

# Divide-and-Conquer

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- **Divide** the problem into a number of sub-problems
  - Similar sub-problems of smaller size
- **Conquer** the sub-problems
  - Solve the sub-problems recursively
  - Sub-problem size small enough  $\Rightarrow$  solve the problems in straightforward manner
- **Combine** the solutions of the sub-problems
  - Obtain the solution for the original problem

# Merge Sort Approach

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- To sort an array  $A[p \dots r]$ :
- **Divide**
  - Divide the  $n$ -element sequence to be sorted into two subsequences of  $n/2$  elements each
- **Conquer**
  - Sort the subsequences recursively using merge sort
  - When the size of the sequences is 1 there is nothing more to do
- **Combine**
  - Merge the two sorted subsequences

# Merge Sort

*Alg.:* MERGE-SORT( $A, p, r$ )

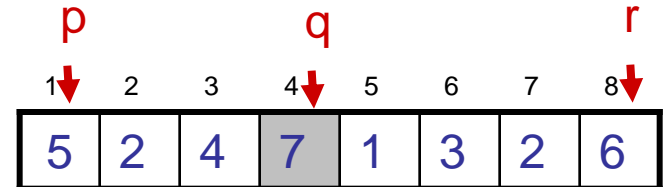
if  $p < r$

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

MERGE-SORT( $A, p, q$ )

MERGE-SORT( $A, q + 1, r$ )

MERGE( $A, p, q, r$ )



▷ Check for base case

▷ Divide

▷ Conquer

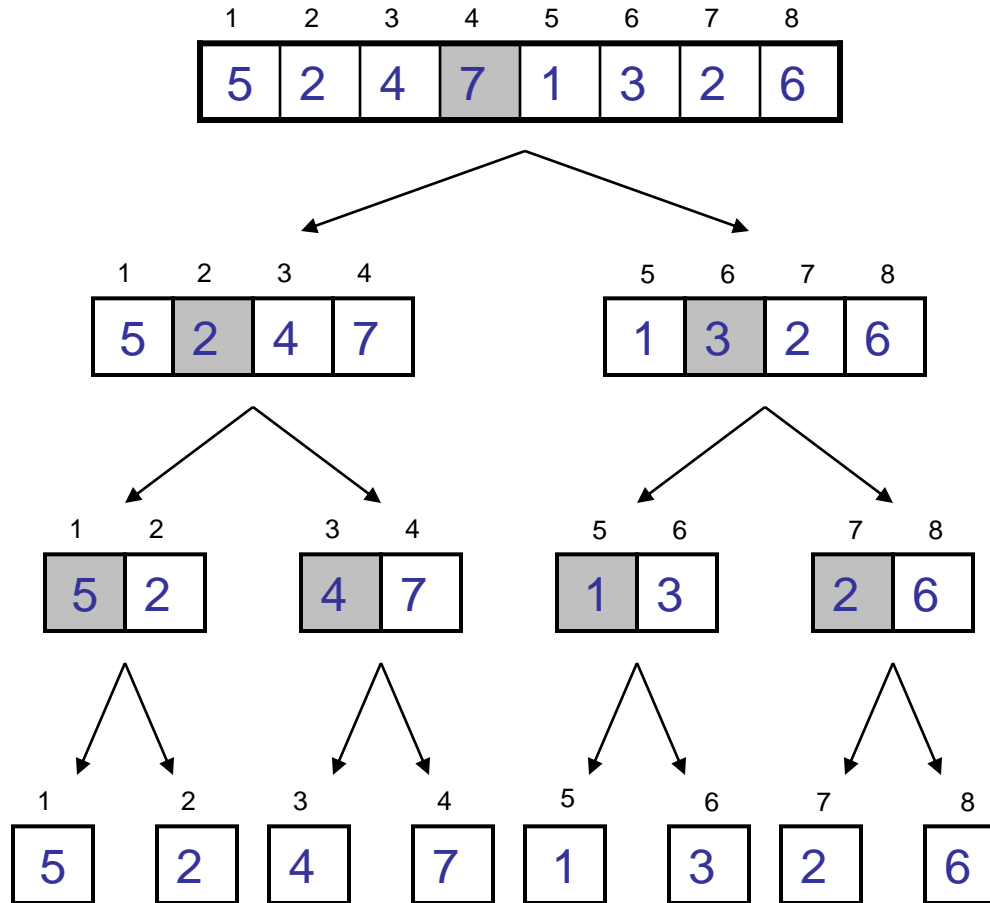
▷ Conquer

▷ Combine

- Initial call: MERGE-SORT( $A, 1, n$ )

