

1	The Master Theorem: Q7.3
2	Binary Heap as a concrete DS for Priority Queues, Heap Sort, Q7.1
3	Quick Sort
4	Sorting Algorithms, Q 7.2 (extended) The k-smallest problem, Q 7.5, 7.6 Lower Bound for the Closest Pair Problem: Q7.4
LAB	Play with some sorting algorithm, especially different choices of pivot for qsort Using the provided code, understand how to do timing

The Master Theorem (for complexity of divide-and-conquer algorithms)

If
$$T(n) = aT\left(\frac{n}{b}\right) + Q(n^d)$$

$$T(1) = Q(1)$$

where $a \geq 1$, $b > 1$, then

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

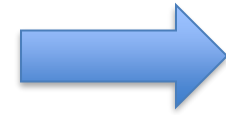
Note:

- Similar results hold for O and Ω
- Also OK if $T(1) = 0$

Q7.3: Find time complexity

$$T(n) = aT\left(\frac{n}{b}\right) + Q(n^d)$$

$$T(1) = Q(1)$$



$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

we can also compare $\log_b a$ & d

(a) $T(n) = 9T\left(\frac{n}{3}\right) + n^3, T(1) = 1$

(b) $T(n) = 64T\left(\frac{n}{4}\right) + n + \log n, T(1) = 1$

(c) $T(n) = 2T\left(\frac{n}{2}\right) + n, T(1) = 1$

(d) $T(n) = 2T\left(\frac{n}{2}\right), T(1) = 1$

(e) $T(n) = 2T(n-1) + 1, T(1)=1$

(a)

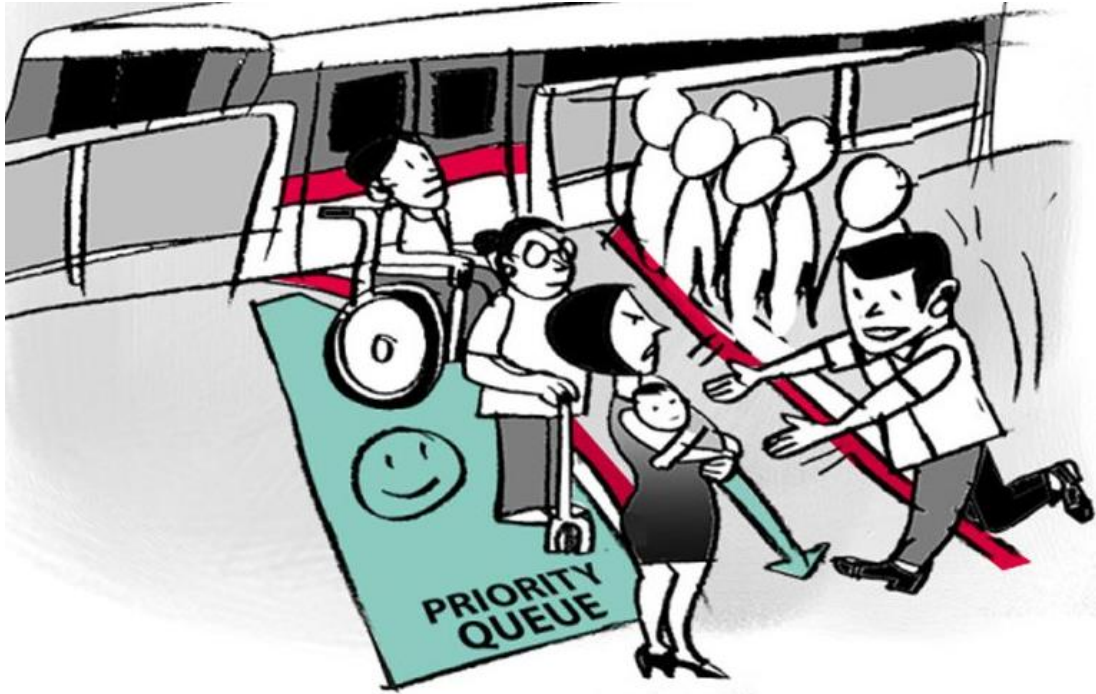
(b)

(c)

(d)

(e)

