

# Search Algorithms



# Linear Search

**Key idea:** search linearly through array from front to back to find item

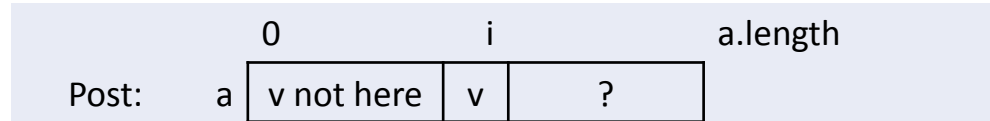
```
/** Returns: the smallest index i such that a[i] == v.  
    Requires: v is in a. */  
int linear_search(int[] a, int v) {  
    int i = 0;  
  
    while (a[i] != v) i++;  
    return i;  
}
```

# Exercise

**State** the loop invariant.

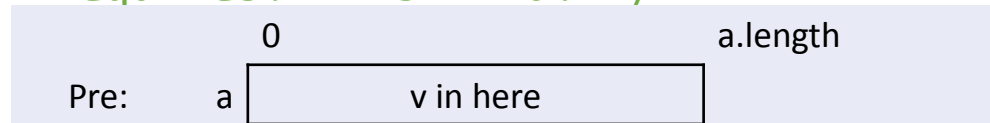
```
/** Returns: the smallest index i such that a[i] == v.  
    Requires: v is in a. */  
int linear_search(int[] a, int v) {  
    int i = 0;  
    // inv: TODO  
    while (a[i] != v) i++;  
    return i;  
}
```

# Discovering the loop invariant

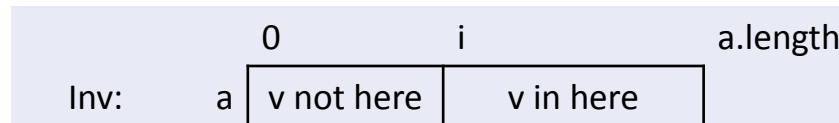


Post:  
v not in a[0..i)  
a[i] == v

*/\*\* Returns: the smallest index i such that a[i] == v.  
Requires: v is in a. \*/*



Pre:  
v in a[0..]



Rule: We never draw indices directly above a line in diagram! Always to left or right.

Inv:  
v not in a[0..i)  
v in a[i..] *in math*

*in array diagrams*

# Discovering the loop invariant

```
/** Returns: the smallest index i such that a[i] == v.  
    Requires: v is in a. */
```

Discovering an invariant from the pre and post conditions requires creativity and practice.

**Theorem.** There is no algorithm that can do it for you.  
**Corollary:** ChatGPT can't replace human programmers yet!

```
Inv:  
v not in  
a[0..i)  
v in a[i..]
```

# Linear Search: with invariant

```
/** Returns: the smallest index i such that a[i] == v.  
    Requires: v is in a. */  
int linear_search(int[] a, int v) {  
    int i = 0;  
    // inv: v not in a[0..i), and v in a[i..]  
    while (a[i] != v) i++;  
    return i;  
}
```

# Linear Search: loop checklist

Does it start right?

Does it maintain the invariant?

Does it end right?

Does it make progress?

```
/** Returns: the smallest index
    i such that a[i] == v.
    Requires: v is in a. */
int linear_search(int[] a, int v) {
    int i = 0;
    // inv: v not in a[0..i), and v in a[i..]
    while (a[i] != v) i++;
    return i;
}
```

# Binary Search

**Key idea:** maintain upper and lower bounds on where value could be.

```
/** Returns: an index i such that a[i] == v.  
    Requires: v is in a, and a is sorted in ascending order. */  
int bin_search(int[] a, int v) {  
    int l = 0;  
    int r = a.length - 1;  
    // inv: 0 <= l <= r < a.length, and v in a[l..r]  
    while (l != r) {  
        int m = (l + r) / 2;  
        if (v <= a[m]) { r = m; }  
        else { l = m + 1; }  
    }  
    return l;  
}
```





