Queue	Remove the item least-recently added	ENQUEUE, DEQUEUE

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Max-PQ	Remove the largest item	Insert, Max, Del-Max	?	

Unordered List:



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- **insert**: $\Theta(1)$ (insert to the end of the list; order does not matter)
- max: $\Theta(n)$ (linearly search for the max)
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Ordered Array:

1 2 3 3 3 4 5 6	7	9
-----------------	---	---

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Binary Heap:

- insert: $O(\log n)$ (how?)
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Complete Binary Tree:

- All levels are full (except possibly the last level).
- Last level is filled left-to-right.





not complete

complete

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Properties:

• Height if there are *n* nodes: $h = \lfloor \log_2 n \rfloor$



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- Height if there are *n* nodes: $h = \lfloor \log_2 n \rfloor$
- There are $\left\lfloor \frac{n+1}{2} \right\rfloor$ leaves.
- All leaves are at level h or h 1.
- Number of nodes at *internal* level $i = 2^i$



Binary Heap: (max-ordered)

- Structure: Must be a complete binary tree.
- Order: Every node is not less than its children.

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Non-Examples:



order property violated



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0	1	2	3	4	5	6	7	8
7	6	6	5	5	4	4	2	2

array has the tree nodes in level-order



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0	1	2	3	4	5	6	7	8
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array has the tree nodes in level-order



Three simple functions.



Binary Heap: (max-ordered)





Three simple functions.

LEFT(i) return 2*i + 1 left child is at index 2*0 + 1 = 1

RIGHT(i)

return 2*i + 2

Right child is at index 2*0 + 2 = 2

PARENT(i)

return (i-1)/2

Parent of the node at 0 is negative (no parent)

Binary Heap: (max-ordered)





Three simple functions.

LEFT(i) return 2*i + 1left child is at index 2*1 + 1 = 3

RIGHT(i)

return 2*i + 2

Right child is at index 2*1 + 2 = 4

PARENT(i)

return (i-1)/2

Parent is at index (1-1)/2 = 0

Binary Heap: (max-ordered)





Three simple functions.

LEFT(i) return $2 \times i + 1$ left child is at index $2 \times 2 + 1 = 5$

RIGHT(i)

return 2*i + 2

Right child is at index 2*2 + 2 = 6

PARENT(i)

return (i-1)/2

Parent is at index (2-1)/2 = 0

Binary Heap: (max-ordered)



Three simple functions.

LEFT(i)	RIGHT (i)	PARENT (i)
return 2*i + 1	return 2*i + 2	return (i-1)/2
left child is at index 2*3 + 1 = 7	Right child is at index 2*3 + 2 = 8	Parent is at index $(3-1)/2 = 1$

6











1. If an item becomes **larger** than its parent, push it **up** the tree to maintain the heap order property.



SWIM(a[], i, size)

also called SiftUp()
(not shiftup) on wikipedia
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SWIM(a[], i, size)

```
while (i>0 and a[i] > a[PARENT(i)]):
    swap(a[i], a[PARENT(i)])
    i = PARENT(i)
```



Running Time. *O*(log *n*)

1 swap and 1 compare per iteration. The number of iterations is bounded by the tree height.







2. If an item becomes **less** than one of its children, push it **down** the tree to maintain the heap order property.



is less than the other child (3)



2. If an item becomes **less** than one of its children, push it **down** the tree to maintain the heap order property.



SINK(a[], i, size)

also called:

- SIFTDOWN on wikipedia
- MAX-HEAPIFY in our text-book
- FIX-HEAP in the slides of the other sections!









```
SINK(a[], i, size)
```

```
while (LEFT(i) < size):
    k = LEFT(i)

if (RIGHT(i) < size):
    if (a[k] < a[RIGHT(i)]): k = RIGHT(i)

if (a[i] < a[k]):
    swap(a[i], a[k])
    i = k
else: break
</pre>
```

2. If an item becomes **less** than one of its children, push it **down** the tree to maintain the heap order property.



SINK(a[], i, size)

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while (LEFT(i) < size):
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if (a[i] < a[k]):
    swap(a[i], a[k])
    i = k

else: break</pre>
```



Running Time. *O*(log *n*) At most 1 swap and 2 comparisons per iteration The number of iterations is bounded by the tree height.

Max: Always at index 0. $\Theta(1)$



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Insert: Insert at the end of the array and then **swim**.

INSERT(a[], k, size)

a[size] = k size = size + 1 SWIM(a, size-1, size)

O(log *n*): Swim at most to the root.



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adding to the last index is equivalent to filling the last level left-to-right

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Answer.

DEL-RANDOM(a[], size)

```
k = random index in [0, size-1]
swap(a[k], a[size-1])
size = size-1
SINK(a, k, size)
SWIM(a, k, size)
```

Example 1.



Implement a max-PQ that supports the following operation: **del-Random**: Removes a random element from the priority queue.

Answer.

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Implement a max-PQ that supports the following operation: **del-Random**: Removes a random element from the priority queue.

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Example 2.

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Example 2.

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swap(a[k], a[size-1])
size = size-1
SINK(a, k, size)
SWIM(a, k, size)
```