ADT Priority Queue and one of its implementation: Binary Heap

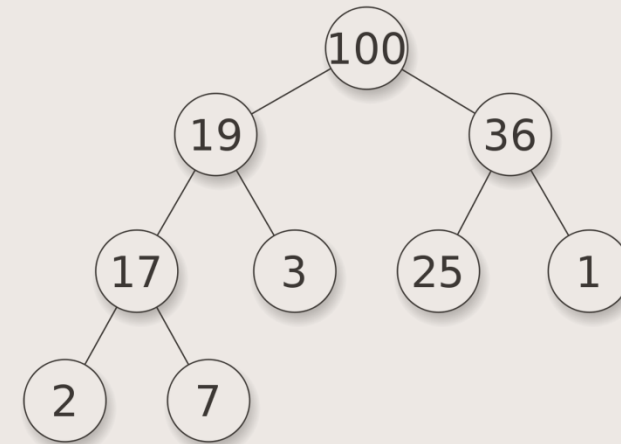


'Can I borrow your baby?...'

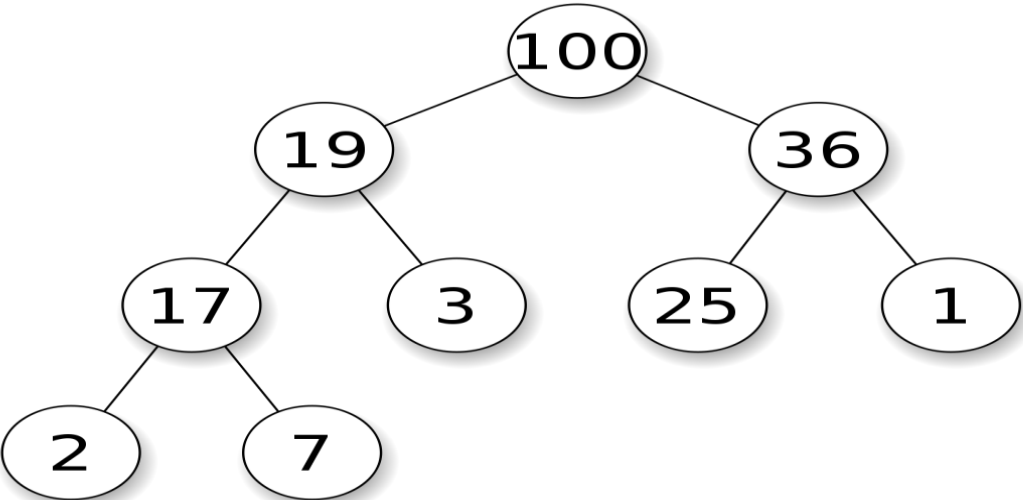
Binary Heap as a *concrete data type* (implementation) for PQ.

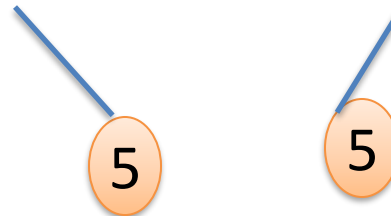
Simple priority: higher (max heap)
lower (min heap)

What is a, say, max heap?
How is it implemented?



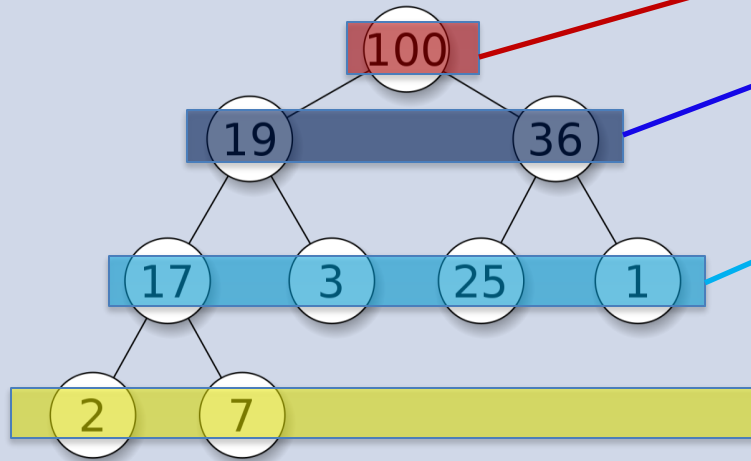
Binary Heap: conceptually, is a binary tree satisfying 2 conditions

Example	Conditions
 <pre>graph TD; 100((100)) --- 19((19)); 100 --- 36((36)); 19 --- 17((17)); 19 --- 3((3)); 17 --- 2((2)); 17 --- 7((7)); 36 --- 25((25)); 36 --- 1((1));</pre>	<ol style="list-style-type: none"><li data-bbox="1281 261 2407 435">1. The tree is <i>complete</i>:<ul style="list-style-type: none">all levels, except for the last, are fullthe last level is filled from left to right<li data-bbox="1281 575 2407 749">2. The <i>heap property</i>: each node has a higher priority (here, is larger) than any of its descendants (or equivalently, just its children).



