



Lecture 5

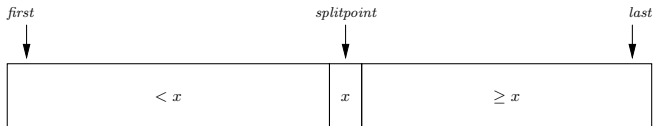
Av. case analysis of quick sort,
divide and conquer,
mergesort

CS 161 Design and Analysis of Algorithms

Ioannis Panageas

Pseudocode for Quicksort

```
def quickSort(A,first,last):  
    if first < last:  
        splitpoint = split(A,first,last)  
        quickSort(A,first,splitpoint-1)  
        quickSort(A,splitpoint+1,last)
```



The split step

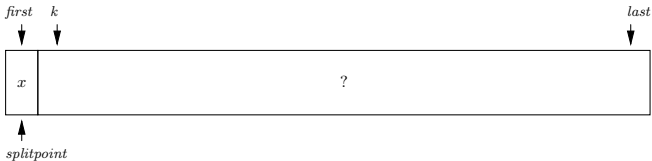
```
def split(A,first,last):  
    splitpoint = first  
    x = A[first]  
    for k = first+1 to last do:  
        if A[k] < x:  
            A[splitpoint+1] ↔ A[k]  
            splitpoint = splitpoint + 1  
    A[first] ↔ A[splitpoint]  
    return splitpoint
```

Loop invariants:

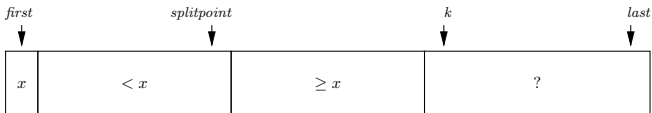
- ▶ $A[\text{first}+1..\text{splitpoint}]$ contains keys $< x$.
- ▶ $A[\text{splitpoint}+1..k-1]$ contains keys $\geq x$.
- ▶ $A[k..\text{last}]$ contains unprocessed keys.

The split step

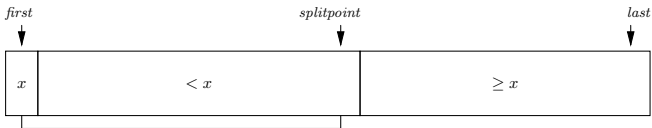
At start:



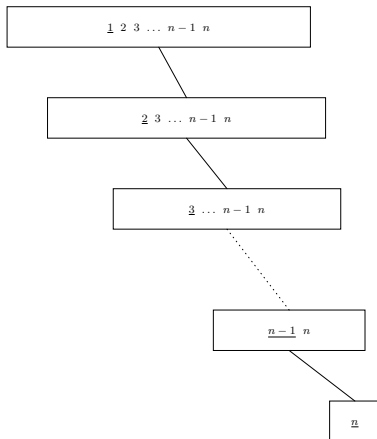
In middle:



At end:

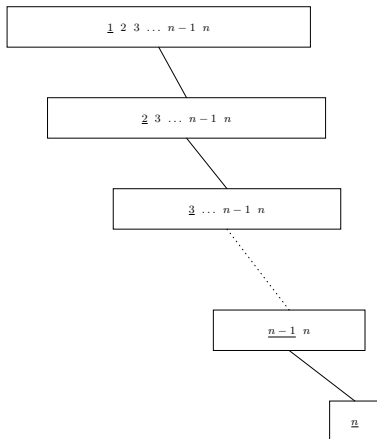


A bad case case for Quicksort: $1, 2, 3, \dots, n - 1, n$



$\binom{n}{2}$ comparisons required. So the **worst-case** running time for Quicksort is $\Theta(n^2)$.

A bad case case for Quicksort: $1, 2, 3, \dots, n-1, n$



$\binom{n}{2}$ comparisons required. So the **worst-case** running time for Quicksort is $\Theta(n^2)$. But what about the **average case** ... ?

Average-case analysis of Quicksort:

Our approach:

Average-case analysis of Quicksort:

Our approach:

1. Use the **binary tree of sorted lists**

Average-case analysis of Quicksort:

Our approach:

1. Use the **binary tree of sorted lists**
2. Number the items in **sorted order**

Average-case analysis of Quicksort:

Our approach:

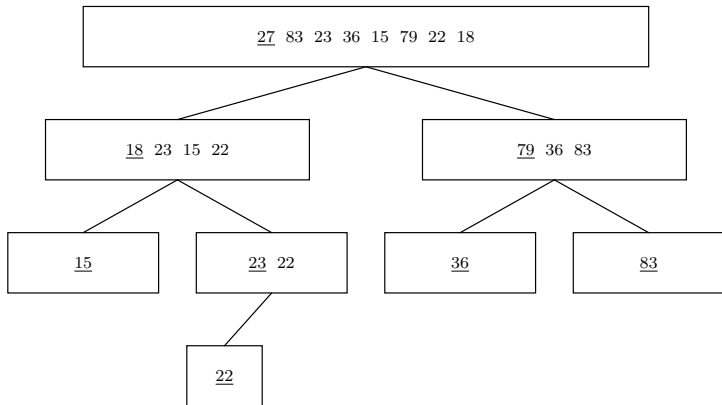
1. Use the **binary tree of sorted lists**
2. Number the items in **sorted order**
3. Calculate the **probability that two items get compared**

Average-case analysis of Quicksort:

Our approach:

1. Use the **binary tree of sorted lists**
2. Number the items in **sorted order**
3. Calculate the **probability that two items get compared**
4. Use this to compute the **expected number of comparisons** performed by Quicksort.

Average-case analysis of Quicksort:



Sorted order: 15 18 22 23 27 36 79 83

Average-case analysis of Quicksort

Average-case analysis of Quicksort

- ▶ Number the keys in sorted order: $S_1 < S_2 < \cdots < S_n$.