HEAP-SORT(a[], size)

```
heap \leftarrow An empty max heap

for i = 0 \rightarrow n-1:

heap.INSERT(a[i])

for i = n-1 \rightarrow 0:

a[i] = heap.MAX()
```

```
heap.DELETE-MAX()
```







Running Time. (number of compares in the worst case)

• Step 1. $\log_2(1) + \log_2(2) + \log_2(3) + \dots + \log_2(n-1) \le \log_2(n!)$



Running Time. (number of compares in the worst case)

• Step 1.
$$\log_2(1) + \log_2(2) + \log_2(3) + \dots + \log_2(n-1) \le \log_2(n!)$$

insert the second
element into a
heap of size 1
insert the last
element into a
heap of size n-1



Running Time. (number of compares in the worst case)

- Step 1. $\log_2(1) + \log_2(2) + \log_2(3) + \dots + \log_2(n-1) \le \log_2(n!) = O(n \log n)$
- Step 2. $2 \times (\log_2(n-1) + \log_2(n-2) + \log_2(n-3) + \dots + \log_2(1))$ $\leq 2 \times \log_2(n!)$



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• Step 2.
$$2 \times (\log_2(n-1) + \log_2(n-2) + \log_2(n-3) + \dots + \log_2(1))$$

 $\leq 2 \times \log_2(n!) = O(n \log n)$
check the analysis
of the SINK operation!



Running Time. (number of compares in the worst case)

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- Total. $O(n \log n)$



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- Total. $O(n \log n)$



Not asymptotically, but we can still improve the actual running time!





CONSTRUCT-HEAP(a, size)

```
while (size > 1):
   swap(a[0], a[size-1])
   size = size-1
   SINK(a, 0, size)
```

construct a max-heap in-place (convert the array to become a heap)

How? stay tuned!

1





size-1





































































































```
HEAP-SORT(a[], size)
```

```
CONSTRUCT-HEAP(a, size)
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while (size > 1):
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CONSTRUCT-HEAP(a, size)

```
for i = size/2 - 1 → 0:
    SINK(a, i, size)
```

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while (size > 1):
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```

CONSTRUCT-HEAP(a, size)

for i = size/2 - 1 → 0:
 SINK(a, i, size)

sink all the non-leaf nodes

at indices 0 to size/2 - 1





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CONSTRUCT-HEAP(a, size)
```

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    size = size-1
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```
for i = size/2 - 1 → 0:
    SINK(a, i, size)
```



Heap Construction Analysis:

number of nodes

HEAP-SORT(a[], size)

```
CONSTRUCT-HEAP(a, size)
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```
while (size > 1):
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for i = size/2 - 1 → 0:
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Heap Construction Analysis:

• Maximum number of swaps: $(1 \bullet h) + 2(h - 1)$



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CONSTRUCT-HEAP(a, size)
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while (size > 1):
    swap(a[0], a[size-1])
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CONSTRUCT-HEAP(a, size)

```
for i = size/2 - 1 → 0:
    SINK(a, i, size)
```



Heap Construction Analysis:

• Maximum number of swaps: $(1 \cdot h) + 2(h - 1) + 4(h - 2)$



```
CONSTRUCT-HEAP(a, size)
```

```
while (size > 1):
    swap(a[0], a[size-1])
    size = size-1
    SINK(a, 0, size)
```

CONSTRUCT-HEAP(a, size)

```
for i = size/2 - 1 → 0:
    SINK(a, i, size)
```



Heap Construction Analysis:

• Maximum number of swaps: $(1 \cdot h) + 2(h-1) + 4(h-2) + \dots + (\frac{n}{4} \cdot 1)$


```
for i = size/2 - 1 → 0:
    SINK(a, i, size)
```



Heap Construction Analysis:

• Maximum number of swaps: $(1 \cdot h) + 2(h-1) + 4(h-2) + \dots + \frac{n}{4}(1) = O(n)$

tricky sum (math skipped)



for i = size/2 - 1
$$\rightarrow$$
 0:
SINK(a, i, size)



Heap Construction Analysis:

• Maximum number of swaps: $(1 \cdot h) + 2(h-1) + 4(h-2) + \dots + \frac{n}{4}(1) = O(n)$

number of swaps is linear!





Heap Construction Analysis:

- Maximum number of swaps: $(1 \cdot h) + 2(h-1) + 4(h-2) + \dots + \frac{n}{4}(1) = O(n)$
- Maximum number of compares: 2 × number of swaps

____ check the analysis of the SINK operation!

HEAP-SORT(a[], size)

```
CONSTRUCT-HEAP(a, size)
```

```
while (size > 1):
    swap(a[0], a[size-1])
    size = size-1
    SINK(a, 0, size)
```

CONSTRUCT-HEAP(a, size)

```
for i = size/2 - 1 → 0:
    SINK(a, i, size)
```

Think. Why does this heap construction code run in *O*(*n*) while inserting all the elements into a heap takes *O*(*n* log *n*) time?



