

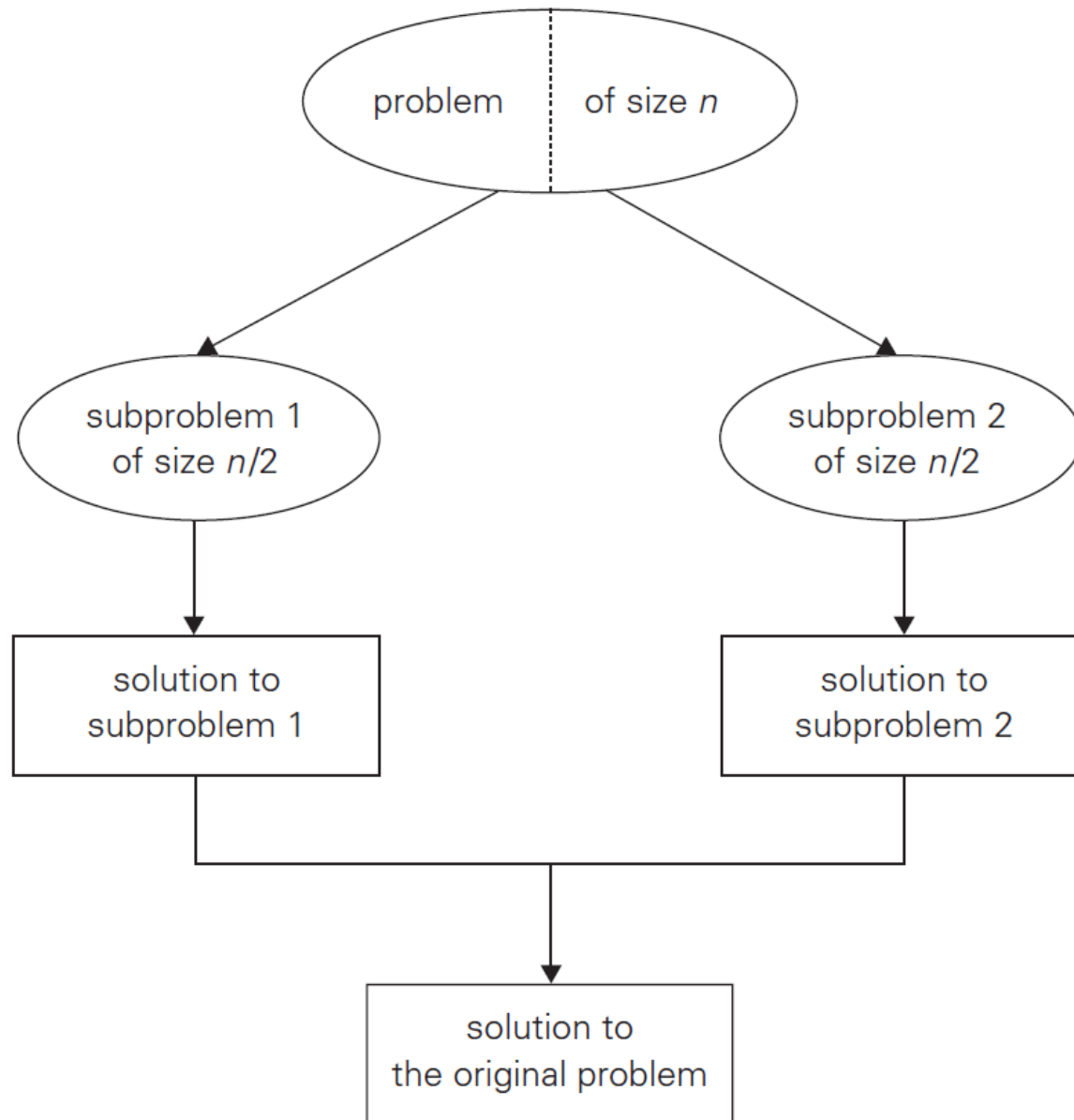
Divide and Conquer

(Chapter 5)

This week:

- Divide and Conquer technique
- Count a specific key in an array
- Master theorem
- Merge sort