Average-case analysis of Quicksort

- ▶ Number the keys in sorted order: $S_1 < S_2 < \dots < S_n$.
- ▶ **Fact about comparisons:** During the run of Quicksort, two keys S_i and S_j get compared **if and only if** the first key from the set of keys $\{S_i, S_{i+1}, \dots, S_j\}$ to be chosen as a pivot is either S_i or S_j .

Average-case analysis of Quicksort

- ▶ Number the keys in sorted order: $S_1 < S_2 < \dots < S_n$.
- ▶ **Fact about comparisons:** During the run of Quicksort, two keys S_i and S_j get compared **if and only if** the first key from the set of keys $\{S_i, S_{i+1}, \dots, S_j\}$ to be chosen as a pivot is either S_i or S_j .
 - ▶ If some key S_k is chosen first with $S_i < S_k < S_j$, then S_i goes in the left half, S_j goes in the right half, and S_i and S_j never get compared.

Average-case analysis of Quicksort

- ▶ Number the keys in sorted order: $S_1 < S_2 < \dots < S_n$.
- ▶ **Fact about comparisons:** During the run of Quicksort, two keys S_i and S_j get compared **if and only if** the first key from the set of keys $\{S_i, S_{i+1}, \dots, S_j\}$ to be chosen as a pivot is either S_i or S_j .
 - ▶ If some key S_k is chosen first with $S_i < S_k < S_j$, then S_i goes in the left half, S_j goes in the right half, and S_i and S_j never get compared.
 - ▶ If S_i is chosen first, it is compared against all the other keys in the set in the split step (including S_j).

Average-case analysis of Quicksort

- ▶ Number the keys in sorted order: $S_1 < S_2 < \dots < S_n$.
- ▶ **Fact about comparisons:** During the run of Quicksort, two keys S_i and S_j get compared **if and only if** the first key from the set of keys $\{S_i, S_{i+1}, \dots, S_j\}$ to be chosen as a pivot is either S_i or S_j .
 - ▶ If some key S_k is chosen first with $S_i < S_k < S_j$, then S_i goes in the left half, S_j goes in the right half, and S_i and S_j never get compared.
 - ▶ If S_i is chosen first, it is compared against all the other keys in the set in the split step (including S_j).
 - ▶ Similar if S_j is chosen first.

Average-case analysis of Quicksort

- ▶ Number the keys in sorted order: $S_1 < S_2 < \dots < S_n$.
- ▶ **Fact about comparisons:** During the run of Quicksort, two keys S_i and S_j get compared **if and only if** the first key from the set of keys $\{S_i, S_{i+1}, \dots, S_j\}$ to be chosen as a pivot is either S_i or S_j .
 - ▶ If some key S_k is chosen first with $S_i < S_k < S_j$, then S_i goes in the left half, S_j goes in the right half, and S_i and S_j never get compared.
 - ▶ If S_i is chosen first, it is compared against all the other keys in the set in the split step (including S_j).
 - ▶ Similar if S_j is chosen first.

Examples:

Average-case analysis of Quicksort

- ▶ Number the keys in sorted order: $S_1 < S_2 < \dots < S_n$.
- ▶ **Fact about comparisons:** During the run of Quicksort, two keys S_i and S_j get compared **if and only if** the first key from the set of keys $\{S_i, S_{i+1}, \dots, S_j\}$ to be chosen as a pivot is either S_i or S_j .
 - ▶ If some key S_k is chosen first with $S_i < S_k < S_j$, then S_i goes in the left half, S_j goes in the right half, and S_i and S_j never get compared.
 - ▶ If S_i is chosen first, it is compared against all the other keys in the set in the split step (including S_j).
 - ▶ Similar if S_j is chosen first.

Examples:

- ▶ 23 and 22 (both statements true)

Average-case analysis of Quicksort

- ▶ Number the keys in sorted order: $S_1 < S_2 < \dots < S_n$.
- ▶ **Fact about comparisons:** During the run of Quicksort, two keys S_i and S_j get compared **if and only if** the first key from the set of keys $\{S_i, S_{i+1}, \dots, S_j\}$ to be chosen as a pivot is either S_i or S_j .
 - ▶ If some key S_k is chosen first with $S_i < S_k < S_j$, then S_i goes in the left half, S_j goes in the right half, and S_i and S_j never get compared.
 - ▶ If S_i is chosen first, it is compared against all the other keys in the set in the split step (including S_j).
 - ▶ Similar if S_j is chosen first.

Examples:

- ▶ 23 and 22 (both statements true)
- ▶ 36 and 83 (both statements false)

Average-case analysis of Quicksort

Average-case analysis of Quicksort

Assume:

Average-case analysis of Quicksort

Assume:

- ▶ All n keys are distinct

Average-case analysis of Quicksort

Assume:

- ▶ All n keys are distinct
- ▶ All permutations are equally likely

Average-case analysis of Quicksort

Assume:

- ▶ All n keys are distinct
- ▶ All permutations are equally likely
- ▶ The keys in **sorted order** are $S_1 < S_2 < \dots < S_n$.

Average-case analysis of Quicksort

Assume:

- ▶ All n keys are distinct
- ▶ All permutations are equally likely
- ▶ The keys in **sorted order** are $S_1 < S_2 < \dots < S_n$.

Let $P_{i,j}$ = The probability that keys S_i and S_j are compared with each other during the invocation of quicksort

Average-case analysis of Quicksort

Assume:

- ▶ All n keys are distinct
- ▶ All permutations are equally likely
- ▶ The keys in **sorted order** are $S_1 < S_2 < \dots < S_n$.

