

Algorithms: Sorting + (Time) Analysis

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Lecture 2, 19.02.2025

Recall Last Lecture

- ▶ CS-250: A lot of interesting and useful material!

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- ▶ An algorithm describes a specific computational procedure for achieving that input/output relationship

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- ▶ A computational problem is defined by an input/output relationship
 - ▶ Example: **INPUT:** n **OUTPUT:** $\sum_{i=1}^n i$
- ▶ An algorithm describes a specific computational procedure for achieving that input/output relationship
 - ▶ Example: return $n(n+1)/2$
- ▶ “Time + Space” is crucial for the usefulness of an algorithm

The Growth of Functions

- Three algorithms: A, B, C with different running times in ms.

	A ($1000 \log_2 n$ ms)	B ($2n^2$ ms)	C (2^n ms)

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$n = 32$	5 s	2 s 48 ms	\approx 49 days 18h

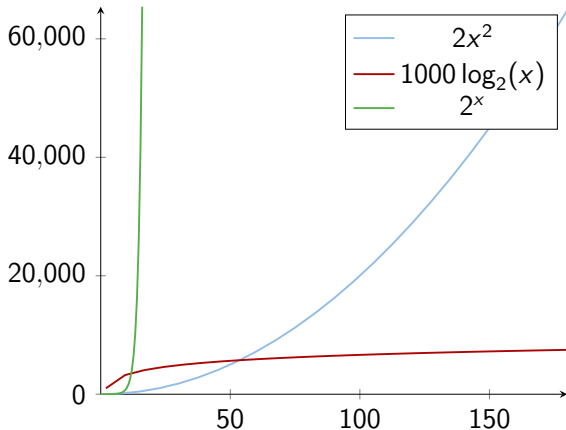
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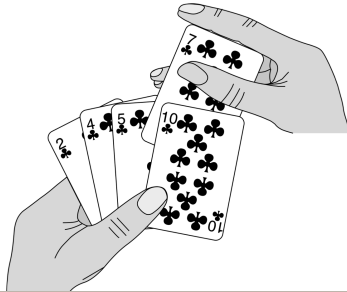
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$n = 64$	6 s	8 s 192 ms	> age of the universe

The Growth of Functions

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SORTING

Insertion Sort

The sorting problem

Definition

INPUT: A sequence of n numbers $\langle a_1, a_2, \dots, a_n \rangle$.

OUTPUT: A permutation (reordering) $\langle a'_1, a'_2, \dots, a'_n \rangle$ of the input sequence such that $a'_1 \leq a'_2 \leq \dots \leq a'_n$.

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For example

- ▶ Given the input $\langle 5, 2, 4, 6, 1, 3 \rangle$

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For example

- ▶ Given the input $\langle 5, 2, 4, 6, 1, 3 \rangle$
- ▶ a correct output is $\langle 1, 2, 3, 4, 5, 6 \rangle$

Insertion Sort - The Idea

Like sorting a hand of playing cards

- ▶ Start with an empty left hand of playing cards and the cards face down on the table



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- ▶ Start with an empty left hand of playing cards and the cards face down on the table
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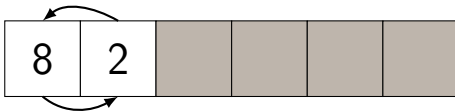
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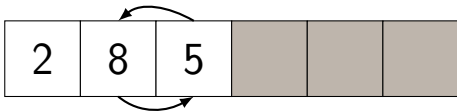
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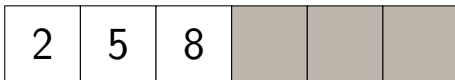
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2	5	8	10		
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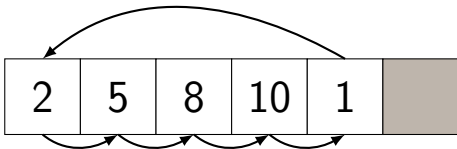
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2	5	8	10	1	
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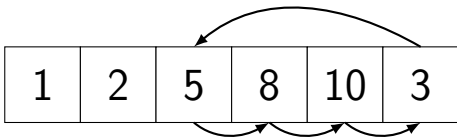
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1	2	5	8	10	3
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- ▶ To find the correct position for a card, compare it with each of the cards already in the hand, from right to left.
- ▶ At all times, the cards, held in the left hand are sorted, and these cards were originally the top cards of the pile on the table