# Example – n Power of 2

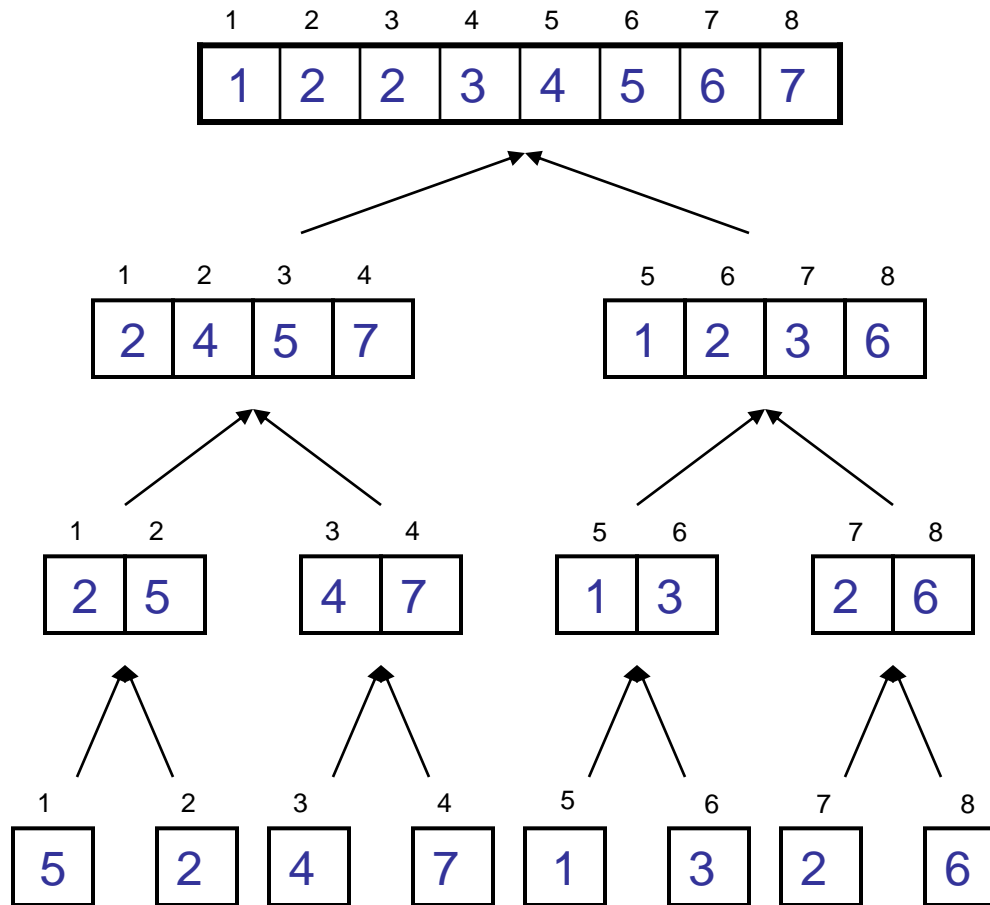
Divide



$q = 4$

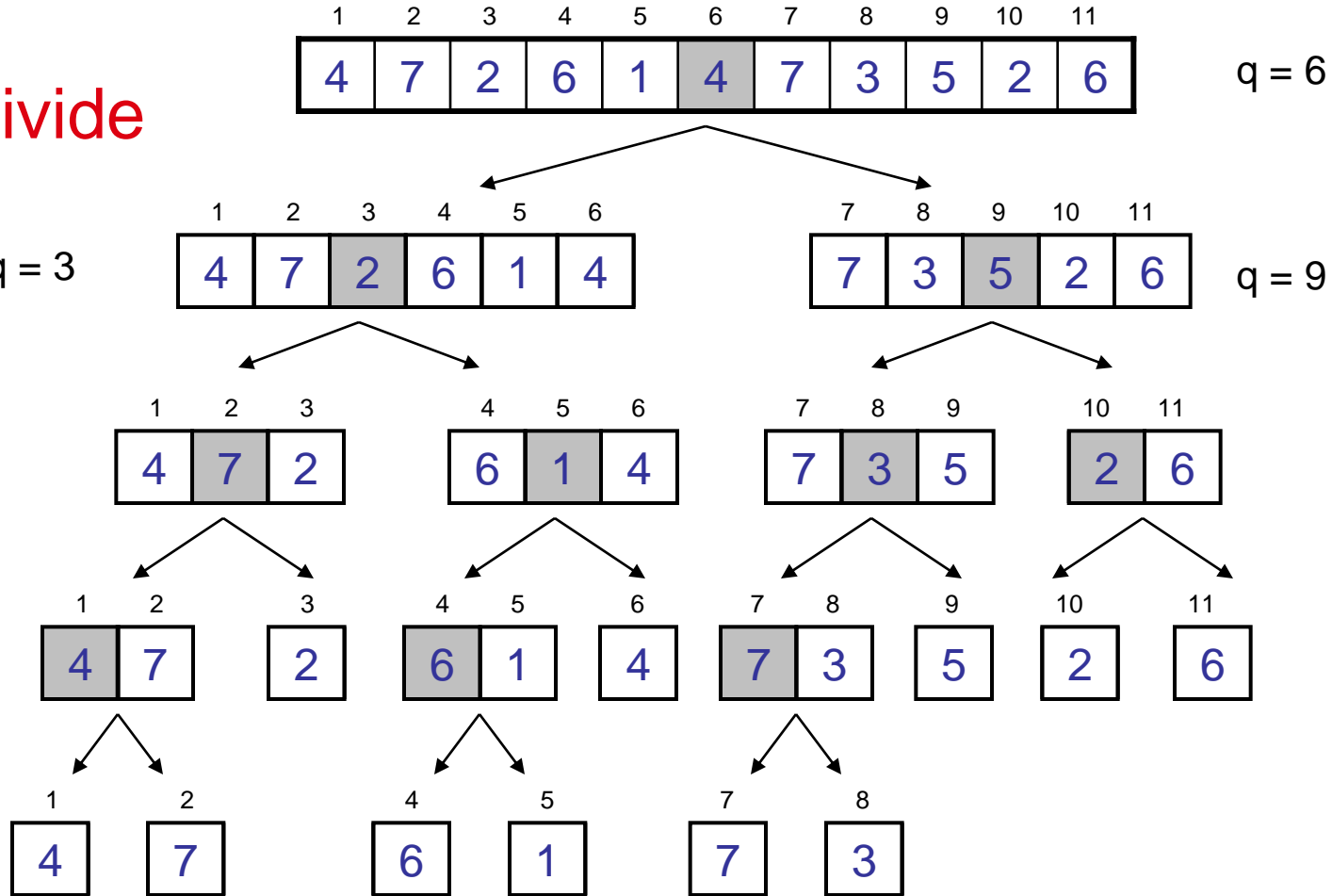
# Example – n Power of 2

Conquer  
and  
Merge



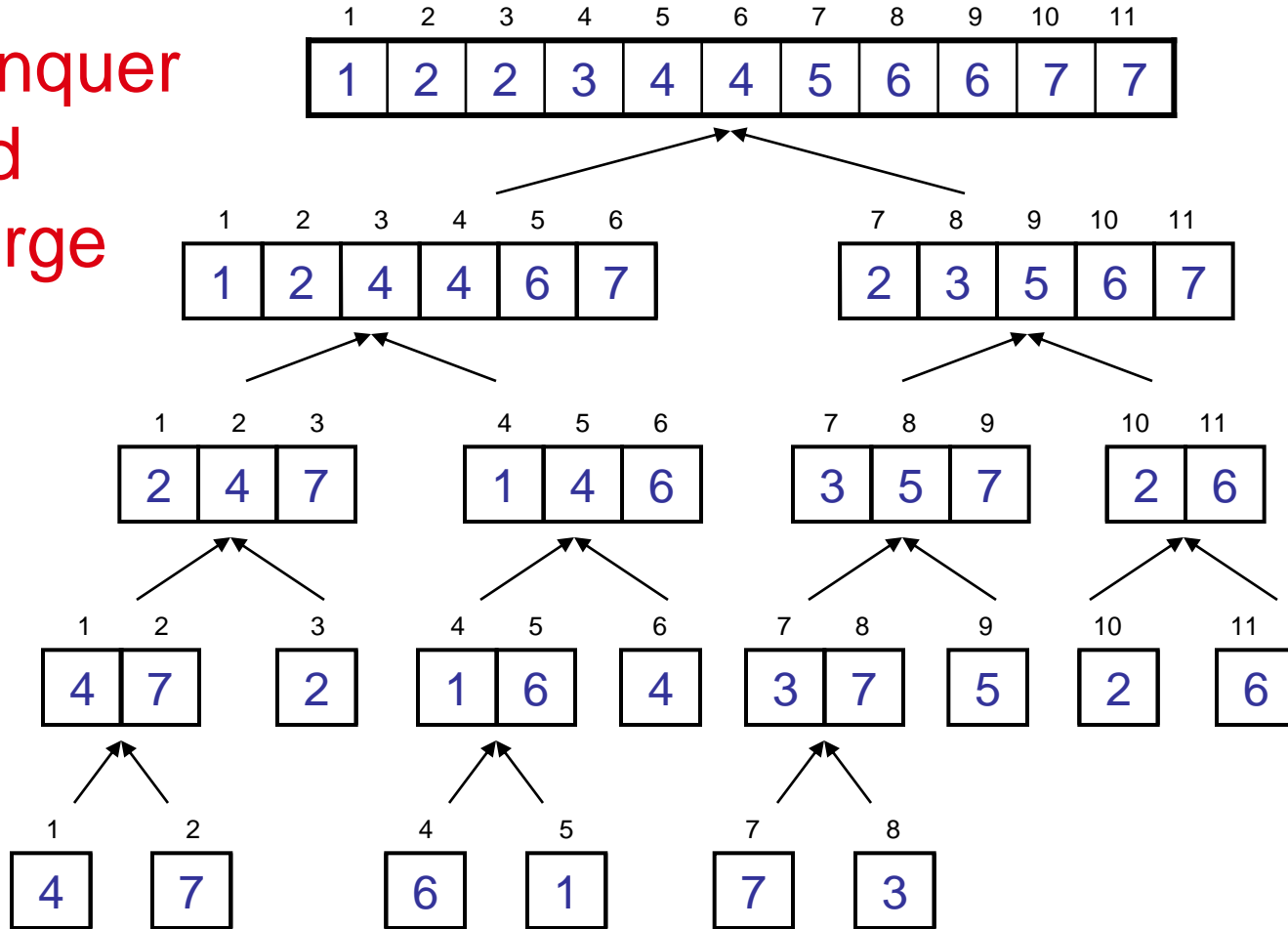
# Example – n Not a Power of 2

Divide



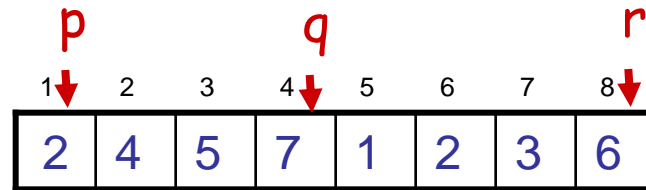
# Example – n Not a Power of 2

Conquer  
and  
Merge



# Merging

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- **Input:** Array  $A$  and indices  $p, q, r$  such that  $p \leq q < r$ 
  - Subarrays  $A[p \dots q]$  and  $A[q + 1 \dots r]$  are sorted
- **Output:** One single sorted subarray  $A[p \dots r]$