# Binary Search: loop checklist

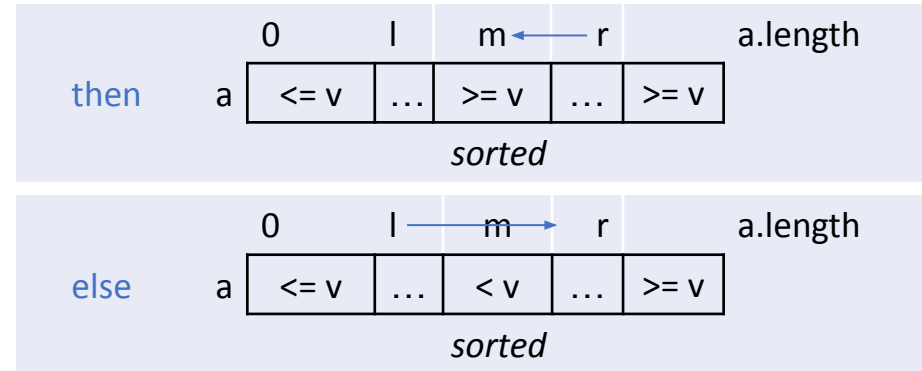
```
/** Returns: an index i such that a[i] == v.  
    Requires: v is in a, and a is sorted in ascending order.  
*/  
int bin_search(int[] a, int v) {  
    int l = 0;  
    int r = a.length - 1;  
    // inv: 0 <= l <= r < a.length, and v in a[l..r]  
    while (l != r) {  
        int m = (l + r) / 2;  
        if (v <= a[m]) { r = m; }  
        else { l = m + 1; }  
    }  
    return l;  
}
```

Does it start right?

Does it maintain the invariant?

Does it end right?

Does it make progress?



The background consists of numerous dark gray, glossy, ribbon-like shapes that swirl and curve across the frame, creating a sense of motion and depth. The ribbons vary in thickness and are layered, with some appearing to wrap around others. The overall effect is a complex, organic pattern that resembles a tangled mass of fibers or a dynamic, swirling vortex.

Loops (in)variants  
are your friends.

# CS 2110

## Lecture 12?

### Sorting

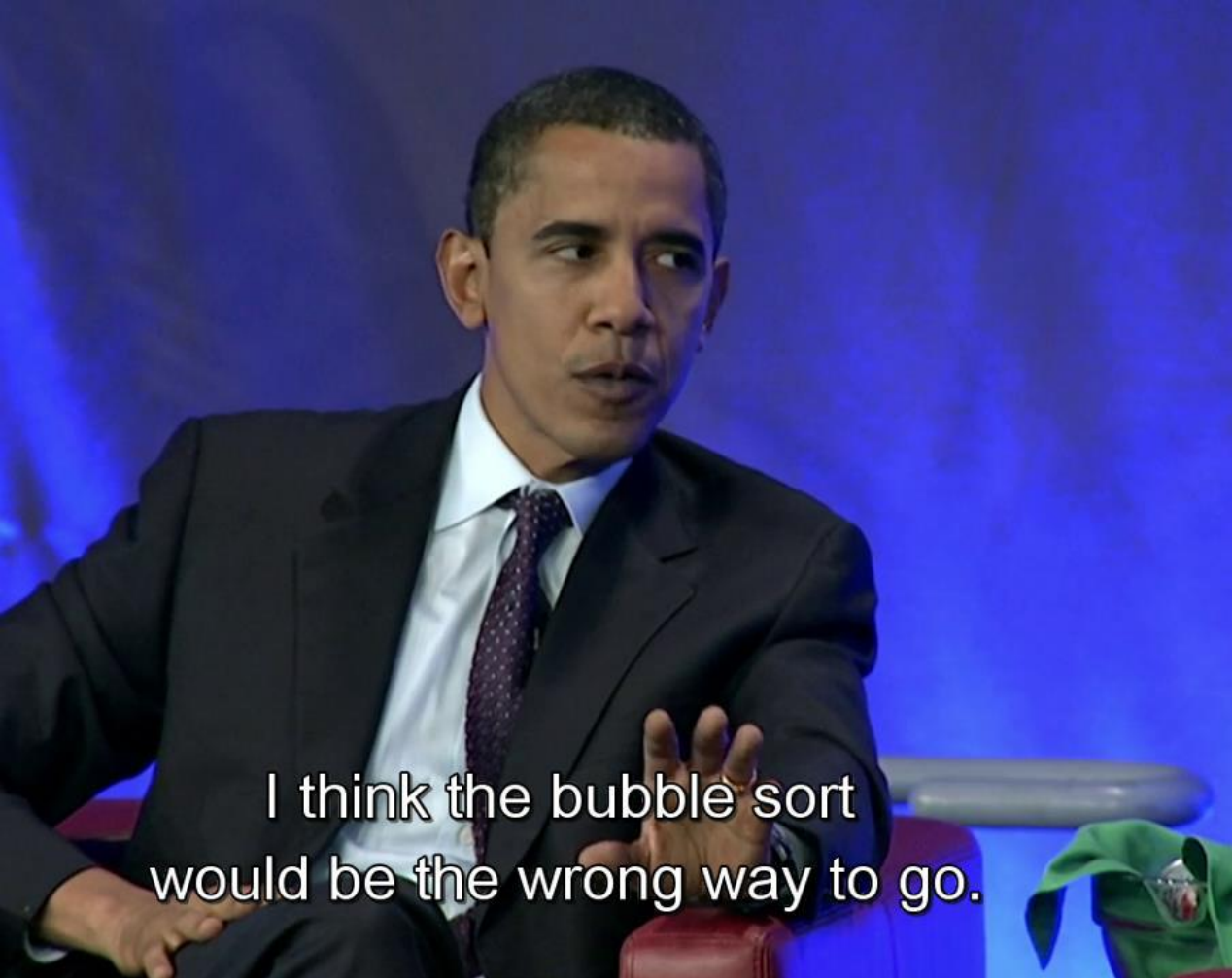
- Selection sort
- Insertion sort
- Merge sort
- Quicksort