1	2	3	5	8	10
---	---	---	---	---	----



Insertion Sort

The Algorithm

- ▶ Takes as parameters an array $A[1 \dots n]$ and the length n of the array

```
INSERTION-SORT( $A, n$ )  
  for  $j = 2$  to  $n$   
     $key = A[j]$   
    // Insert  $A[j]$  into the sorted sequence  $A[1 \dots j - 1]$ .  
     $i = j - 1$   
    while  $i > 0$  and  $A[i] > key$   
       $A[i + 1] = A[i]$   
       $i = i - 1$   
     $A[i + 1] = key$ 
```

Insertion Sort

Example on $\langle 8, 2, 5, 10, 1, 3 \rangle$

key:

j:

i:

A:

8	2	5	10	1	3
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Insertion Sort

Example on $\langle 8, 2, 5, 10, 1, 3 \rangle$

key:	0
j:	2
i:	0

A:	8	2	5	10	1	3
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PROVING ALGORITHMS CORRECT

Loop invariants

Insertion Sort

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Loop invariant:

At the start of each iteration of the “outer” **for** loop – the loop indexed by j – the subarray $A[1 \dots, j - 1]$ consists of the elements originally in $A[1, \dots, j - 1]$ but in sorted order.

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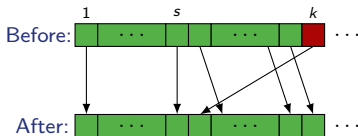
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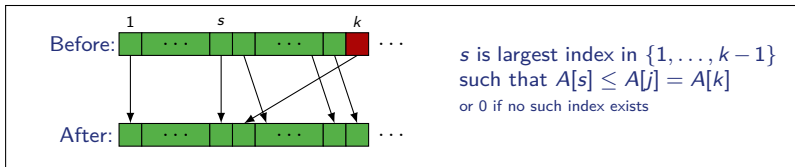
s is largest index in $\{1, \dots, k - 1\}$
such that $A[s] \leq A[k]$
or 0 if no such index exists

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- ▶ The subarray $A[1 \dots k]$ then consists of the elements originally in $A[1 \dots k]$ in a sorted order. Incrementing j (to $k + 1$) for the next iteration of the **for** loop then preserves the loop invariant :)

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ANALYZING ALGORITHMS

Computational Model

We want to predict the resources that the algorithm requires. Usually, running time.

For that we need a computational model

Random-access machine (RAM) model

- ▶ Instructions are executed one after another
- ▶ Simplification basic instructions take constant ($O(1)$) time
 - ▶ Arithmetic: add, subtract, multiply, divide, remainder, floor, ceiling
 - ▶ Data movement: load, store, copy.
 - ▶ Control: conditional/unconditional branch, subroutine call and return
- ▶ We don't worry about precision, although it is crucial in certain numerical applications

Analyzing an algorithm's running time (1/2)

Time it takes depend on the input

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- ▶ Usually, the number of items in the input. Like the size n of the array being sorted
- ▶ If multiplying two integers, could be the total number of bits in the two integers
- ▶ Could be described by more than one number: e.g. graph algorithm running times are usually expressed in terms of the number of vertices and the number of edges in the input graph.