Let $P_{i,j}$ = The probability that keys S_i and S_j are compared with each other during the invocation of quicksort

Then by **Fact about comparisons** on previous slide:

$P_{i,j}$ = The probability that the first key from $\{S_i, S_{i+1}, \dots, S_j\}$ to be chosen as a pivot value is either S_i or S_j

Average-case analysis of Quicksort

Assume:

- ▶ All n keys are distinct
- ▶ All permutations are equally likely
- ▶ The keys in **sorted order** are $S_1 < S_2 < \dots < S_n$.

Let $P_{i,j}$ = The probability that keys S_i and S_j are compared with each other during the invocation of quicksort

Then by **Fact about comparisons** on previous slide:

$$\begin{aligned}
 P_{i,j} &= \text{The probability that the first key from} \\
 &\quad \{S_i, S_{i+1}, \dots, S_j\} \text{ to be chosen as a pivot value is} \\
 &\quad \text{either } S_i \text{ or } S_j \\
 &= \frac{2}{j - i + 1}
 \end{aligned}$$

Average-case analysis of Quicksort

Define indicator random variables $\{X_{i,j} : 1 \leq i < j \leq n\}$

$$X_{i,j} = \begin{cases} 1 & \text{if keys } S_i \text{ and } S_j \text{ get compared} \\ 0 & \text{if keys } S_i \text{ and } S_j \text{ do not get compared} \end{cases}$$

Average-case analysis of Quicksort

Define indicator random variables $\{X_{i,j} : 1 \leq i < j \leq n\}$

$$X_{i,j} = \begin{cases} 1 & \text{if keys } S_i \text{ and } S_j \text{ get compared} \\ 0 & \text{if keys } S_i \text{ and } S_j \text{ do not get compared} \end{cases}$$

1. The total number of comparisons is:

$$\sum_{i=1}^n \sum_{j=i+1}^n X_{i,j}$$

Average-case analysis of Quicksort

Define indicator random variables $\{X_{i,j} : 1 \leq i < j \leq n\}$

$$X_{i,j} = \begin{cases} 1 & \text{if keys } S_i \text{ and } S_j \text{ get compared} \\ 0 & \text{if keys } S_i \text{ and } S_j \text{ do not get compared} \end{cases}$$

1. The total number of comparisons is:

$$\sum_{i=1}^n \sum_{j=i+1}^n X_{i,j}$$

2. The expected (average) total number of comparisons is:

$$E \left(\sum_{i=1}^n \sum_{j=i+1}^n X_{i,j} \right) = \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j})$$

Average-case analysis of Quicksort

Define indicator random variables $\{X_{i,j} : 1 \leq i < j \leq n\}$

$$X_{i,j} = \begin{cases} 1 & \text{if keys } S_i \text{ and } S_j \text{ get compared} \\ 0 & \text{if keys } S_i \text{ and } S_j \text{ do not get compared} \end{cases}$$

1. The total number of comparisons is:

$$\sum_{i=1}^n \sum_{j=i+1}^n X_{i,j}$$

2. The expected (average) total number of comparisons is:

$$E \left(\sum_{i=1}^n \sum_{j=i+1}^n X_{i,j} \right) = \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j})$$

3. The expected value of $X_{i,j}$ is:

$$E(X_{i,j}) = P_{i,j} = \frac{2}{j-i+1}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j})$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) = \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\begin{aligned} \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\ &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \quad (k=j-i+1) \end{aligned}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\begin{aligned}
 \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\
 &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \quad (k=j-i+1) \\
 &< \sum_{i=1}^n \sum_{k=1}^n \frac{2}{k}
 \end{aligned}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\begin{aligned}
 \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\
 &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \quad (k=j-i+1) \\
 &< \sum_{i=1}^n \sum_{k=1}^n \frac{2}{k} \\
 &= 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k}
 \end{aligned}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\begin{aligned}
 \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\
 &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \quad (k=j-i+1) \\
 &< \sum_{i=1}^n \sum_{k=1}^n \frac{2}{k} \\
 &= 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k} \\
 &= 2 \sum_{i=1}^n H_n
 \end{aligned}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\begin{aligned}
 \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\
 &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \quad (k=j-i+1) \\
 &< \sum_{i=1}^n \sum_{k=1}^n \frac{2}{k} \\
 &= 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k} \\
 &= 2 \sum_{i=1}^n H_n = 2nH_n
 \end{aligned}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\begin{aligned}
 \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\
 &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \quad (k=j-i+1) \\
 &< \sum_{i=1}^n \sum_{k=1}^n \frac{2}{k} \\
 &= 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k} \\
 &= 2 \sum_{i=1}^n H_n = 2nH_n \in O(n \lg n).
 \end{aligned}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\begin{aligned}
 \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\
 &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \quad (k=j-i+1) \\
 &< \sum_{i=1}^n \sum_{k=1}^n \frac{2}{k} \\
 &= 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k} \\
 &= 2 \sum_{i=1}^n H_n = 2nH_n \in O(n \lg n).
 \end{aligned}$$

Average-case analysis of Quicksort

Hence the expected number of comparisons is

$$\begin{aligned}
 \sum_{i=1}^n \sum_{j=i+1}^n E(X_{i,j}) &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\
 &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \quad (k=j-i+1) \\
 &< \sum_{i=1}^n \sum_{k=1}^n \frac{2}{k} \\
 &= 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k} \\
 &= 2 \sum_{i=1}^n H_n = 2nH_n \in O(n \lg n).
 \end{aligned}$$

So the **average time** for Quicksort is $O(n \lg n)$.