# Merging

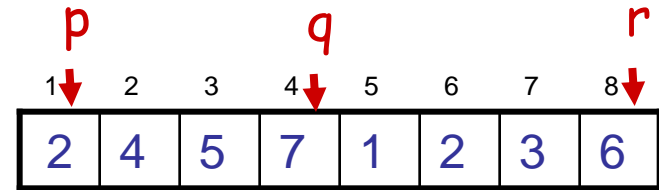
- Idea for merging:

- Two piles of sorted cards

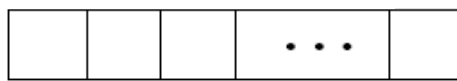
- Choose the smaller of the two top cards
- Remove it and place it in the output pile

- Repeat the process until one pile is empty

- Take the remaining input pile and place it face-down onto the output pile



$A_1 \leftarrow A[p, q]$



$A_2 \leftarrow A[q+1, r]$

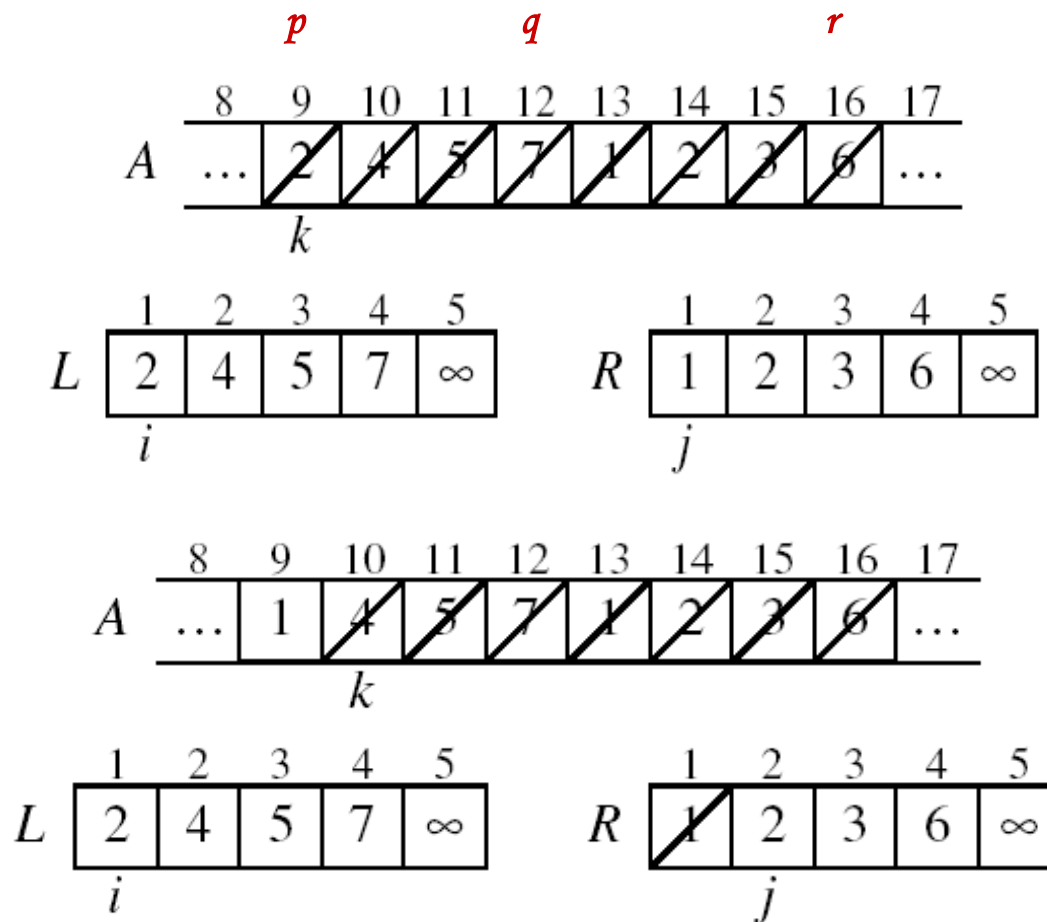


choose the smaller  
element from the subarrays

$A[p, r]$

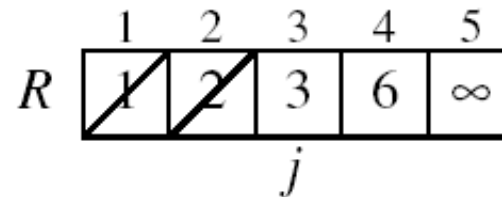
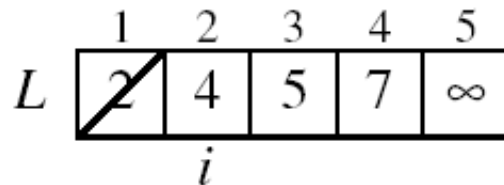
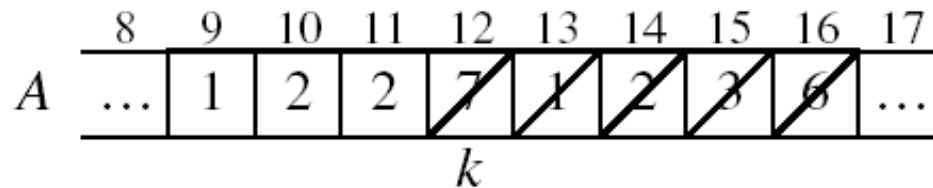
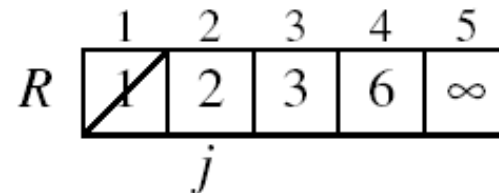
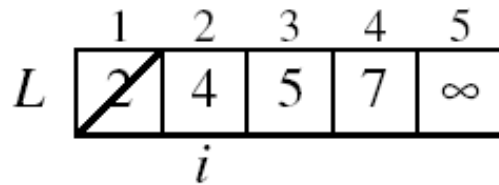
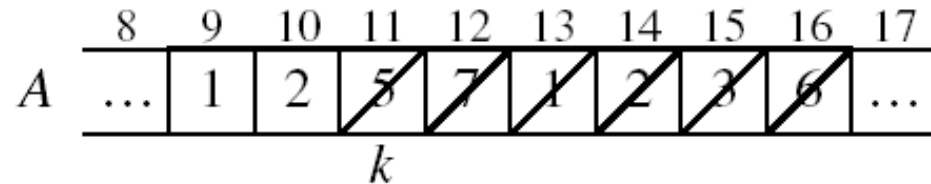


# Example: MERGE(A, 9, 12, 16)



# Example: MERGE(A, 9, 12, 16)

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# Example (cont.)