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- ▶ Figure that each line of pseudocode requires a constant amount of time
- ▶ One line may take a different amount of time than another, but each execution of line i takes the same amount of time c_i
- ▶ This is assuming that the line consists only of primitive operations
 - ▶ If the line is a subroutine call, then the actual call takes constant time, but the execution of the subroutine might not
 - ▶ If the line specifies operations other than primitive ones, then it might take more than constant time. Example: "sort the points by x-coordinate"

Analysis of insertion sort

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Analysis of insertion sort

INSERTION-SORT(A, n)	<i>cost</i>	<i>times</i>
for $j = 2$ to n	c_1	n
$key = A[j]$	c_2	$n - 1$
// Insert $A[j]$ into the sorted sequence $A[1 \dots j - 1]$.	0	$n - 1$
$i = j - 1$	c_4	$n - 1$
while $i > 0$ and $A[i] > key$	c_5	$\sum_{j=2}^n t_j$
$A[i + 1] = A[i]$	c_6	$\sum_{j=2}^n (t_j - 1)$
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$$T(n) = c_1 n + c_2(n - 1) + c_4(n - 1) + c_5(n - 1) + c_8(n - 1) = \Theta(n)$$

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$$\begin{aligned} T(n) &= c_1 n + c_2(n - 1) + c_4(n - 1) + c_5 \frac{n(n + 1) - 2}{2} \\ &\quad + (c_6 + c_7) \frac{n \cdot (n - 1)}{2} + c_8(n - 1) = \Theta(n^2) \end{aligned}$$

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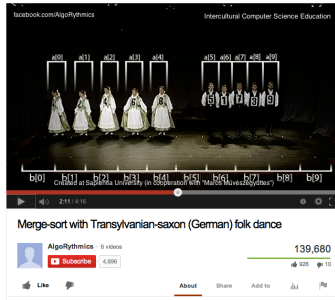
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- ▶ Gives a guaranteed upper bound on the running time for any input
- ▶ For some algorithms, the worst case occurs often. For example, when searching, the worst case often occurs when the item being searched for is not present
- ▶ Average case often as bad as worst-case: Suppose that we randomly choose n numbers as the input to insertion sort

Order of growth: Focus on the important features

- ▶ Drop lower-order terms
- ▶ Ignore the constant coefficient in the leading term



SORTING BY DIVIDE-AND-CONQUER

Merge Sort

Divide-and-Conquer

Powerful algorithmic approach:

recursively divide problem into smaller subproblems

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Combine the subproblem solutions to give a solution to the original problem