Binary Heap: is implemented as an array!

Visualisation: as a complete binary tree



Implementation: using arrays

idx	0	1	2	3	4	5	6	7	8	9
val	×	100	19	36	17	3	25	1	2	7

note: for convenience:
 $H[1]$ is for the root
 $H[0]$ not used (or
used for a sentinel)

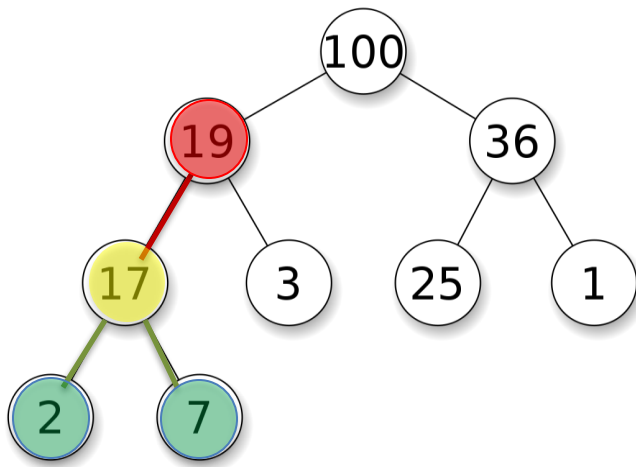
Heap is $H[1..n]$

- level i occupies 2^i cells in array $H[1..n]$
(except for the last level)
- if root is level 1, then level i starts from $H[2^{i-1}]$

Binary Heap: parent and children of a node

- left child of $H[i]$ is $H[2*i]$
- right child of $H[i]$ is $H[2*i+1]$

parent of $H[i]$ is $H[i/2]$



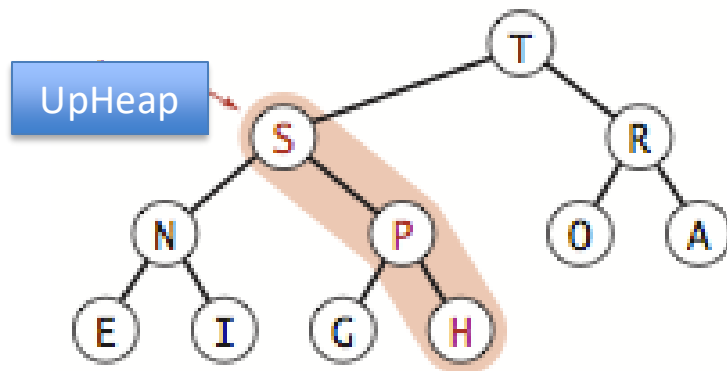
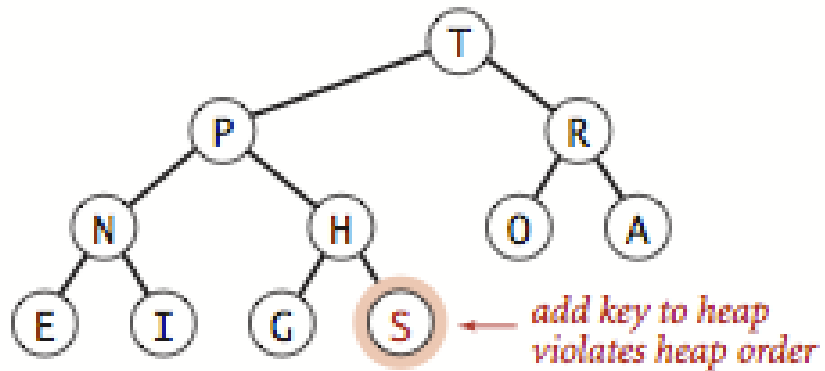
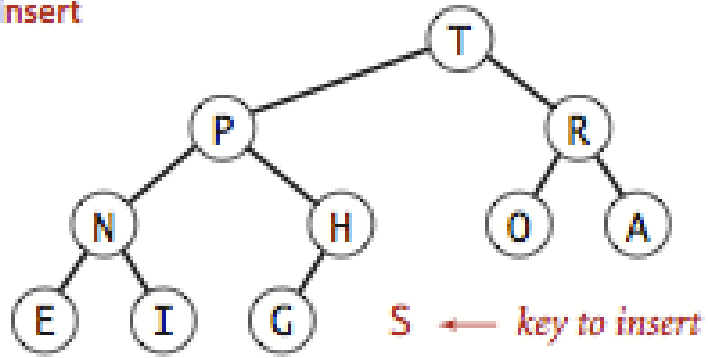
idx	0	1	2	3	4	5	6	7	8	9
val	×	100	19	36	17	3	25	1	2	7