# Why sort things?

- Makes looking things up faster
  - Binary search
- Compute robust statistics
  - Median, quantiles
  - Top-10 lists
- Prioritize/optimize
  - Search results
  - Drawing order





I think the bubble sort  
would be the wrong way to go.

## Why multiple algorithms?

- Tradeoffs: no one “best” algorithm
  - Speed
  - Memory
  - Expected vs. worst case
  - Stability
  - R/W locality
- You will be responsible for choosing appropriate methods



# Setting: arrays

- Why arrays?
  - Data to be sorted is often in an array (or ArrayList)
  - Arrays are familiar
  - Good opportunity to visualize loop invariants with array diagrams
- Implications
  - Fast to read/write arbitrary locations, iterate in reverse
  - Swaps are cheap
  - Insertions are expensive
- Most algorithms generalize to linked structures

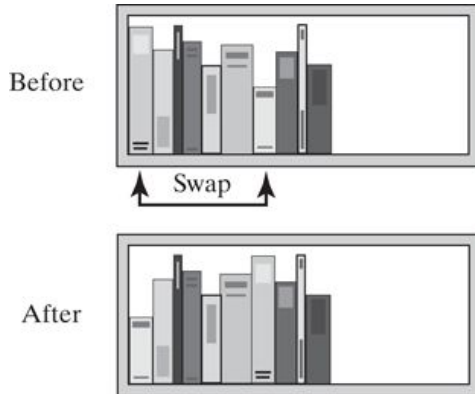
# Selection sort





# Analogy: bookshelf

- Find the shortest remaining (unsorted) book
- Move it just after all the already sorted (and shorter) books



- How to “move” it?
  - Push subsequent books out of the way
    - Difficult; analogous to insertion
  - Trade places with book in desired position
    - Easy; analogous to swapping

# Selection sort example

<b>i=0</b>	<b>3</b>	<b>1</b>	<b>4</b>	<b>1</b>	<b>5</b>
<b>i=1</b>					
<b>i=2</b>					
<b>i=3</b>					
<b>i=4</b>					

# Selection sort invariant

Pre a 

$\emptyset$	$h$
?	

 a.length  
:

Inv a 

$\emptyset$	$i$	$a.length$
sorted	$\geq a[0..i)$	

 h  
:

Post a 

$\emptyset$	$h$
sorted	

 a.length  
:

# Selection sort code

```
// Invariant: a[0..i) is sorted, a[i..] >= a[0..i)
int i = 0;
while (i < a.length - 1) {
    // Find index of smallest element in a[i..]
    int jSmallest = i;
    for (int j = i + 1; j < a.length; ++j) {
        if (a[j] < a[jSmallest]) {
            jSmallest = j;
        }
    }
    // Swap smallest element to extend sorted portion
    swap(a, i, jSmallest);
    i += 1;
}
```

- Time complexity analysis ( $N = a.length$ )
  - $i=0$ :  $N-1$  comparisons
  - $i=1$ :  $N-2$  comparisons
  - $i=2$ :  $N-3$  comparisons
  - ...
  - $i=N-2$ : 1 comparison
- Total comparisons:  
 $1 + 2 + \dots + (N-1)$

$$O(N^2)$$

# Algorithm properties

Algorithm	Best case time complexity	Worst case time complexity	Space complexity
Selection sort			