Divide and Conquer

Divide and conquer paradigm

1. Split problem into subproblem(s)
2. Solve each subproblem (usually via recursive call)
3. Combine solution of subproblem(s) into solution of original problem

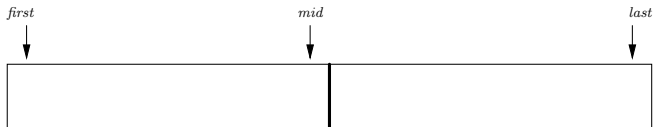
We will discuss two sorting algorithms based on this paradigm:

- ▶ Quicksort (done)
- ▶ Mergesort

MergeSort

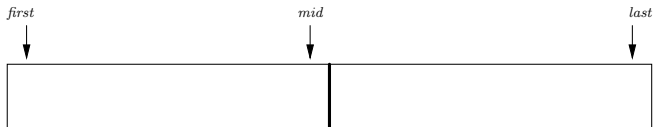
MergeSort

- ▶ Split array into two equal subarrays



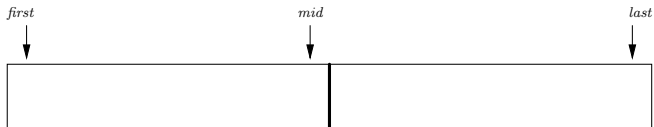
MergeSort

- ▶ Split array into two equal subarrays
- ▶ Sort both subarrays (recursively)



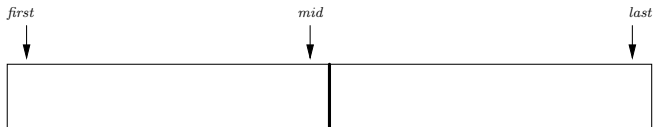
MergeSort

- ▶ Split array into two equal subarrays
- ▶ Sort both subarrays (recursively)
- ▶ Merge two sorted subarrays



MergeSort

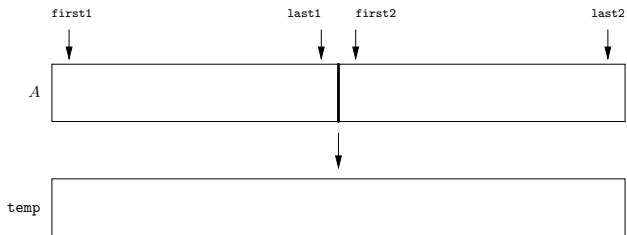
- ▶ Split array into two equal subarrays
- ▶ Sort both subarrays (recursively)
- ▶ Merge two sorted subarrays



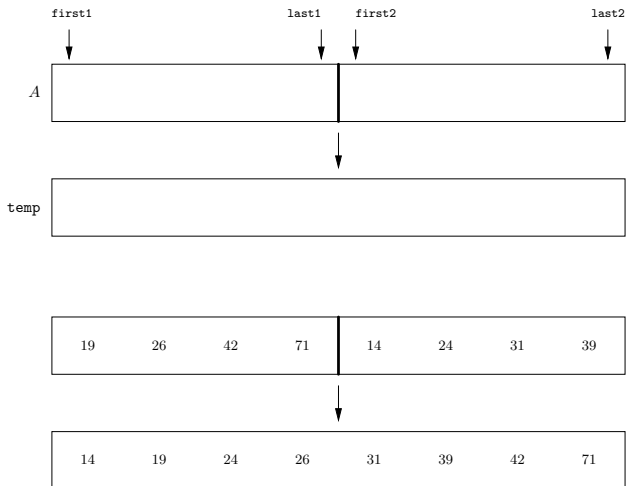
```
def mergeSort(A,first,last):  
    if first < last:  
        mid = [(first + last)/2]  
        mergeSort(A,first,mid)  
        mergeSort(A,mid+1,last)  
        merge(A,first,mid,mid+1,last)
```

The merge step

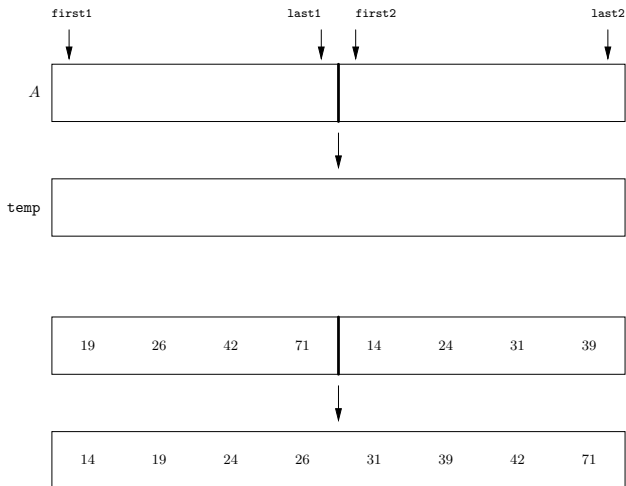
The merge step



The merge step



The merge step



Merging two lists of total size n requires at most $n - 1$ comparisons.

Code for the merge step

```
def merge(A,first1,last1,first2,last2):
    index1 = first1; index2 = first2; tempIndex = 0
    // Merge into temp array until one input array is exhausted
    while (index1 <= last1) and (index2 <= last2)
        if A[index1] <= A[index2]:
            temp[tempIndex++] = A[index1++]
        else:
            temp[tempIndex++] = A[index2++]
    // Copy appropriate trailer portion
    while (index1 <= last1): temp[tempIndex++] = A[index1++]
    while (index2 <= last2): temp[tempIndex++] = A[index2++]
    // Copy temp array back to A array
    tempIndex = 0; index = first1
    while (index <= last2): A[index++] = temp[tempIndex++]
```

Analysis of Mergesort

Analysis of Mergesort

$T(n)$ = number of comparisons required to sort n items in the worst case

Analysis of Mergesort

$T(n)$ = number of comparisons required to sort n items in the worst case

$$T(n) = \begin{cases} T(\lceil \frac{n}{2} \rceil) + T(\lfloor \frac{n}{2} \rfloor) + n - 1, & n > 1 \\ 0, & n = 1 \end{cases}$$