Note:

- The use of indices 1..n is not a rule
- index 0..n-1 can also be used for heaps, but the formula for parent/children are not so nice

inject = enPQ = Insert a new elem into a heap using UpHeap

Insert



Notes:

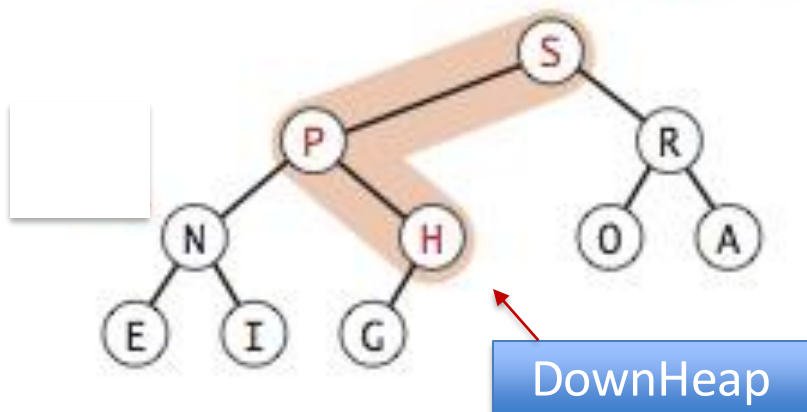
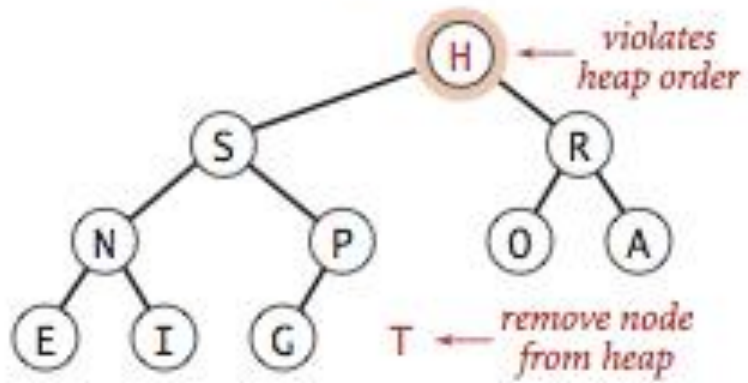
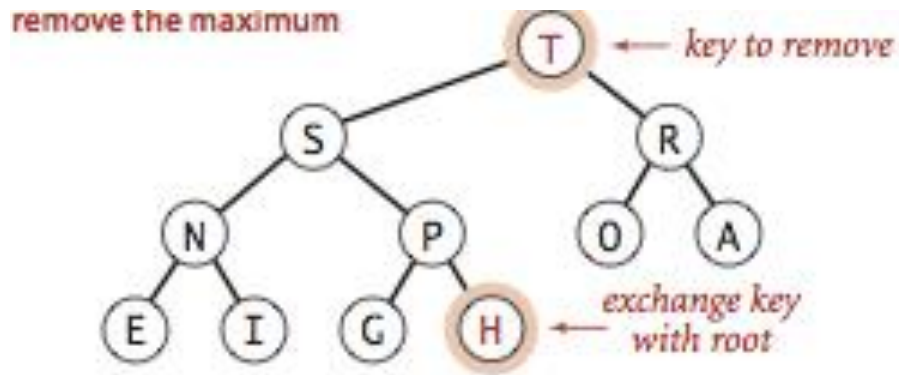
Procedure:

Complexity:

UpHeap (aka. SiftUp)

while (has parent and parent has lower priority):
 swap up with the parent

ject = delete (and returns the heaviest) and repair using DownHeap



Notes:
Procedure:

Complexity:

DownHeap (aka. SiftDown):

while (has children & the heavier child has higher priority):
swap down with the *heavier* child