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	8	9	10	11	12	13	14	15	16	17
A	...	1	2	2	3	/	/	/	/	...

$k$

$L$	1	2	3	4	5
	/	4	5	7	∞

$i$

$R$	1	2	3	4	5
	/	/	/	6	∞

$j$

	8	9	10	11	12	13	14	15	16	17
A	...	1	2	2	3	4	/	/	/	...

$k$

$L$	1	2	3	4	5
	/	/	5	7	∞

$i$

$R$	1	2	3	4	5
	/	/	/	6	∞

$j$

# Example (cont.)

	8	9	10	11	12	13	14	15	16	17
A	...	1	2	2	3	4	5	<del>3</del>	<del>6</del>	...
								$k$		

$L$	1	2	3	4	5
	<del>2</del>	<del>4</del>	<del>5</del>	7	$\infty$
			$i$		

$R$	1	2	3	4	5
	<del>1</del>	<del>2</del>	<del>3</del>	6	$\infty$
				$j$	

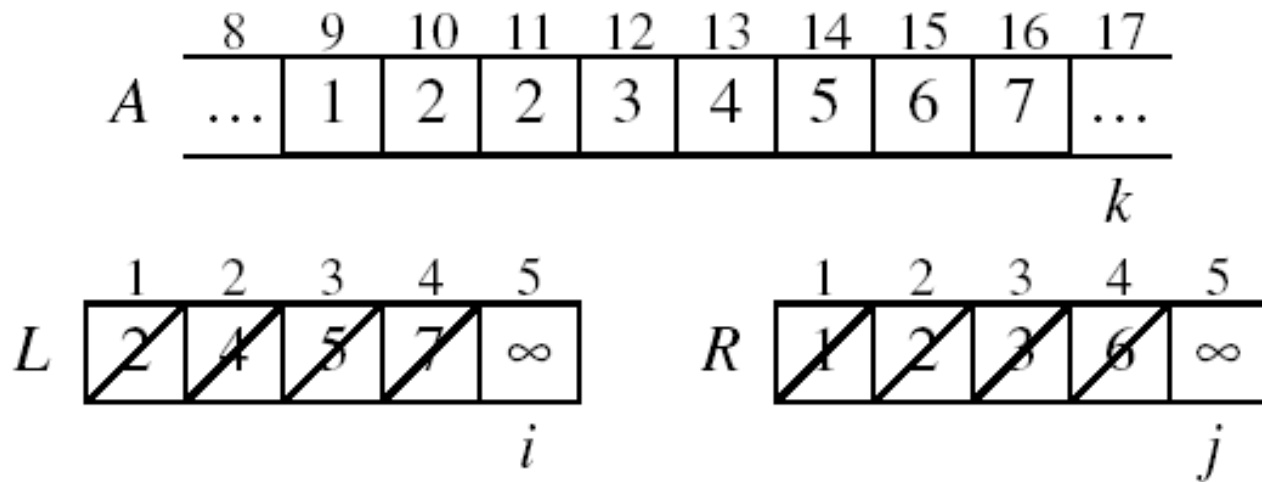
	8	9	10	11	12	13	14	15	16	17
A	...	1	2	2	3	4	5	6	<del>6</del>	...
									$k$	

$L$	1	2	3	4	5
	<del>2</del>	<del>4</del>	<del>5</del>	7	$\infty$
			$i$		

$R$	1	2	3	4	5
	<del>1</del>	<del>2</del>	<del>3</del>	<del>6</del>	$\infty$
				$j$	

# Example (cont.)

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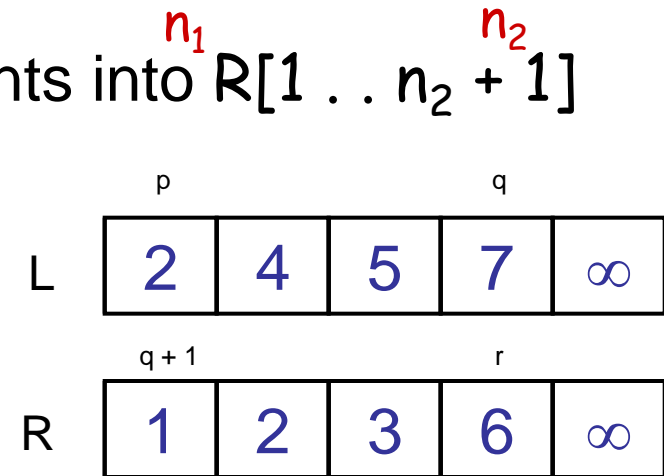
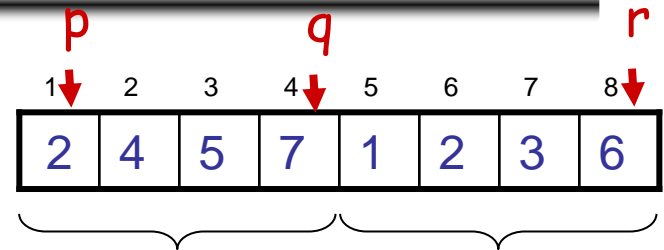


Done!

# Merge - Pseudocode

*Alg.:* MERGE(A, p, q, r)

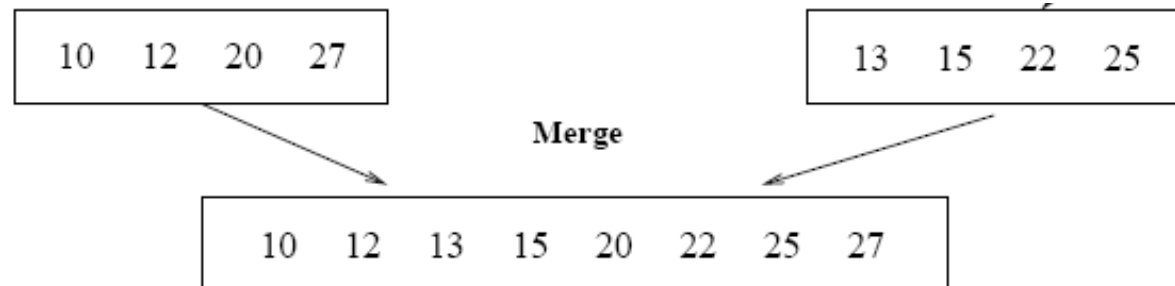
1. Compute  $n_1$  and  $n_2$
2. Copy the first  $n_1$  elements into  $L[1 \dots n_1 + 1]$  and the next  $n_2$  elements into  $R[1 \dots n_2 + 1]$
3.  $L[n_1 + 1] \leftarrow \infty$ ;  $R[n_2 + 1] \leftarrow \infty$
4.  $i \leftarrow 1$ ;  $j \leftarrow 1$
5. **for**  $k \leftarrow p$  **to**  $r$
6.     **do if**  $L[i] \leq R[j]$
7.         **then**  $A[k] \leftarrow L[i]$
8.          $i \leftarrow i + 1$
9.     **else**  $A[k] \leftarrow R[j]$
10.      $j \leftarrow j + 1$



# Running Time of Merge (assume last **for** loop)

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- Initialization (copying into temporary arrays):
  - $\Theta(n_1 + n_2) = \Theta(n)$
- Adding the elements to the final array:
  - $n$  iterations, each taking constant time  $\Rightarrow \Theta(n)$
- Total time for Merge:
  - $\Theta(n)$