# Insertion sort

# Analogy: a hand of playing cards

- Left hand holds cards that have already been sorted
- Take next card from right hand, insert it where it belongs in left hand
- How to “insert”
  - Push all bigger cards out of the way
  - Swap with cards to left until in position



# Insertion sort example

<b>i=0</b>	<b>3</b>	<b>1</b>	<b>4</b>	<b>1</b>	<b>5</b>
<b>i=1</b>					
<b>i=2</b>					
<b>i=3</b>					
<b>i=4</b>					





# Insertion sort code

```
// Invariant: a[0..i] is sorted
int i = 0;
while (i < a.length) {
    // Slide a[i] to its sorted position in a[0..i]
    // Invariant: a[j] < a[j+1..i]
    int j = i;
    while (j > 0 && a[j - 1] > a[j]) {
        swap(a, j - 1, j);
        j -= 1;
    }
    i += 1;
}
```

- Time complexity analysis ( $N = a.length$ )
  - $i=1$ : 1 comparison
  - $i=2$ : < 2 comparisons
  - $i=3$ : < 3 comparisons
  - ...
  - $i=N-1$ : <  $N-1$  comparisons
- Total comparisons (worst-case):  
 $1 + 2 + \dots + (N-1)$

$$O(N^2)$$

# Poll: Complexity if array is already sorted?

- How many comparisons does **Insertion Sort** evaluate if the array is already sorted?

( $N$  is the number of elements in the array)

- A.  $O(1)$
- B.  $O(\log N)$
- C.  $O(N)$
- D.  $O(N^2)$



# Insertion sort extras

- What if array is already sorted?
  - Each “insert” requires only 1 comparison
  - Overall complexity (best case) is  $\Omega(N)$
- Fast in practice for *small N*
  - Often used as a “base case” in implementations of other algs
- What if there are duplicates?
  - E.g. sorting Students by last name
  - **Stable**: relative order of equal elements is preserved
  - Insertion sort is stable because elements only move right-to-left and stop when they hit a duplicate
  - Selection sort is *not* stable because long-range swaps can change order

# Algorithm properties

Algorithm	Best case time complexity	Worst case time complexity	Space complexity	Stability
Selection sort				Unstable
Insertion sort				Stable



Merge sort



# Merging sorted subarrays

- Given two sorted sequences, how hard is it to merge them?
  - Easy! Repeatedly take the smaller of what's left of the two sequences
  - Complexity:  $O(N)$  – easier than sorting (but requires  $O(N)$  scratch space)
- What if, when tasked to sort, you outsourced the job to *two* assistants, who each sorted half of the list
  - Their jobs are easier (maybe much easier), since their lists are smaller
  - Your job is easier, since you only have to merge
- What if your assistants outsourced their tasks...?

# Divide and conquer

- Divide task into *multiple* smaller subtasks, then assemble results into solution
- Natural fit for recursion