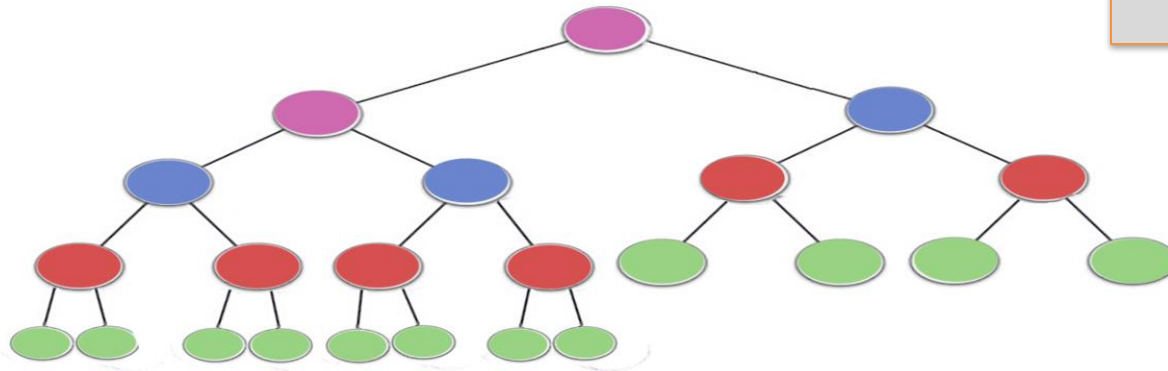
Heapify: Turning an array $H[1..n]$ into a heap

```
function Heapify( $H[1..n]$ )  
  for  $i \leftarrow n/2$  downto 1 do  
    downheap( $H, i$ )
```

= $\Theta(n)$ (see lectures and/or ask Google for a proof)

The operation is aka. **Heapify**/Makeheap/ Bottom-Up
Heap Construction

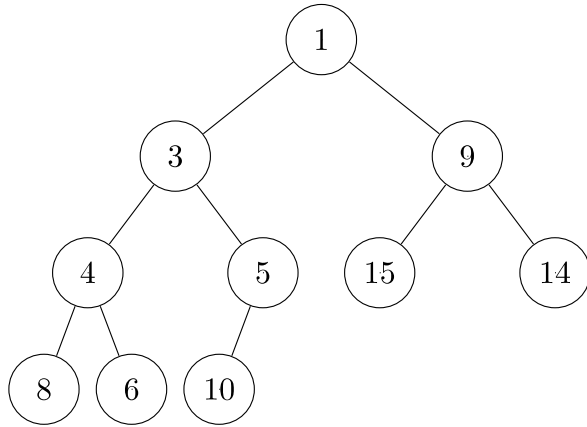
Example: build maxheap for keys **E X A M P**



Notes:

Complexity:

Q7.1



- a) Show how this heap would be stored in an array as discussed in lectures (root is at index 1; node at index i has children at indices $2i$ and $2i+1$)
- b) Run the `RemoveRootFromHeap` (eject) algorithm from lectures on this heap by hand
- c) Run the `InsertIntoHeap` (inject) algorithm and insert the value 2 into the heap

Your answers:

- a) array is: [???]
- b) Run the `RemoveRootFromHeap`:
- c) Run the `InsertIntoHeap(2)`:

Heap Sort

Describe an algorithm of using a heap for:

- Sorting an array in increasing order
- Sorting an array in decreasing order

Then:

- Do Complexity Analysis, use correct O or Θ
- Compare Heap Sort with Selection Sort
- Compare Heap Sort with fast sorting algorithms

Priority Queues: A Summary

A Priority Queue (PQ) is an Abstract Data Type (ADT) that enables the efficient removal or examination of the highest-priority element. Main operations:

- inject, or enPQ, or insert
- eject, or dePQ, or deleteMin (deleteMax)

Priority Queues are particularly useful in scenarios requiring:

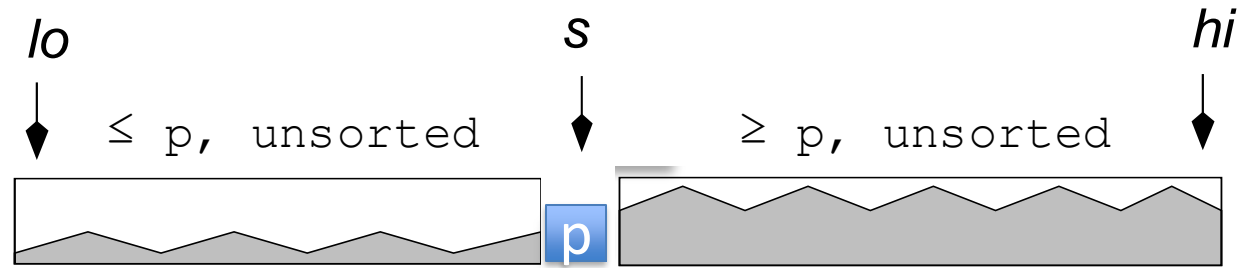
- Repeated removal of the smallest or largest element.
- Processing a set of elements in ascending or descending order of priority.

Priority Queues can be efficiently implemented using a Binary Heap, providing the following time complexities:

- $O(\log n)$ for **insertion**, deletion, and **priority change** operations.
- $\Theta(n)$ for **building** a priority queue from an unsorted collection

Quicksort idea (recursive, usage: `Quicksort(A[0..n-1])`)

```
function QuickSort(A[lo..hi])  
  if lo < hi then  
    s := Partition(A[lo..hi])  
    QuickSort(A[lo..s-1])  
    QuickSort(A[s+1..hi])
```



$p = A[s]$ is called the *pivot* of this partitioning

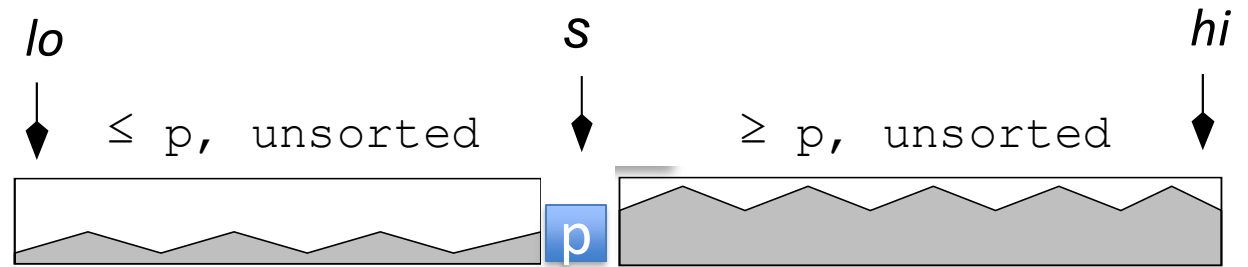
Note: a Partition of n elements has the complexity of $\Theta(n)$

Questions:

- What is the (additional) space complexity of Quicksort?
- What is the time complexity?
- Is it input-sensitive?
- Is it in-place?
- Is it stable?

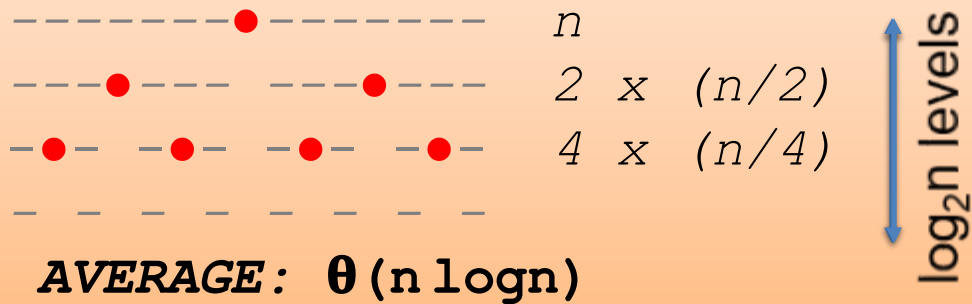
Quicksort Properties - Check your answers

```
function QuickSort(A[lo..hi])
  if lo < hi then
    s := Partition(A[lo..hi])
    QuickSort(A[lo..s-1])
    QuickSort(A[s+1..hi])
```

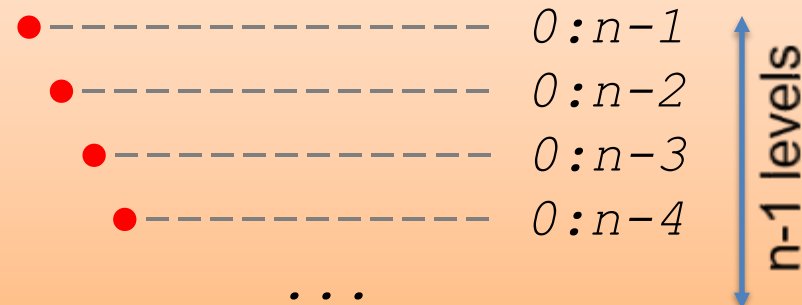


Quicksort complexity depends on the relative lengths of the left and the right parts in partitioning.