Heap Construction Analysis:

- Maximum number of swaps: $(1 \cdot h) + 2(h-1) + 4(h-2) + \dots + \frac{n}{4}(1) = O(n)$
- Maximum number of compares: 2 × number of swaps

____ check the analysis of the SINK operation!

Optional

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

$$(1 \cdot h) + 2(h-1) + 4(h-2) + \dots + \frac{n}{4}(1)$$

$$= 2^{0}(1 \cdot h) + 2^{1}(h-1) + 2^{2}(h-2) + \dots + 2^{h-1}$$

$$= \sum_{i=0}^{h-1} 2^{i}(h-i) = \left(\sum_{i=0}^{h-1} 2^{i}h\right) - \left(\sum_{i=0}^{h-1} i2^{i}\right) = h(2^{h}-1) - \left(\sum_{i=0}^{h-1} i2^{i}\right)$$

$$= h(2^{h}-1) - ((h-2)2^{h}+2)$$

$$= h(2^{h}-1) - (h2^{h}-2^{h+1}+2)$$

$$= h2^{h}-h-h2^{h}+2^{h+1}-2$$

$$= 2^{h+1}-2 \quad \longleftarrow h \sim \log_{2} n$$

$$= O(n)$$

Worst Case: $\Theta(n)$ to **construct** the heap and $\Theta(n \log n)$ to **heapsort**.

Average Case: $\Theta(n \log n)$

- **Worst Case:** $\Theta(n)$ to **construct** the heap and $\Theta(n \log n)$ to **heapsort**.
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- **Best Case:** $\Theta(n)$ if all the elements are the same.

Worst Case: $\Theta(n)$ to construct the heap and $\Theta(n \log n)$ to heapsort.Average Case: $\Theta(n \log n)$

Best Case:

 $\Theta(n)$ if all the elements are the same.

Why? Trace on a piece of paper to see why!

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- Average Case: $\Theta(n \log n)$
- **Best Case:** $\Theta(n)$ if all the elements are the same.

Running Time:

• Number of compares: At most $\sim 2n \log_2 n$.

~ $n \log_2 n$ for merge sort and ~ $1.39n \log_2 n$ for quicksort (on random data)

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Average Case:	$\Theta(n \log n)$
Best Case:	$\Theta(n)$ if all the elements are the same.

Running Time:

- Number of compares: At most $\sim 2n \log_2 n$.
- Actual running time: Slower than merge sort and quicksort because
- of the higher number of comparisons and the the poor use of cache.

optimizations are possible

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Memory. Heapsort is an *in-place* sorting algorithm.

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Memory. Heapsort is an *in-place* sorting algorithm.

Bottom line.

- Θ(n log n) in the worst case and also sorts in-place at the same time.
 (Merge Sort is not in-place and Quicksort has a theoretical worst case of Θ(n²))
- Practically, not frequently used because it is slower than merge sort and quicksort.



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Introsort

From Wikipedia, the free encyclopedia

Introsort or introspective sort is a hybrid sorting algorithm that provides both fast average performance and (asymptotically) optimal worst-case performance. It begins with quicksort, it switches to heapsort when the recursion depth exceeds a level based

Introso	Introsort			
Class	Sorting algorithm			
Data structure	Array			
Worst-case performance	O(<i>n</i> log <i>n</i>)			
Average performance	O(<i>n</i> log <i>n</i>)			

on (the logarithm of) the number of elements being sorted and it switches to insertion sort when the number of elements is below some threshold. This combines the good parts of the three algorithms, with practical performance comparable to quicksort on typical data sets and worst-case $O(n \log n)$ runtime due to the heap sort. Since the three algorithms it uses are comparison sorts, it is also a comparison sort.



Sorting algorithms: summary

	inplace?	stable?	best	average	worst	remarks
selection	~		$\frac{1}{2} n^2$	$\frac{1}{2} n^2$	$\frac{1}{2} n^2$	n exchanges
insertion	~	~	п	$\frac{1}{4} n^2$	$\frac{1}{2} n^2$	use for small <i>n</i> or partially ordered
merge		~	$\frac{1}{2} n \log_2 n$	$n \log_2 n$	$n \log_2 n$	$\Theta(n \log n)$ guarantee; stable
timsort		~	п	$n \log_2 n$	$n \log_2 n$	improves mergesort when pre-existing order
quick	•		$n \log_2 n$	$2 n \ln n$	$\frac{1}{2} n^2$	$\Theta(n \log n)$ probabilistic guarantee; fastest in practice
3-way quick	~		n	$2 n \ln n$	$\frac{1}{2} n^2$	improves quicksort when duplicate keys
heap	~		3 n	$2 n \log_2 n$	$2 n \log_2 n$	$\Theta(n \log n)$ guarantee; in-place
?	~	~	n	$n \log_2 n$	$n \log_2 n$	holy sorting grail

number of compares to sort an array of n elements