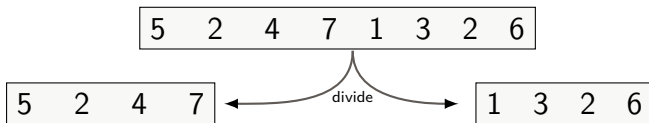
Merge Sort = D & C applied to sorting

Example $\langle 5, 2, 4, 7, 1, 3, 2, 6 \rangle$

5	2	4	7	1	3	2	6
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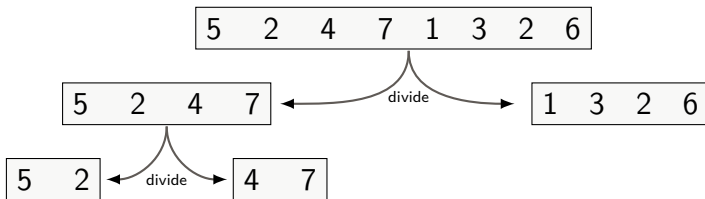
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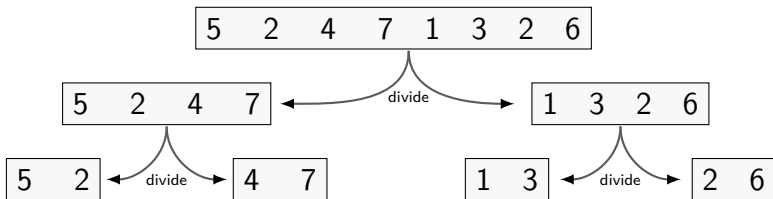
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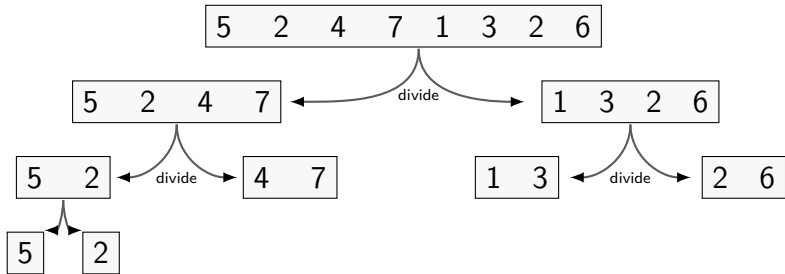
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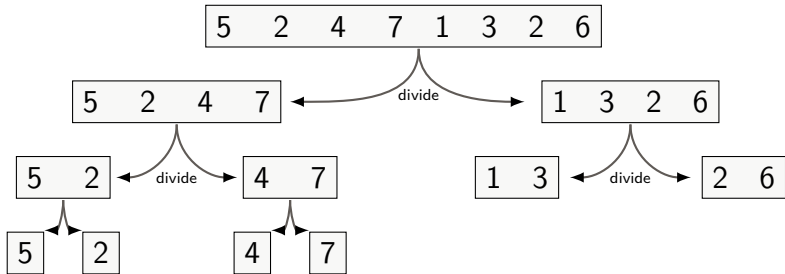
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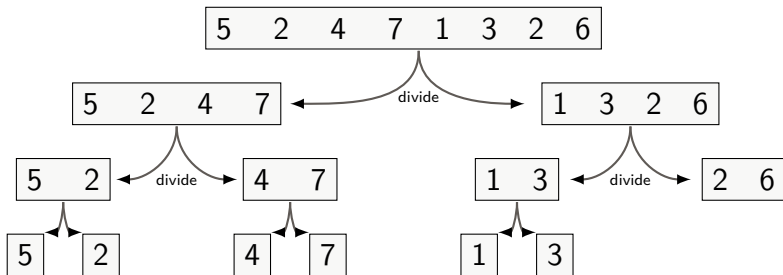
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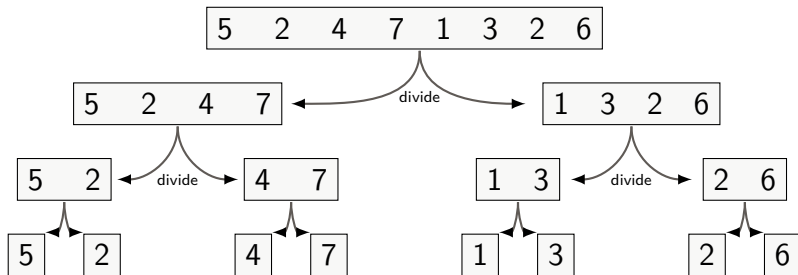
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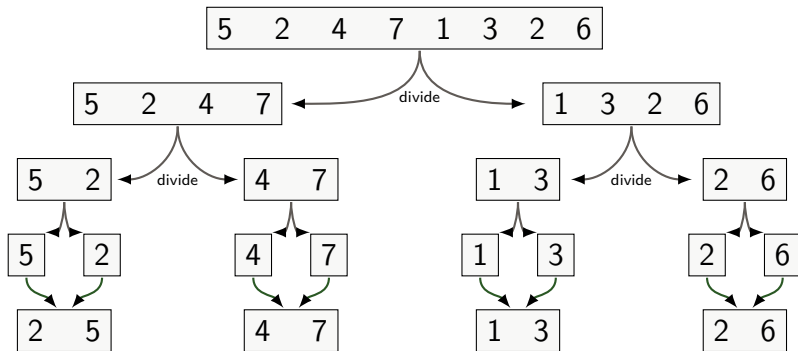
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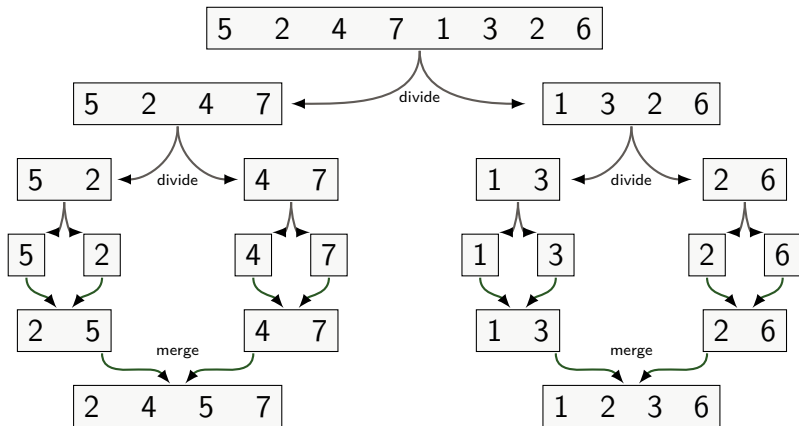
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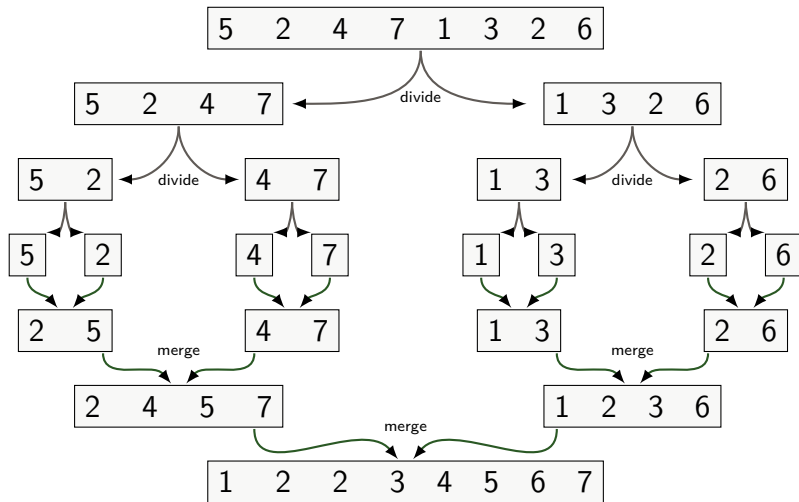
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Merge sort