BEST: always balanced: $\Theta(n \log n)$



WORST: one half always empty: $\Theta(n^2)$



...

What is the (additional) space complexity of **Quicksort**? for recursion:
BEST/AVERAGE $O(\log n)$ WORST $O(n)$

- Is it input-sensitive? Y
- Is it in-place? Y
- Is it stable? N – but we need to check with Partitioning

loop invariant	l	i	j	h
	$A[l+1..i-1] \leq P$ $A[i..j]$ un-examined $A[j+1..h] \geq P$			

function Partition(A[l..h])

$i \leftarrow l; j \leftarrow h$

$P \leftarrow A[l]$ # loop init

repeat

move i forward until $A[i] > P$

while $i < h$ and $A[i] \leq P$ do $i \leftarrow i+1$

move j backward until $A[j] \leq P$

while $A[j] > P$ do $j \leftarrow j-1$

extend yellow and green area

at the same time by swapping ---->

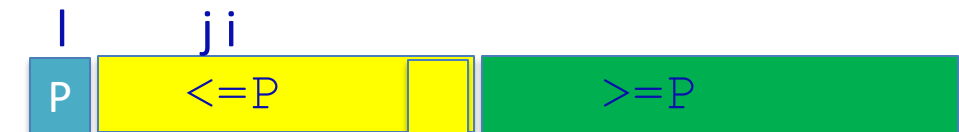
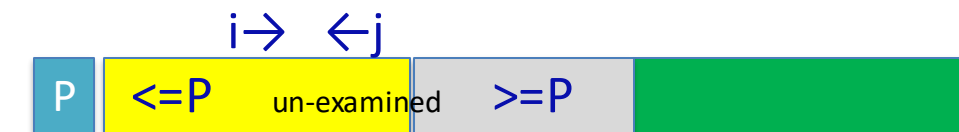
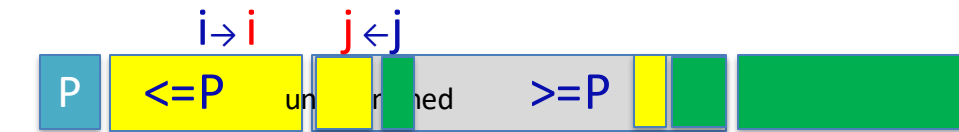
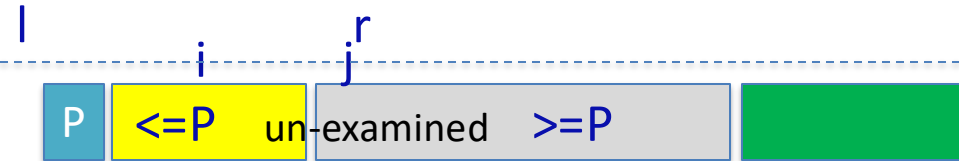
if ($i < j$) then Swap($A[i], A[j]$)

until $i \geq j$

at loop's exit: i and j crossed

Swap($A[l], A[j]$)

return j



<div data-bbox="25 49 152 107" data-label="Text"><p>Note</p></div>	<div data-bbox="242 21 2446 142" data-label="Text"><p>This slide shows that the algorithm presented here and the one in the lectures are basically the same. They are just one variation of implementation of Hoare's Partitioning</p></div> <div data-bbox="25 214 1401 1213" data-label="Code-Block"><pre>function Partition(A[l..h]) P ← A[l]; i ← l; j ← h repeat # move i forward until A[i]>P while i<h and A[i]≤P do i ← i+1 # move j backward until A[j]≤P while A[j]>P do j ← j-1 # extend yellow and green area if (i<j) then Swap(A[i], A[j]) until i ≥ j # at loop's exit: i and j crossed Swap(A[l], A[j]) return j</pre></div> <div data-bbox="1477 214 2522 1313" data-label="Code-Block"><pre>function PARTITION(A[lo..hi]) p ← A[lo]; i ← lo; j ← hi repeat while i < hi and A[i] ≤ p do i ← i + 1 while j ≥ lo and A[j] > p do j ← j - 1 swap(A[i], A[j]) until i ≥ j swap(A[i], A[j]) — undo the last swap swap(A[lo], A[j]) — bring pivot to its correct return j end function</pre></div>
--	--

```

function HoarePartition(A[l..h]) \
  P ← A[l]; i ← l+1; j ← h

  repeat
    # move i forward until A[i]>P
    while i<h and A[i]≤P do i ← i+1
    # move j backward until A[j]≤P
    while A[j]>P do j ← j-1
    # extend yellow and green area
    if (i<j) then Swap(A[i], A[j])
  until i ≥ j

  # at loop's exit: i and j crossed
  Swap(A[l], A[j])

  return j