Analysis of Mergesort

$T(n)$ = number of comparisons required to sort n items in the worst case

$$T(n) = \begin{cases} T(\lceil \frac{n}{2} \rceil) + T(\lfloor \frac{n}{2} \rfloor) + n - 1, & n > 1 \\ 0, & n = 1 \end{cases}$$

The **asymptotic** solution of this recurrence equation is

$$T(n) = \Theta(n \log n)$$

Analysis of Mergesort

$T(n)$ = number of comparisons required to sort n items in the worst case

$$T(n) = \begin{cases} T(\lceil \frac{n}{2} \rceil) + T(\lfloor \frac{n}{2} \rfloor) + n - 1, & n > 1 \\ 0, & n = 1 \end{cases}$$

The **asymptotic** solution of this recurrence equation is

$$T(n) = \Theta(n \log n)$$

The **exact** solution of this recurrence equation is

$$T(n) = n \lceil \lg n \rceil - 2^{\lceil \lg n \rceil} + 1$$

Geometrical Application: Counting line intersections

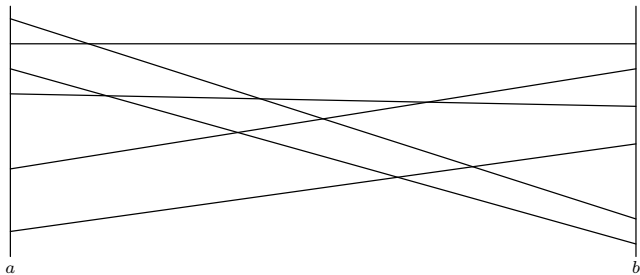
Geometrical Application: Counting line intersections

- ▶ **Input:** n lines in the plane, none of which are vertical; two vertical lines $x = a$ and $x = b$ (with $a < b$).

Geometrical Application: Counting line intersections

- ▶ **Input:** n lines in the plane, none of which are vertical; two vertical lines $x = a$ and $x = b$ (with $a < b$).

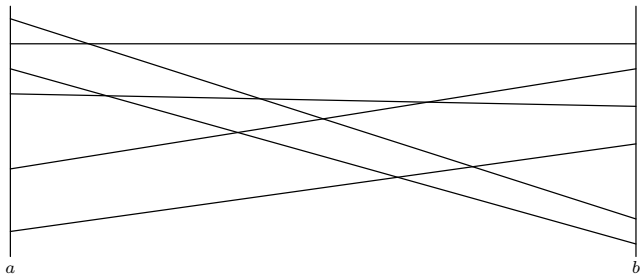
Example: $n = 6$



Geometrical Application: Counting line intersections

- ▶ **Input:** n lines in the plane, none of which are vertical; two vertical lines $x = a$ and $x = b$ (with $a < b$).
- ▶ **Problem:** Count/report all pairs of lines that intersect between the two vertical lines $x = a$ and $x = b$.

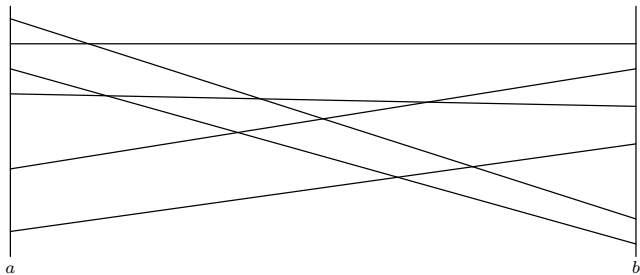
Example: $n = 6$



Geometrical Application: Counting line intersections

- ▶ **Input:** n lines in the plane, none of which are vertical; two vertical lines $x = a$ and $x = b$ (with $a < b$).
- ▶ **Problem:** Count/report all pairs of lines that intersect between the two vertical lines $x = a$ and $x = b$.

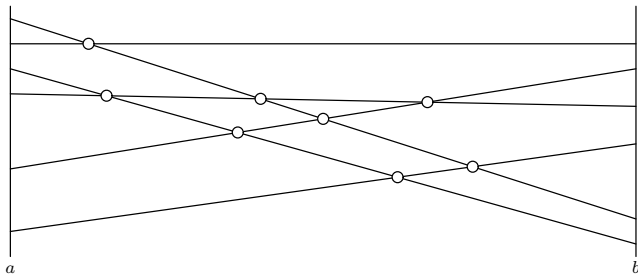
Example: $n = 6$ 8 intersections



Geometrical Application: Counting line intersections

- ▶ **Input:** n lines in the plane, none of which are vertical; two vertical lines $x = a$ and $x = b$ (with $a < b$).
- ▶ **Problem:** Count/report all pairs of lines that intersect between the two vertical lines $x = a$ and $x = b$.

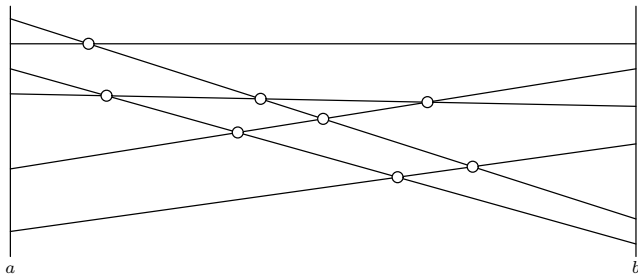
Example: $n = 6$ 8 intersections



Geometrical Application: Counting line intersections

- ▶ **Input:** n lines in the plane, none of which are vertical; two vertical lines $x = a$ and $x = b$ (with $a < b$).
- ▶ **Problem:** Count/report all pairs of lines that intersect between the two vertical lines $x = a$ and $x = b$.

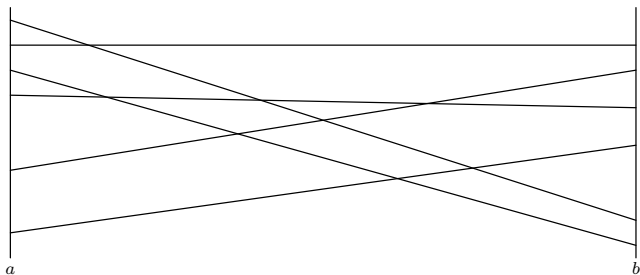
Example: $n = 6$ 8 intersections



Checking every pair of lines takes $\Theta(n^2)$ time. We can do better.