# Analyzing Divide-and Conquer Algorithms

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- The recurrence is based on the three steps of the paradigm:
  - $T(n)$  – running time on a problem of size  $n$
  - **Divide** the problem into  $a$  subproblems, each of size  $n/b$ : takes  $D(n)$
  - **Conquer** (solve) the subproblems  $aT(n/b)$
  - **Combine** the solutions  $C(n)$

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

# MERGE-SORT Running Time

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- **Divide:**

- compute  $q$  as the average of  $p$  and  $r$ :  $D(n) = \Theta(1)$

- **Conquer:**

- recursively solve 2 subproblems, each of size  $n/2$   
 $\Rightarrow 2T(n/2)$

- **Combine:**

- MERGE on an  $n$ -element subarray takes  $\Theta(n)$  time  
 $\Rightarrow C(n) = \Theta(n)$

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

# Solve the Recurrence

---

$$T(n) = \begin{cases} c & \text{if } n = 1 \\ 2T(n/2) + cn & \text{if } n > 1 \end{cases}$$

Use Master's Theorem:

Compare  $n$  with  $f(n) = cn$

Case 2:  $T(n) = \Theta(n \lg n)$

# Merge Sort - Discussion

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- Running time insensitive of the input
- Advantages:
  - Guaranteed to run in  $\Theta(n \lg n)$
- Disadvantage
  - Requires extra space  $\approx N$

# Sorting Challenge 1

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**Problem:** Sort a file of huge records with tiny keys

Example application: Reorganize your MP-3 files

**Which method to use?**

- A. merge sort, guaranteed to run in time  $\sim N \lg N$
- B. selection sort
- C. bubble sort
- D. a custom algorithm for huge records/tiny keys
- E. insertion sort

# Sorting Files with Huge Records and Small Keys

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- Insertion sort or bubble sort?
  - NO, too many exchanges
- Selection sort?
  - YES, it takes **linear** time for exchanges
- Merge sort or custom method?
  - Probably not: selection sort simpler, does less swaps

# Sorting Challenge 2

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**Problem:** Sort a huge randomly-ordered file of small records

**Application:** Process transaction record for a phone company

**Which sorting method to use?**

- A. Bubble sort
- B. Selection sort
- C. Mergesort guaranteed to run in time  $\sim N \lg N$
- D. Insertion sort

# Sorting Huge, Randomly - Ordered Files

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- Selection sort?
  - NO, always takes quadratic time
- Bubble sort?
  - NO, quadratic time for randomly-ordered keys
- Insertion sort?
  - NO, quadratic time for randomly-ordered keys
- Mergesort?
  - YES, it is designed for this problem



# Sorting Challenge 3

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**Problem:** sort a file that is already almost in order

Applications:

- Re-sort a huge database after a few changes
- Doublecheck that someone else sorted a file

**Which sorting method to use?**

- A. Mergesort, guaranteed to run in time  $\sim N \lg N$
- B. Selection sort
- C. Bubble sort
- D. A custom algorithm for almost in-order files
- E. Insertion sort

# Sorting Files That are Almost in Order

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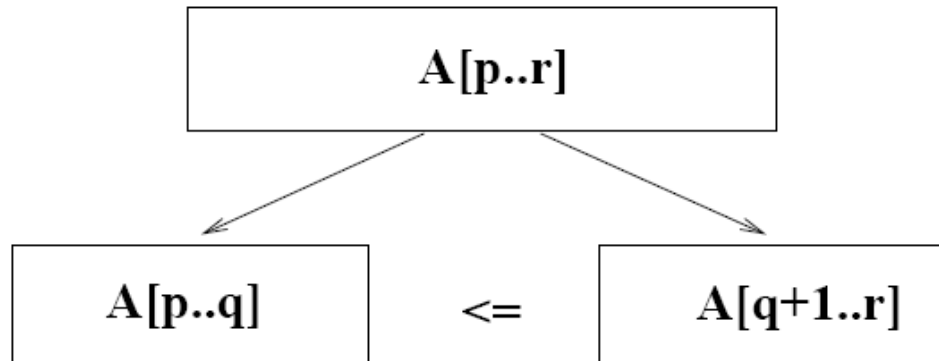
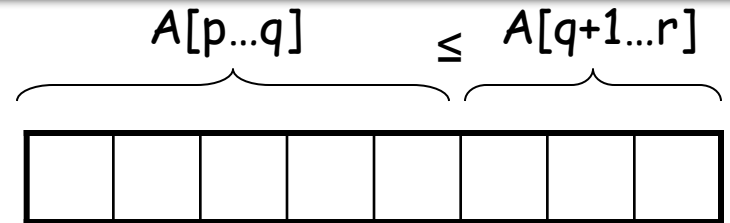
- Selection sort?
  - NO, always takes quadratic time
- Bubble sort?
  - NO, bad for some definitions of “almost in order”
  - Ex: B C D E F G H I J K L M N O P Q R S T U V W X Y Z A
- Insertion sort?
  - YES, takes linear time for most definitions of “almost in order”
- Mergesort or custom method?
  - Probably not: insertion sort simpler and faster

# Quicksort

- Sort an array  $A[p..r]$

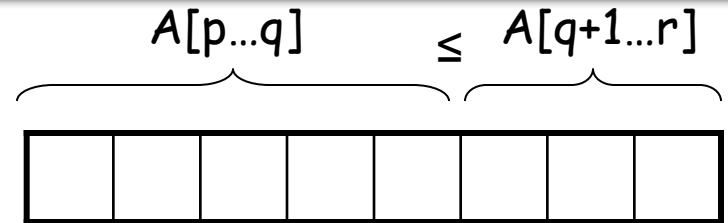
- **Divide**

- Partition the array  $A$  into 2 subarrays  $A[p..q]$  and  $A[q+1..r]$ , such that each element of  $A[p..q]$  is smaller than or equal to each element in  $A[q+1..r]$
- Need to find index  $q$  to partition the array



# Quicksort

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- **Conquer**

- Recursively sort  $A[p\dots q]$  and  $A[q+1\dots r]$  using Quicksort

- **Combine**

- Trivial: the arrays are sorted in place
- No additional work is required to combine them
- The entire array is now sorted

# QUICKSORT

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*Alg.:* QUICKSORT( $A, p, r$ )

Initially:  $p=1, r=n$

**if**  $p < r$

**then**  $q \leftarrow \text{PARTITION}(A, p, r)$

    QUICKSORT ( $A, p, q$ )

    QUICKSORT ( $A, q+1, r$ )

Recurrence:

$$T(n) = T(q) + T(n - q) + f(n) \quad (f(n) \text{ depends on } \text{PARTITION}())$$