To sort $A[p \dots r]$:

- Divide** by splitting into two subarrays $A[p \dots q]$ and $A[q + 1, \dots, r]$, where q is the halfway point of $A[p \dots r]$
- Conquer** by recursively sorting the two subarrays $A[p \dots q]$ and $A[q + 1, \dots, r]$
- Combine** by merging the two sorted subarrays $A[p \dots q]$ and $A[q + 1, \dots, r]$ to produce a single sorted subarray $A[p \dots r]$

Merge sort

To sort $A[p \dots r]$:

- Divide** by splitting into two subarrays $A[p \dots q]$ and $A[q + 1, \dots, r]$, where q is the halfway point of $A[p \dots r]$
- Conquer** by recursively sorting the two subarrays $A[p \dots q]$ and $A[q + 1, \dots, r]$
- Combine** by merging the two sorted subarrays $A[p \dots q]$ and $A[q + 1, \dots, r]$ to produce a single sorted subarray $A[p \dots r]$

```
MERGE-SORT( $A, p, r$ )  
  if  $p < r$                                 // check for base case  
     $q = \lfloor (p + r)/2 \rfloor$                   // divide  
    MERGE-SORT( $A, p, q$ )                    // conquer  
    MERGE-SORT( $A, q + 1, r$ )                // conquer  
    MERGE( $A, p, q, r$ )                      // combine
```


Merging

What remains is the Merge procedure to solve the “merge” problem:

Definition

INPUT: Array A and indices $p \leq q < r$ such that subarrays $A[p \dots q]$, $A[q + 1 \dots r]$ are sorted.