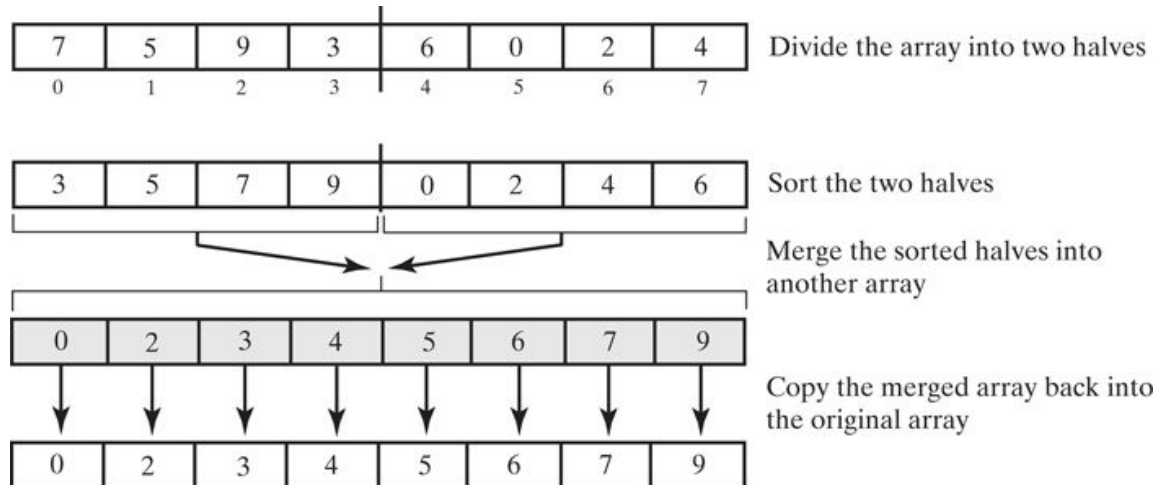


# Merge example

1	3	4	1	5	9

# Merge sort (high level)

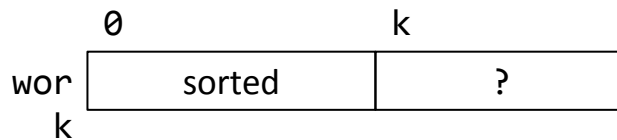
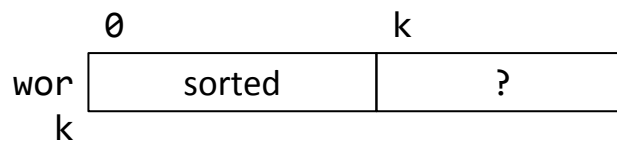
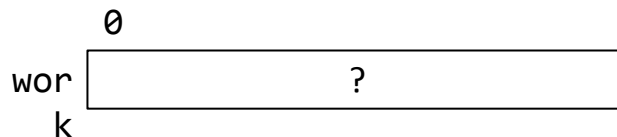
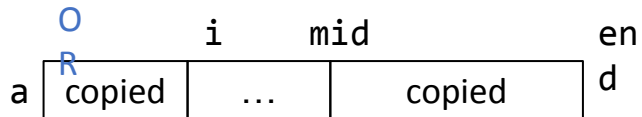
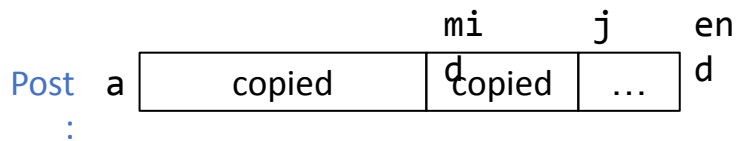
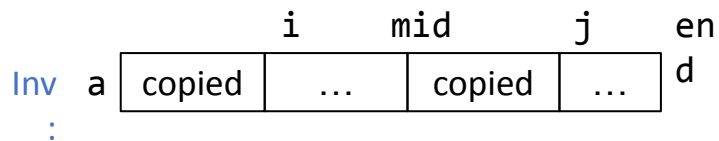
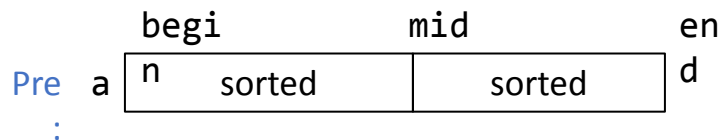
1. Sort left half of array (using merge sort)
2. Sort right half of array (using merge sort)
3. Merge left and right subarrays