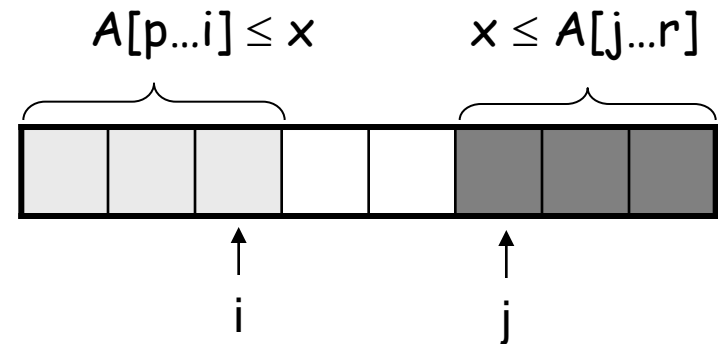
# Partitioning the Array

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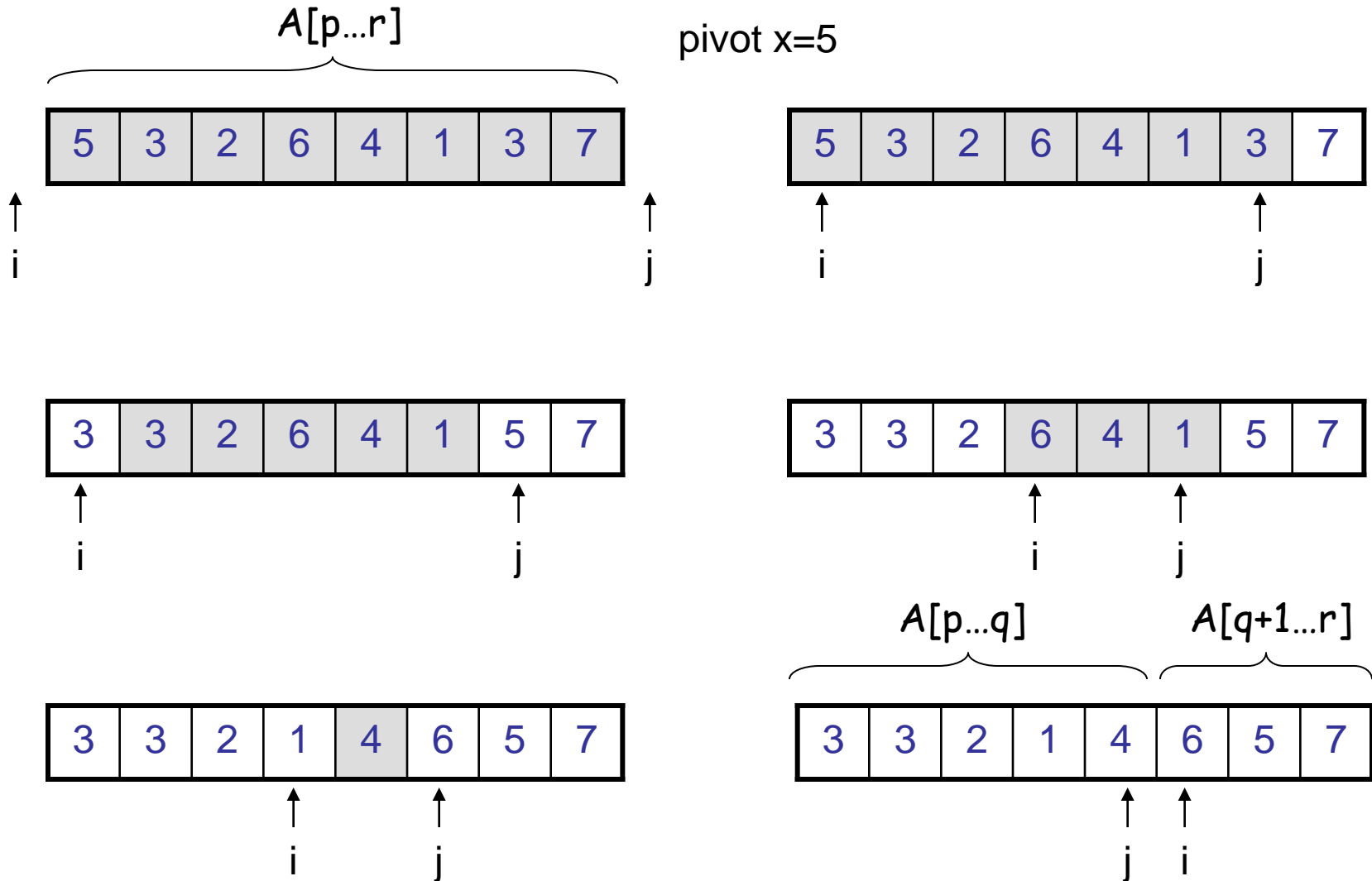
- Choosing PARTITION()
  - There are different ways to do this
  - Each has its own advantages/disadvantages
- Hoare partition (see prob. 7-1, page 159)
  - Select a pivot element  $x$  around which to partition
  - Grows two regions