OUTPUT: The two subarrays are merged into a single sorted subarray in $A[p \dots r]$.

We will give a procedure that solves this problem in time $\Theta(n)$ where n is the size of the subproblem, i.e.,

$$n = r - p + 1$$

Idea behind linear-time merging

Think of two pile of cards that are placed face up

- ▶ Basic step: pick the smaller of the two cards and place it in the output pile



Idea behind linear-time merging

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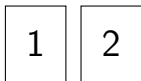
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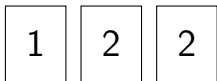
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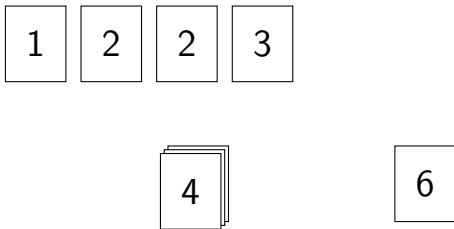
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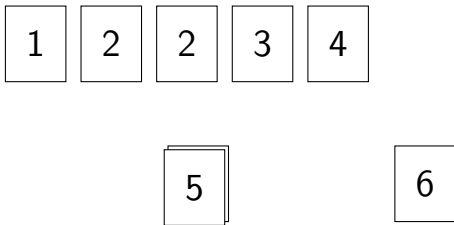
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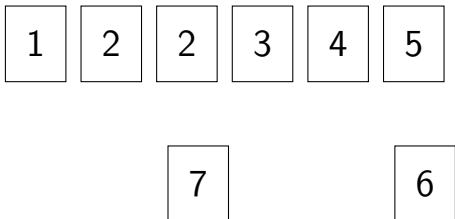
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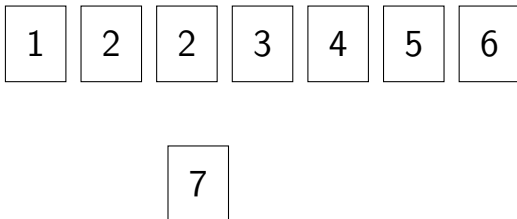
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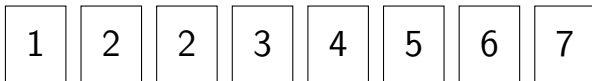
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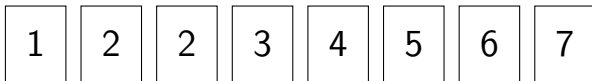
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Idea behind linear-time merging

Think of two pile of cards that are placed face up

- ▶ Basic step: pick the smaller of the two cards and place it in the output pile
- ▶ There are $\leq n$ basic steps, since each basic step removes one card from the input piles, and we started with n cards in the input pile
- ▶ Therefore the procedure should take $\theta(n)$ time



Implementation Simplification

Instead of checking whether a pile is empty:

- ▶ Put in the bottom of each input pile a special **sentinel** card of value ∞
- ▶ Stop once we have performed $n = r - p + 1$ basic steps (picked n cards)



Implementation Simplification

Instead of checking whether a pile is empty:

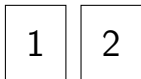
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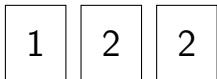
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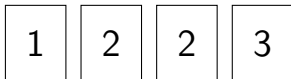
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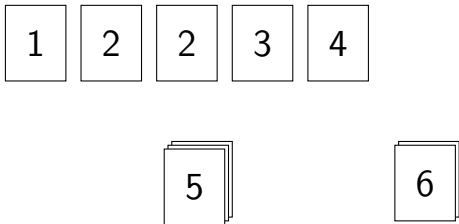
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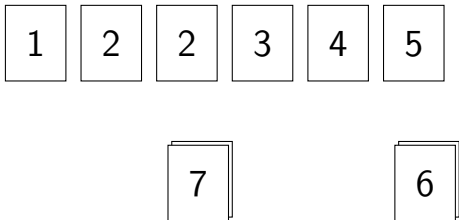
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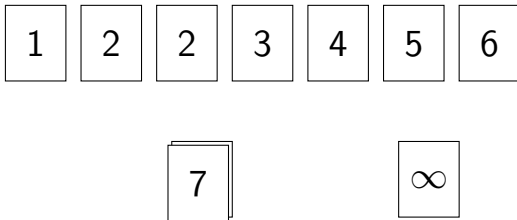
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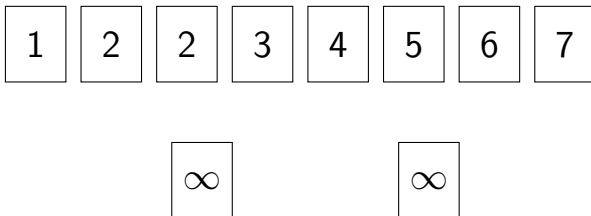
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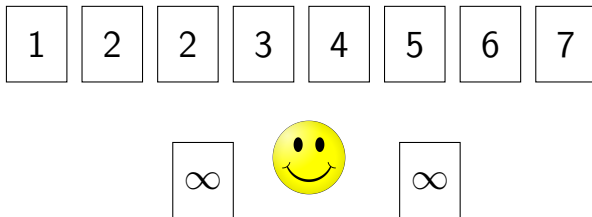
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Merging Algorithm

A:

	A[p]			A[q]				A[r]
	2	4	5	7	1	2	3	6

MERGE(A, p, q, r)

$n_1 = q - p + 1$

$n_2 = r - q$

let $L[1 \dots n_1 + 1]$ and $R[1 \dots n_2 + 1]$ be new arrays

for $i = 1$ **to** n_1

$L[i] = A[p + i - 1]$

for $j = 1$ **to** n_2

$R[j] = A[q + j]$

$L[n_1 + 1] = \infty$

$R[n_2 + 1] = \infty$

$i = 1$

$j = 1$

for $k = p$ **to** r

if $L[i] \leq R[j]$

$A[k] = L[i]$

$i = i + 1$

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Merging Algorithm

	A[p]		A[q]			A[r]		
A:	2	4	5	7	1	2	3	6

L:					
----	--	--	--	--	--

R:					
----	--	--	--	--	--

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Merging Algorithm

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Merging Algorithm

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Merging Algorithm

	A[p]		A[q]				A[r]	
A:	2	4	5	7	1	2	3	6

L:	2	4	5	7	∞	R:	1	2	3	6	∞
----	---	---	---	---	----------	----	---	---	---	---	----------

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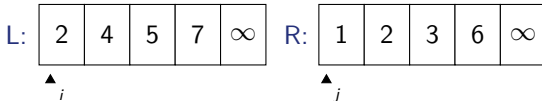
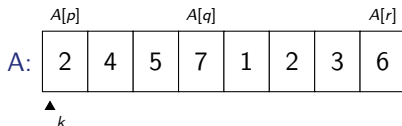
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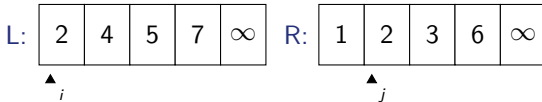
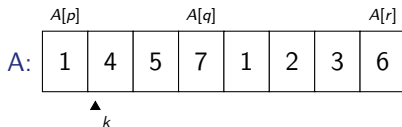
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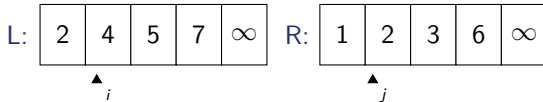
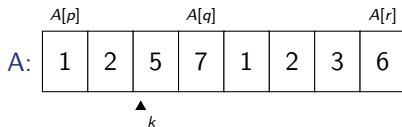
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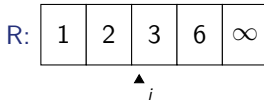
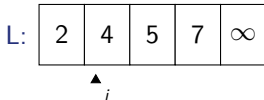
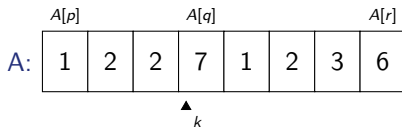
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Merging Algorithm