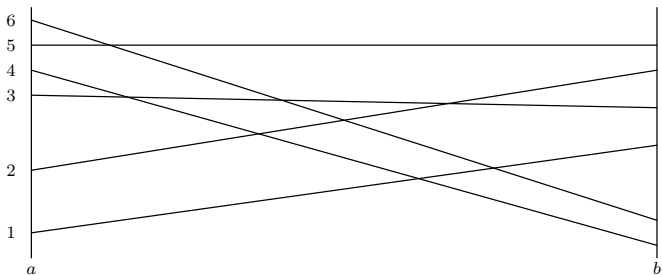
Geometrical Application: Counting line intersections

Geometrical Application: Counting line intersections



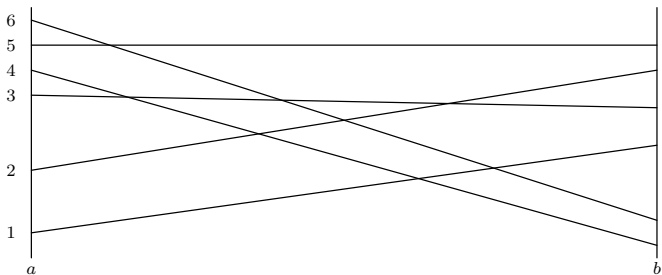
Geometrical Application: Counting line intersections

1. Sort the lines according to the y -coordinate of their intersection with the line $x = a$. Number the lines in sorted order.



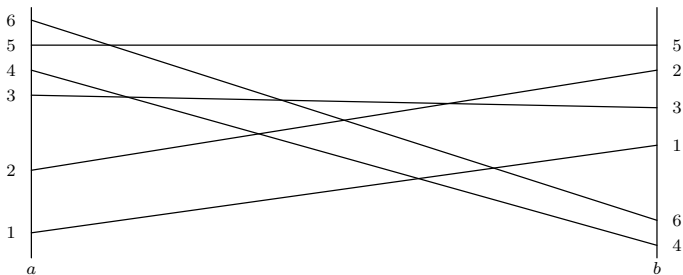
Geometrical Application: Counting line intersections

1. Sort the lines according to the y -coordinate of their intersection with the line $x = a$. Number the lines in sorted order. [$O(n \log n)$ time]



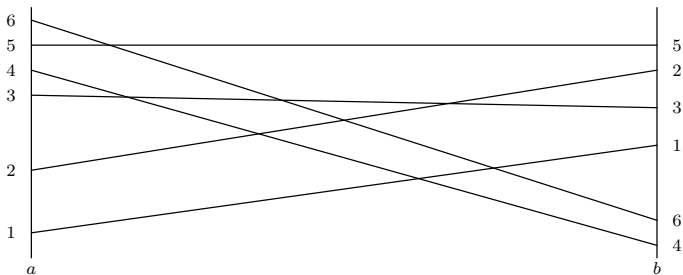
Geometrical Application: Counting line intersections

- Sort the lines according to the y -coordinate of their intersection with the line $x = a$. Number the lines in sorted order. [$O(n \log n)$ time]
- Produce the sequence of line numbers sorted according to the y -coordinate of their intersection with the line $x = b$



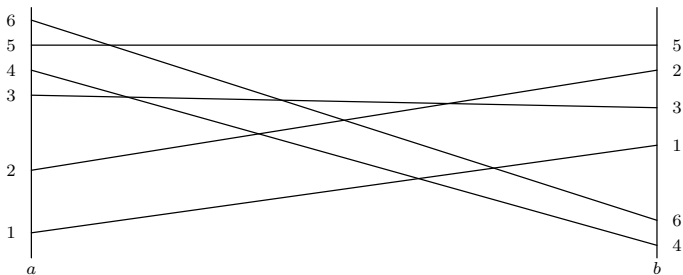
Geometrical Application: Counting line intersections

- Sort the lines according to the y -coordinate of their intersection with the line $x = a$. Number the lines in sorted order. [$O(n \log n)$ time]
- Produce the sequence of line numbers sorted according to the y -coordinate of their intersection with the line $x = b$ [$O(n \log n)$ time]



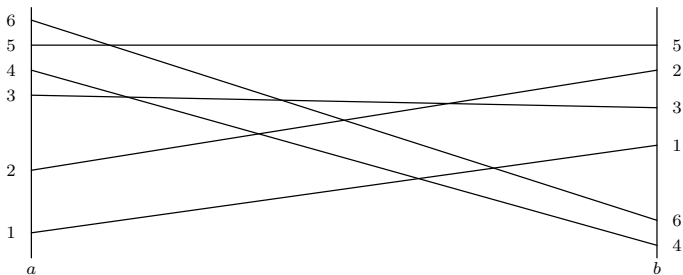
Geometrical Application: Counting line intersections

- Sort the lines according to the y -coordinate of their intersection with the line $x = a$. Number the lines in sorted order. [$O(n \log n)$ time]
- Produce the sequence of line numbers sorted according to the y -coordinate of their intersection with the line $x = b$ [$O(n \log n)$ time]
- Count/report inversions in the sequence produced in step 2.



Geometrical Application: Counting line intersections

- Sort the lines according to the y -coordinate of their intersection with the line $x = a$. Number the lines in sorted order. [$O(n \log n)$ time]
- Produce the sequence of line numbers sorted according to the y -coordinate of their intersection with the line $x = b$ [$O(n \log n)$ time]
- Count/report inversions in the sequence produced in step 2.