$$A[p\dots i] \leq x$$

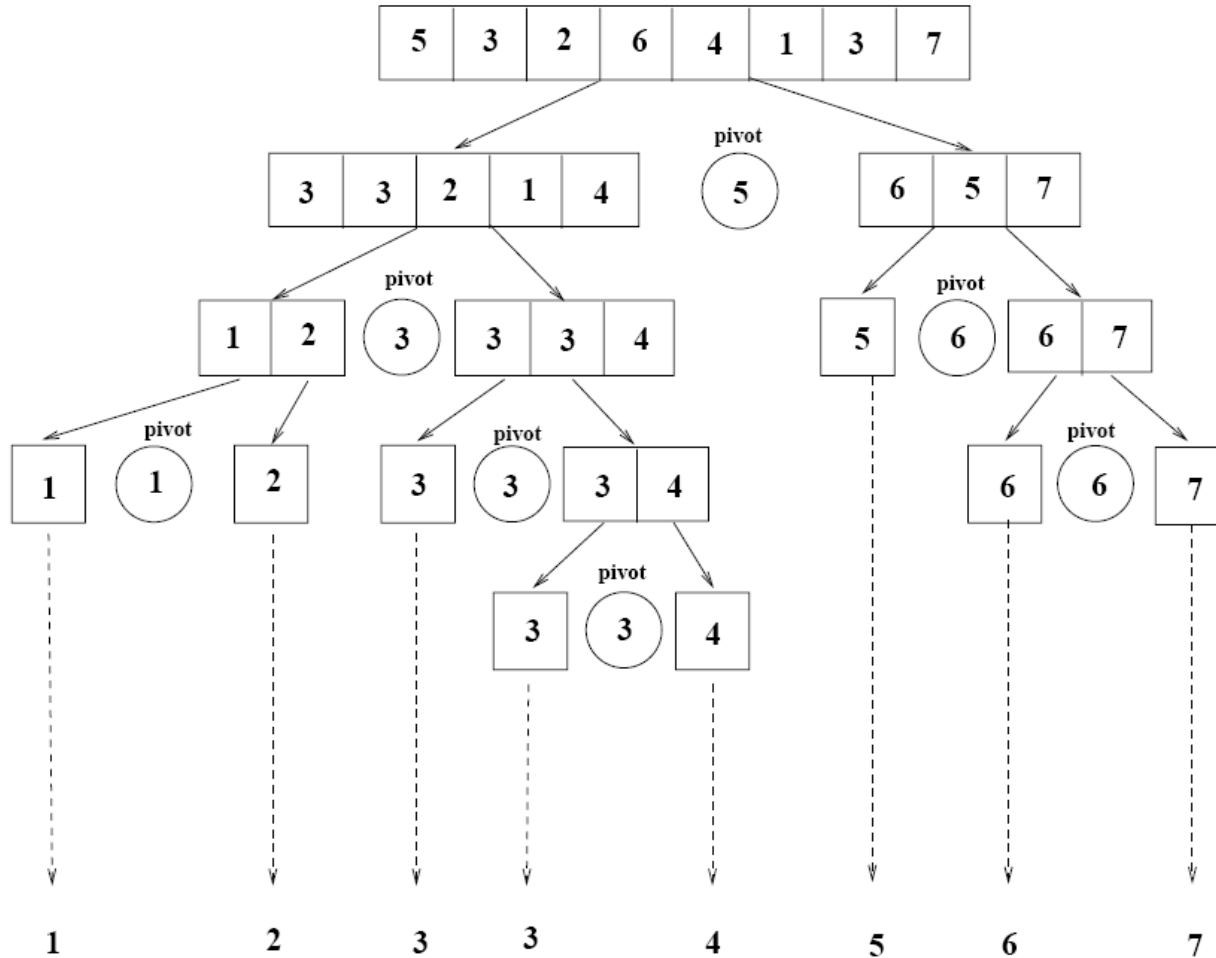
$$x \leq A[j\dots r]$$



# Example



# Example

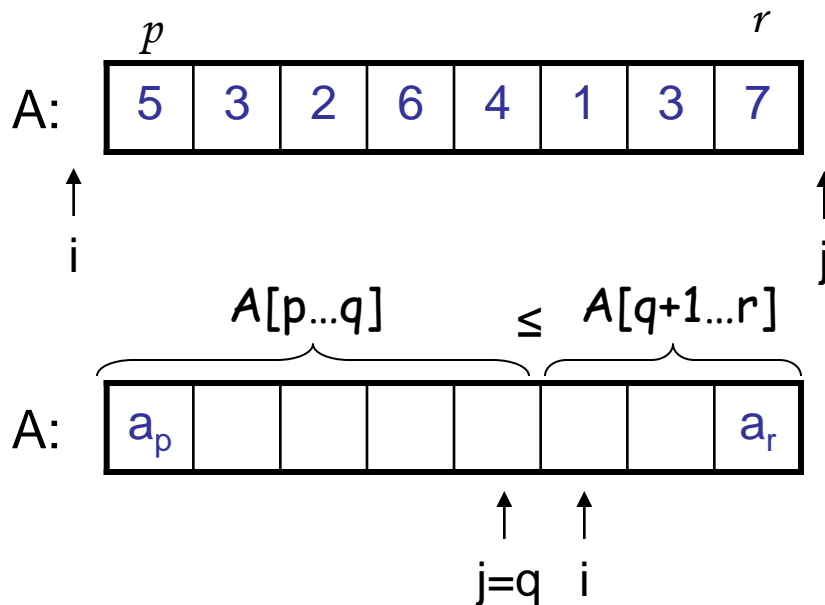




# Partitioning the Array

*Alg.* PARTITION ( $A, p, r$ )

1.  $x \leftarrow A[p]$
2.  $i \leftarrow p - 1$
3.  $j \leftarrow r + 1$
4. **while** TRUE
5.     **do repeat**  $j \leftarrow j - 1$
6.         **until**  $A[j] \leq x$
7.     **do repeat**  $i \leftarrow i + 1$
8.         **until**  $A[i] \geq x$
9.     **if**  $i < j$
10.         **then** exchange  $A[i] \leftrightarrow A[j]$
11.     **else return**  $j$



Each element is visited once!

Running time:  $\Theta(n)$   
 $n = r - p + 1$

# Recurrence

---

*Alg.*: QUICKSORT( $A, p, r$ )

Initially:  $p=1, r=n$

**if**  $p < r$

**then**  $q \leftarrow$  PARTITION( $A, p, r$ )

    QUICKSORT ( $A, p, q$ )

    QUICKSORT ( $A, q+1, r$ )

Recurrence:

$$T(n) = T(q) + T(n - q) + n$$

# Worst Case Partitioning

- Worst-case partitioning

- One region has one element and the other has  $n - 1$  elements
- Maximally unbalanced

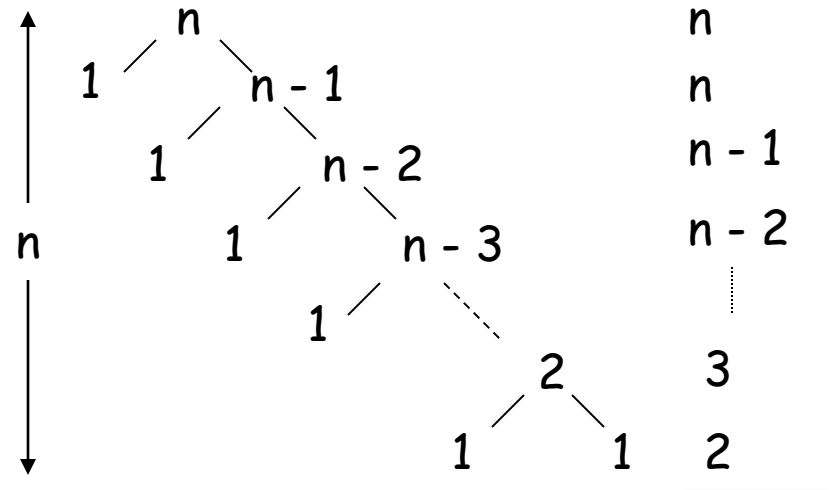
- Recurrence:  $q=1$

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

$$T(1) = \Theta(1)$$

$$T(n) = T(n - 1) + n$$

$$= n + \left( \sum_{k=1}^n k \right) - 1 = \Theta(n) + \Theta(n^2) = \Theta(n^2)$$



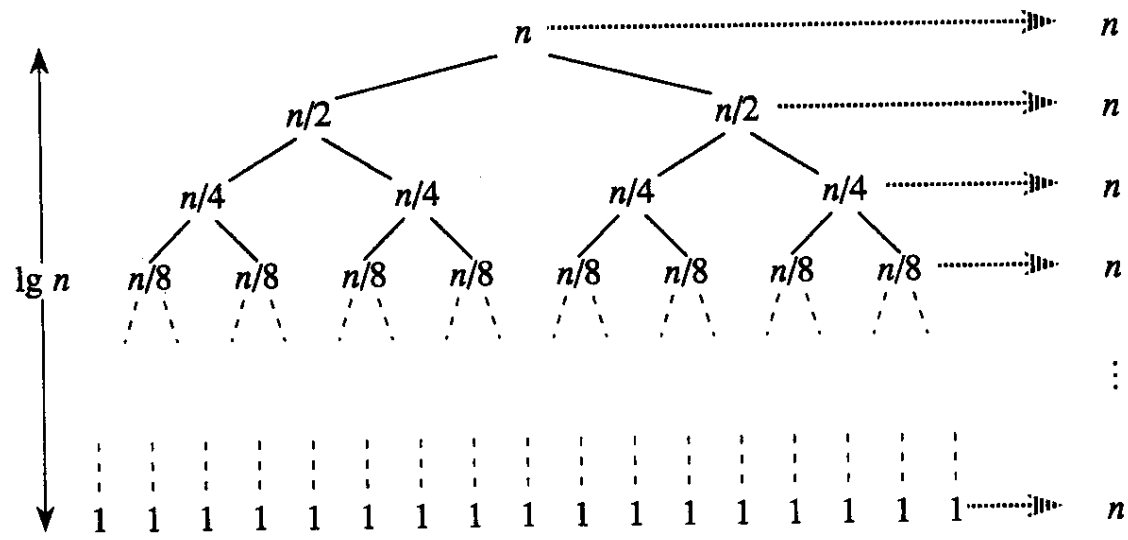
When does the worst case happen?