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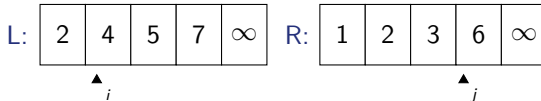
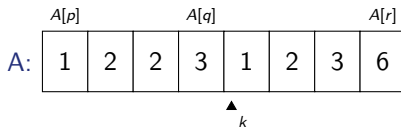
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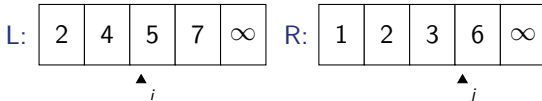
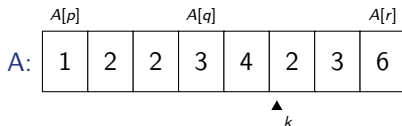
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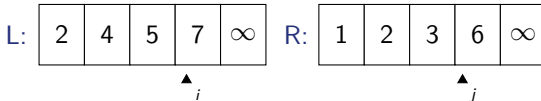
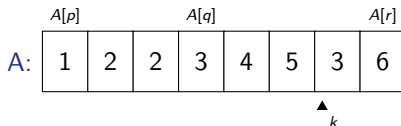
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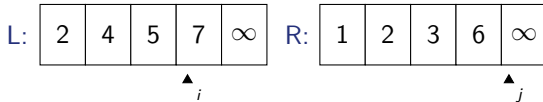
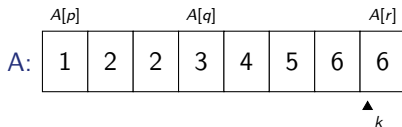
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Merging Algorithm

A:

	A[p]		A[q]				A[r]
1	2	2	3	4	5	6	7

L:

2	4	5	7	∞
---	---	---	---	----------

R:

1	2	3	6	∞
---	---	---	---	----------

\blacktriangle i
 \blacktriangle j

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Merging Algorithm

► Runtime analysis?

MERGE(A, p, q, r)

$n_1 = q - p + 1$

$n_2 = r - q$

let $L[1 \dots n_1 + 1]$ and $R[1 \dots n_2 + 1]$ be new arrays

for $i = 1$ **to** n_1

$L[i] = A[p + i - 1]$

for $j = 1$ **to** n_2

$R[j] = A[q + j]$

$L[n_1 + 1] = \infty$

$R[n_2 + 1] = \infty$

$i = 1$

$j = 1$

for $k = p$ **to** r

if $L[i] \leq R[j]$

$A[k] = L[i]$

$i = i + 1$

else $A[k] = R[j]$

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Merging Algorithm

► Runtime analysis?

MERGE(A, p, q, r)

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$A[k] = L[i]$

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else $A[k] = R[j]$

$j = j + 1$

Merge runs in time $\Theta(n)$ where n is the number of elements in the subarray, i.e.,

$$n = r - p + 1$$

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- ▶ Let $D(n)$ be the time to divide and let $C(n)$ the time to combine solutions.
- ▶ We get the recurrence

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq c, \\ aT(n/b) + D(n) + C(n) & \text{otherwise.} \end{cases}$$

Analysis of Merge Sort

MERGE-SORT(A, p, r)

if $p < r$

$q = \lfloor (p + r)/2 \rfloor$

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// check for base case

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Recurrence for merge sort running time is

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1, \\ 2T(n/2) + \Theta(n) & \text{otherwise.} \end{cases}$$

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- ▶ For small instances insertion sort can still be faster
- ▶ Insertion sort is also **in place**: the numbers are rearranged within the array (with at most a constant number outside the array at any time)
- ▶ Merge sort is not in place!