So the problem reduces to counting/reporting inversions.

Counting Inversions: An Application of Mergesort

Counting Inversions: An Application of Mergesort

An **inversion** in a sequence or list is a pair of items such that the larger one precedes the smaller one.

Counting Inversions: An Application of Mergesort

An **inversion** in a sequence or list is a pair of items such that the larger one precedes the smaller one.

Example: The list `[18, 29, 12, 15, 32, 10]` has 9 inversions:

`(18, 12), (18, 15), (18, 10), (29, 12), (29, 15), (29, 10), (12, 10), (15, 10), (32, 10)`

Counting Inversions: An Application of Mergesort

An **inversion** in a sequence or list is a pair of items such that the larger one precedes the smaller one.

Example: The list $[18, 29, 12, 15, 32, 10]$ has 9 inversions:

$(18, 12), (18, 15), (18, 10), (29, 12), (29, 15), (29, 10), (12, 10), (15, 10), (32, 10)$

In a list of size n , there can be as many as $\binom{n}{2}$ inversions.

Counting Inversions: An Application of Mergesort

An **inversion** in a sequence or list is a pair of items such that the larger one precedes the smaller one.

Example: The list $[18, 29, 12, 15, 32, 10]$ has 9 inversions:

$(18, 12), (18, 15), (18, 10), (29, 12), (29, 15), (29, 10), (12, 10), (15, 10), (32, 10)$

In a list of size n , there can be as many as $\binom{n}{2}$ inversions.

Problem: Given a list, compute the number of inversions.

Counting Inversions: An Application of Mergesort

An **inversion** in a sequence or list is a pair of items such that the larger one precedes the smaller one.

Example: The list $[18, 29, 12, 15, 32, 10]$ has 9 inversions:

$(18, 12), (18, 15), (18, 10), (29, 12), (29, 15), (29, 10), (12, 10), (15, 10), (32, 10)$

In a list of size n , there can be as many as $\binom{n}{2}$ inversions.

Problem: Given a list, compute the number of inversions.

Brute force solution: Check each pair i, j with $i < j$ to see if $L[i] > L[j]$.

Counting Inversions: An Application of Mergesort

An **inversion** in a sequence or list is a pair of items such that the larger one precedes the smaller one.

Example: The list $[18, 29, 12, 15, 32, 10]$ has 9 inversions:

$(18, 12), (18, 15), (18, 10), (29, 12), (29, 15), (29, 10), (12, 10), (15, 10), (32, 10)$

In a list of size n , there can be as many as $\binom{n}{2}$ inversions.

Problem: Given a list, compute the number of inversions.

Brute force solution: Check each pair i, j with $i < j$ to see if $L[i] > L[j]$. This gives a $\Theta(n^2)$ algorithm.

Counting Inversions: An Application of Mergesort

An **inversion** in a sequence or list is a pair of items such that the larger one precedes the smaller one.

Example: The list $[18, 29, 12, 15, 32, 10]$ has 9 inversions:

$(18, 12), (18, 15), (18, 10), (29, 12), (29, 15), (29, 10), (12, 10), (15, 10), (32, 10)$

In a list of size n , there can be as many as $\binom{n}{2}$ inversions.

Problem: Given a list, compute the number of inversions.

Brute force solution: Check each pair i, j with $i < j$ to see if $L[i] > L[j]$. This gives a $\Theta(n^2)$ algorithm. We can do better.

Inversion Counting

Inversion Counting

Sorting is the process of removing inversions. So to count inversions:

Inversion Counting

Sorting is the process of removing inversions. So to count inversions:

- ▶ Run a sorting algorithm

Inversion Counting

Sorting is the process of removing inversions. So to count inversions:

- ▶ Run a sorting algorithm
- ▶ Every time data is rearranged, keep track of how many inversions are being removed.

Inversion Counting

Sorting is the process of removing inversions. So to count inversions:

- ▶ Run a sorting algorithm
- ▶ Every time data is rearranged, keep track of how many inversions are being removed.

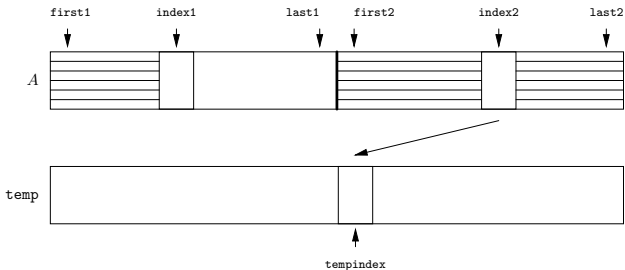
In principle, we can use any sorting algorithm to count inversions. Mergesort works particularly nicely.

Inversion Counting with MergeSort

In Mergesort, the only time we rearrange data is during the merge step.

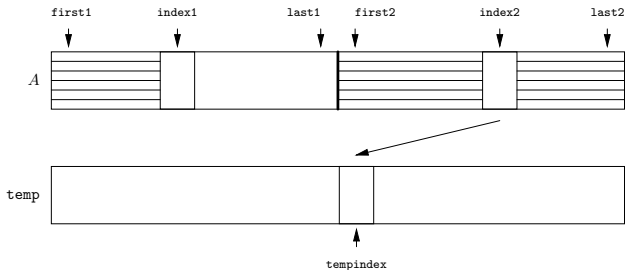
Inversion Counting with MergeSort

In Mergesort, the only time we rearrange data is during the merge step.



Inversion Counting with MergeSort

In Mergesort, the only time we rearrange data is during the merge step.