# Best Case Partitioning

- Best-case partitioning
  - Partitioning produces two regions of size  $n/2$
- Recurrence:  $q=n/2$

$$T(n) = 2T(n/2) + \Theta(n)$$

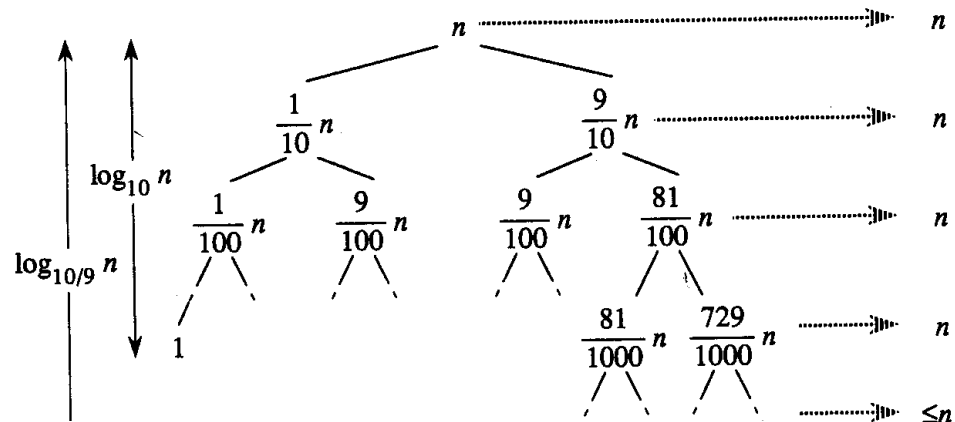
$$T(n) = \Theta(n \lg n) \text{ (Master theorem)}$$



# Case Between Worst and Best

- 9-to-1 proportional split

$$Q(n) = Q(9n/10) + Q(n/10) + n$$



- Using the recursion tree:

$$\text{longest path: } Q(n) \leq n \sum_{i=0}^{\log_{10/9} n} 1 = n(\log_{10/9} n + 1) = c_2 n \lg n \quad \Theta(n \lg n)$$

$$\text{shortest path: } Q(n) \geq n \sum_{i=0}^{\log_{10} n} 1 = n \log_{10} n = c_1 n \lg n$$

Thus,  $Q(n) = \Theta(n \lg n)$

# How does partition affect performance?

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- **Any splitting of constant proportionality** yields  $\Theta(n \lg n)$  time !!!

- Consider the  $(1 : n - 1)$  splitting:

$$\text{ratio} = 1/(n - 1) \text{ not a constant !!!}$$

- Consider the  $(n/2 : n/2)$  splitting:

$$\text{ratio} = (n/2)/(n/2) = 1 \text{ it is a constant !!}$$

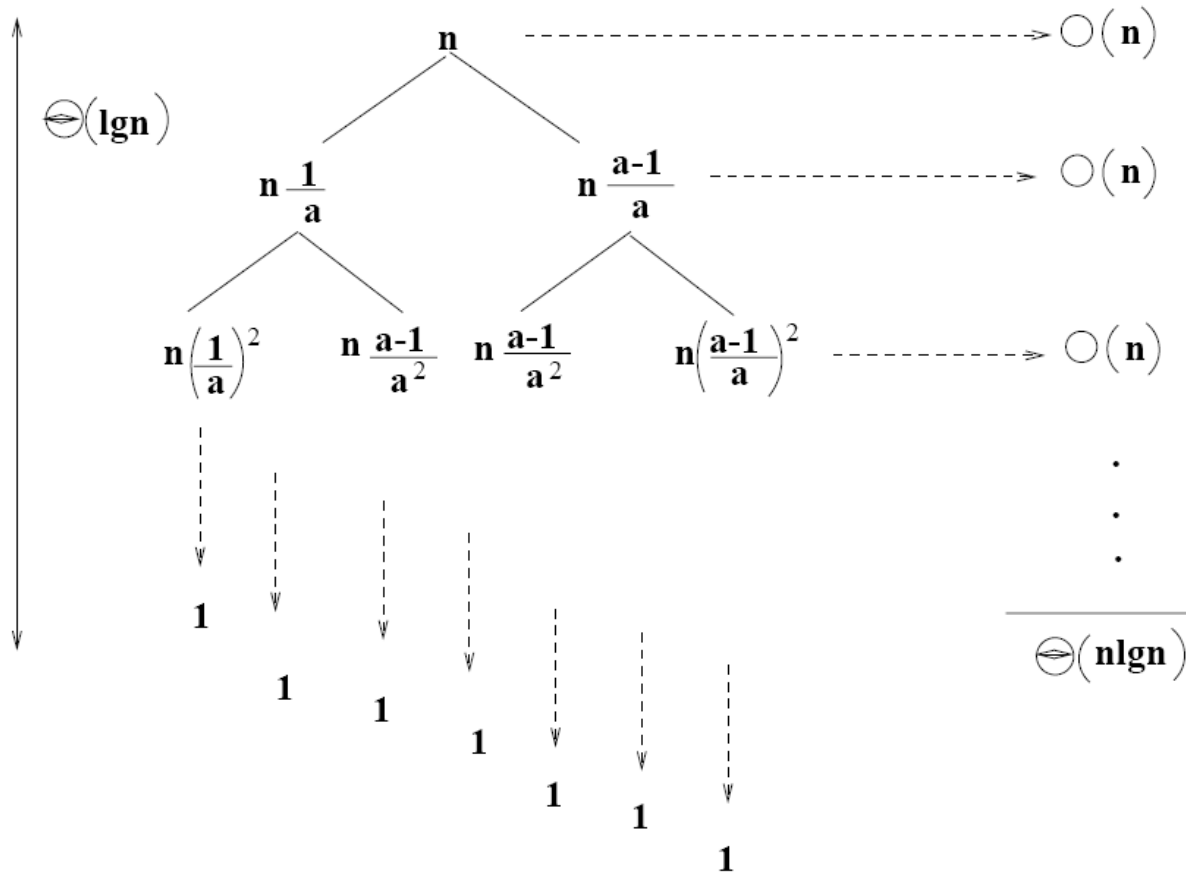
- Consider the  $(9n/10 : n/10)$  splitting:

$$\text{ratio} = (9n/10)/(n/10) = 9 \text{ it is a constant !!}$$

# How does partition affect performance?

- Any  $((a - 1)n/a : n/a)$  splitting:

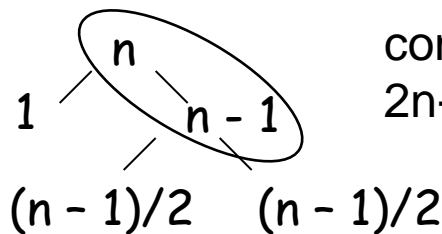
ratio= $((a - 1)n/a)/(n/a) = a - 1$  it is a constant !!



# Performance of Quicksort

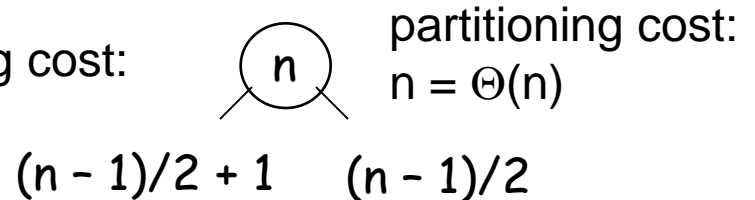
- Average case

- All permutations of the input numbers are equally likely
- On a random input array, we will have a **mix** of well balanced and unbalanced splits
- Good and bad splits are randomly distributed across throughout the tree



Alternate of a good and a bad split

combined partitioning cost:  
 $2n-1 = \Theta(n)$



Nearly well balanced split

- Running time of Quicksort when levels alternate between good and bad splits is  $O(n \lg n)$