```

Start with:

2_i 4 1 3_j

To simplify, we will write out the sequence only:

- at the beginning and end of the loop, and
- after a swap.

Question 7.1: Group Work

Know how to run by hand the following algorithms:

- a) Selection Sort
- b) Merge Sort
- c) Quick Sort with Hoare's Partitioning

For each algorithm: Run the algorithm on the array:

- [A N A L Y S I S]

Question 9.1

For each sorting algorithm:

- i. Run the algorithm on the following input array:
[A N A L Y S I S]
- ii. What is the time complexity of the algorithm?
- iii. Is the sorting algorithm stable?
- iv. Does the algorithm sort in-place?
- v. Is the algorithm input sensitive?

- skip stuffs that, after a short discussion, your group agrees that it's easy!
- be careful with letter ordering, perhaps write down:

A B C D E F G H I J K L M N O P Q R S T U V W X Y Z

Sorting Algorithms

Insertion Sort

Selection Sort

Heap Sort

Quick Sort

Merge Sort

Basic Idea

Complexity

Expected
Complexity

In-place?

Stable?

Input-
Sensitive?

The
Good

The Bad

Work with Your Classmates

As part of (extended) Question 7.2:

- fill in the table,
- imagine about scenarios when each sorting method would be the choice (e.g. better than other), that is
- be able to choose the best sorting algorithm for a particular application.

Also... Mergesort

Make sure you can run (by hand) Merge Sort

[4 1 3 5 2]

[A N A L Y S I S]

And, review the lectures for the remaining questions of problem 1. For each sorting algorithm, think:
which is the best situations when we want to employ that algorithm?
in which situations when we definitely don't want that algorithm?

Group Work on Algorithm Design: The k-th smallest problem

Problem: Given an array, find its k-th smallest value.

Task: Design and compare the complexity of multiple (say, 5) different algorithms.

A reasonable plan:

- Discuss and briefly describe algorithms with complexity analysis in class.
- Choose two distinct optimal methods for pseudocode implementation in class.
- Complete remaining work at home, if desired.

Note: Do **Not** read questions 7.5 and 7.6. This discussion covers these questions anyway.

Q7.4: Closest-pair and element-distinction

The element distinction problem takes as input a collection of n elements and determines whether or not all elements are distinct. It has been proved that if we disallow the usage of a hash table then the element distinction problem cannot be solved in less than $n \log n$ time (i.e., this class of problems is $\Omega(n \log n)$).

- Describe how we could use the closest pair algorithm from class to solve the element distinction problem (where the input is a collection of floating point numbers), and hence

```
function ElemDistinction(?)
```

- explain why this proves that the closest pair problem must not be able to be solved in less than $n \log n$ time (and is thus $\Omega(n \log n)$).

```
?
```


- Play with some sorting algorithm, especially different choices of pivot for qsort
- Discuss: What's the best method of choosing qsort?
- Using the provided code, understand how to do timing in C

CHECK: qsort with Hoare's Partitioning

Q9.1: Run quicksort with Hoare's for [A N A L Y S I S] .

Partitioning of
A N A L Y S I S

A_i N A L Y S S_j
A N_i A_j L Y S I S
A A_j N_i L Y S I S

[A] A_j [N L Y S I S]

Partitioning of
N L Y S I S

N_i L Y S I S_j
N L I_j S_i Y S
N L I_j S_i Y S

[I L] N_j [S Y S]

Partitioning of
I L

I_i L_j
 I_j L_i
 I_j [L]

Partitioning of
S Y S

S_i Y S_j
S Y_i S_j
S S_j Y_i

[S] S_j [Y]

Final sorted sequence
A I L N S S Y