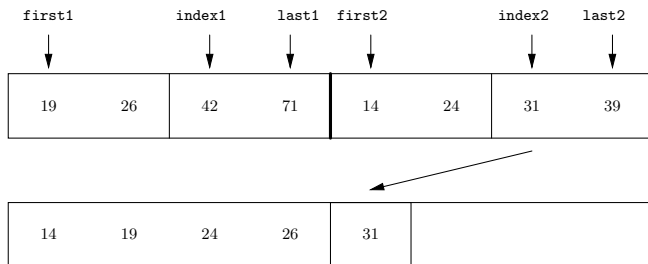


The number of inversions removed is:

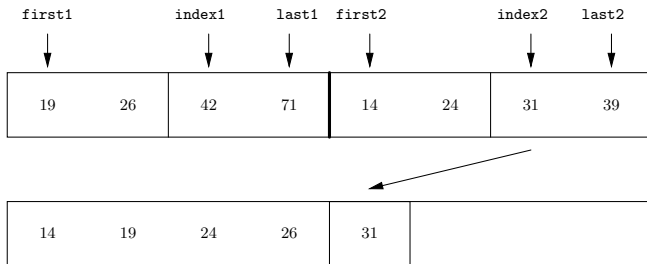
$$\text{last1} - \text{index1} + 1$$

Example

Example



Example



2 inversions removed: $(42, 31)$ and $(71, 31)$

Pseudocode for the merge step with inversion counting

Pseudocode for the merge step with inversion counting

```
def merge(A,first1,last1,first2,last2):
    index1 = first1; index2 = first2; tempIndex = 0
    invCount = 0
    // Merge into temp array until one input array is exhausted
    while (index1 <= last1) and (index2 <= last2)
        if A[index1] <= A[index2]:
            temp[tempIndex++] = A[index1++]
        else:
            temp[tempIndex++] = A[index2++]
            invCount += last1 - index1 + 1;
    // Copy appropriate trailer portion
    while (index1 <= last1): temp[tempIndex++] = A[index1++]
    while (index2 <= last2): temp[tempIndex++] = A[index2++]
    // Copy temp array back to A array
    tempIndex = 0; index = first1
    while (index <= last2): A[index++] = temp[tempIndex++]
    return invCount
```

Pseudocode for MergeSort with inversion counting

Pseudocode for MergeSort with inversion counting

```
def mergeSort(A,first,last):
    invCount = 0
    if first < last:
        mid = [(first + last)/2]
        invCount += mergeSort(A,first,mid)
        invCount += mergeSort(A,mid+1,last)
        invCount += merge(A,first,mid,mid+1,last)
    return invCount
```

Pseudocode for MergeSort with inversion counting

```
def mergeSort(A,first,last):
    invCount = 0
    if first < last:
        mid = [(first + last)/2]
        invCount += mergeSort(A,first,mid)
        invCount += mergeSort(A,mid+1,last)
        invCount += merge(A,first,mid,mid+1,last)
    return invCount
```

Running time is the same as standard mergeSort: $O(n \log n)$

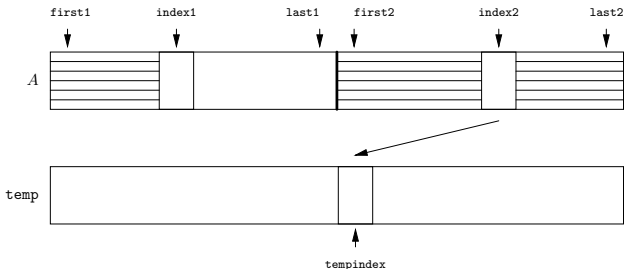
Listing inversions

Listing inversions

We have just seen that we can count inversions without increasing the asymptotic running time of Mergesort. Suppose we want to **list** inversions.

Listing inversions

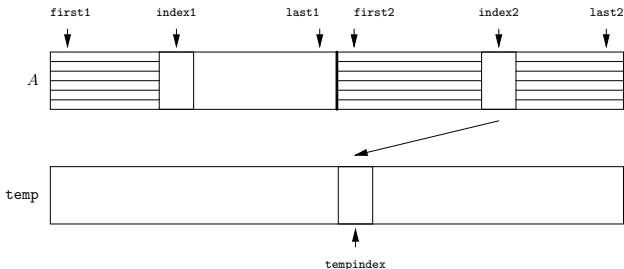
We have just seen that we can count inversions without increasing the asymptotic running time of Mergesort. Suppose we want to **list** inversions. When we remove inversions, we list all inversions removed:



$(A[index1], A[index2]), (A[index1+1], A[index2]), \dots,$
 $(A[last1], A[index2]).$

Listing inversions

We have just seen that we can count inversions without increasing the asymptotic running time of Mergesort. Suppose we want to **list** inversions. When we remove inversions, we list all inversions removed:



$(A[index1], A[index2]), (A[index1+1], A[index2]), \dots,$
 $(A[last1], A[index2]).$

The extra work to do the reporting is proportional to the number of inversions reported.

Inversion counting summary

Inversion counting summary

Using a slight modification of Mergesort, we can ...

Inversion counting summary

Using a slight modification of Mergesort, we can ...

- ▶ **Count** inversions in $O(n \log n)$ time.

Inversion counting summary

Using a slight modification of Mergesort, we can ...

- ▶ **Count** inversions in $O(n \log n)$ time.
- ▶ **Report** inversions in $O(n \log n + k)$ time, where k is the number of inversions.

Inversion counting summary

Using a slight modification of Mergesort, we can ...

- ▶ **Count** inversions in $O(n \log n)$ time.
- ▶ **Report** inversions in $O(n \log n + k)$ time, where k is the number of inversions.

The same results hold for the line-intersection counting problem.

Inversion counting summary

Using a slight modification of Mergesort, we can ...

- ▶ **Count** inversions in $O(n \log n)$ time.
- ▶ **Report** inversions in $O(n \log n + k)$ time, where k is the number of inversions.

The same results hold for the line-intersection counting problem.

The reporting algorithm is an example of an **output-sensitive** algorithm. The performance of the algorithm depends on the size of the output as well as the size of the input.