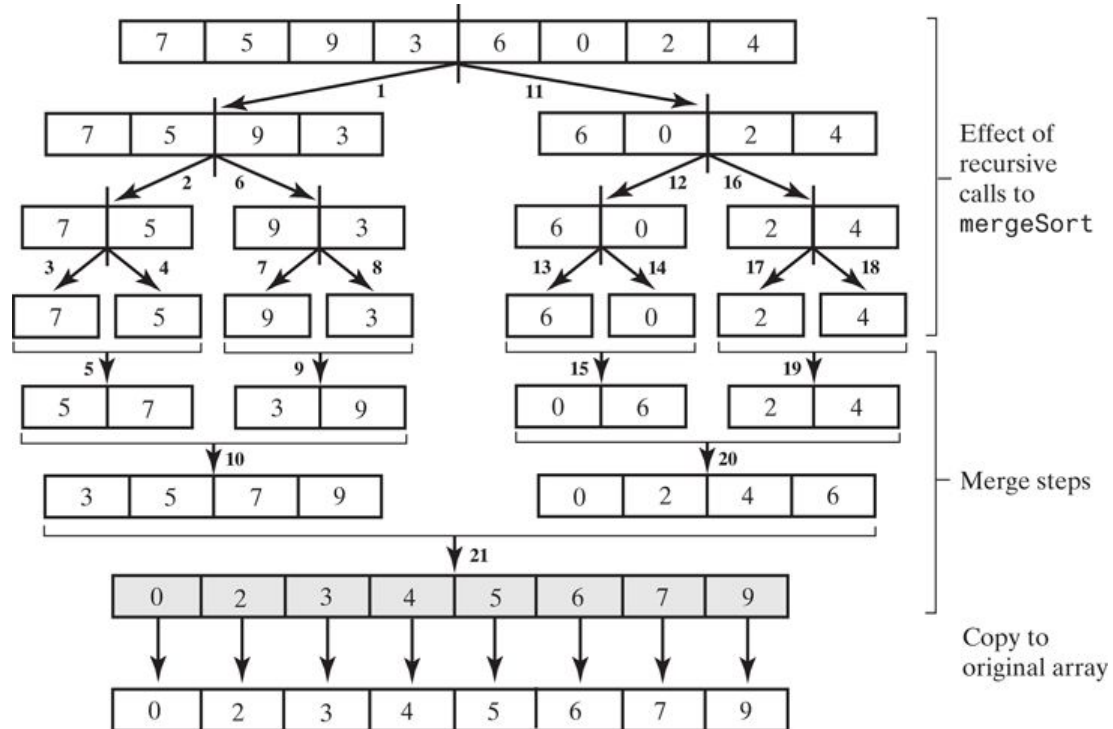
# Merge sort code

- Demo

# Merge invariant



# Analysis



# Algorithm properties

Algorithm	Best case time complexity	Worst case time complexity	Space complexity	Stability
Selection sort				Unstable
Insertion sort				Stable
Merge sort				Stable

# Merge sort in practice

- Usually the go-to stable sort (default in many language libraries)
- Since merging is always left-to-right, can be performed on data that does not fit in RAM



# Quicksort





# Quicksort on one slide

$$\text{sort}(a) = [ \text{sort}(a[a < p]), p, \text{sort}(a[a \geq p]) ]$$

1. Partition array about a “pivot”
2. Sort the subarray of values less than the pivot
3. Sort the subarray of values greater than the pivot

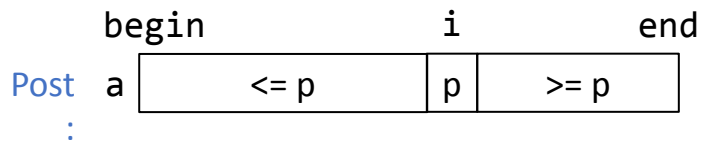
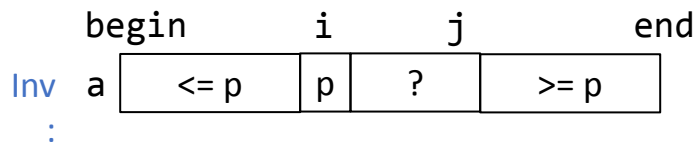
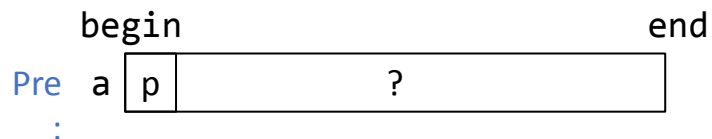
Sort via repeated partitioning

- How efficient is partitioning?
- How many times will you need to partition?

# Partition example

3	1	4	1	5	9

# Partition invariant



# Quicksort code

- Demo

# Analysis

## Best case

- Pivot is median value
- Each subarray is less than half the size of the original
- Depth of recursion:  $O(\log N)$   
Cost of partitioning one level:  $O(N)$
- Overall complexity:  $\Omega(N \log N)$

## Worst case

- Pivot is smallest (or largest) value
- One subarray is only 1 element shorter than original array
- Depth of recursion:  $O(N)$   
Cost of partitioning one level:  $O(N)$
- Overall complexity:  $O(N^2)$

# Choice of pivot

- Using first value is a bad choice!
  - In practice, many arrays are partially sorted
- Computing true median is not cost-effective
- Common heuristic: `med3(a[begin], a[mid], a[end-1])`
- Consequences of a bad pivot can be severe!
  - “Complexity attacks” to deny service

# Algorithm properties

Algorithm	Best case time complexity	Worst case time complexity	Space complexity	Stability
Selection sort				Unstable
Insertion sort				Stable
Merge sort				Stable
Quicksort				Unstable

\* Naïve implementation requires  $O(N)$  worst-case space, but can use tail recursion to reduce to  $O(\log N)$ .

# Quicksort in practice

- Despite poor worst-case complexity, Quicksort is often the fastest sort in practice (default unstable sort in many language libraries)
- Often augmented to detect and avoid worst-case behavior (e.g. fall back to heap sort)