Divide and Conquer technique

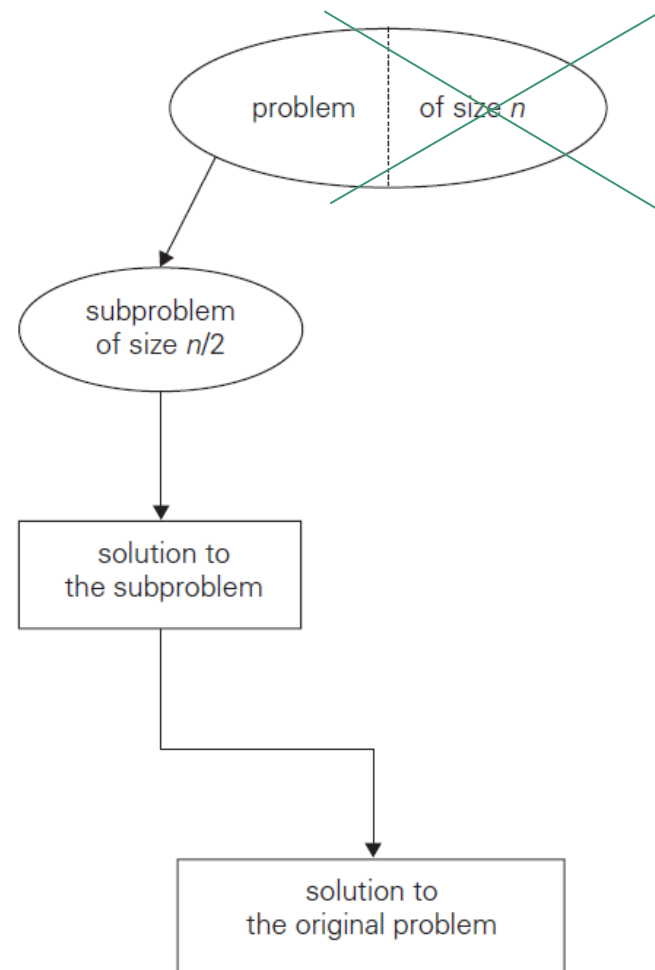


Divide and Conquer technique

A well known algorithm design technique:

1. **Divide** instance of problem into two or more smaller instances
2. **Solve** smaller instances (usually recursively)
3. Obtain solution to original (larger) instance by **combining** these solutions

Decrease and Conquer (last week)



A Natural Question

- How is this different from decrease and conquer technique
- Think of the fake coin problem:
 - We discarded half the coins at each step
 - So we didn't do any work on those “sub problems”
- For divide and conquer...
 - You need to solve all of the sub problems

This week:

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Count a specific key in an array

Problem:

- Count the number of times a specific key occurs in an array.

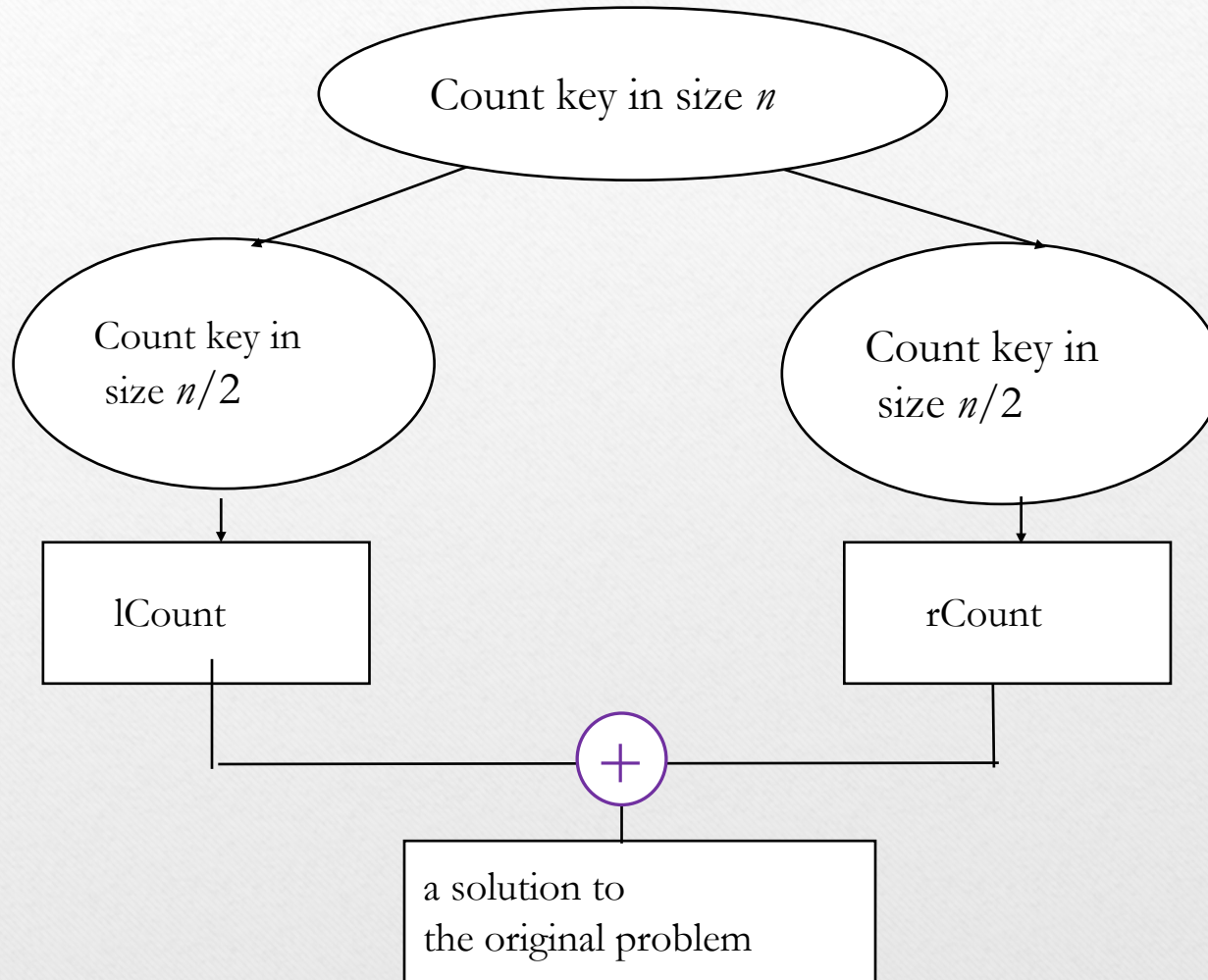
For example:

- If input array is $A=[2,7,6,6,2,4,6,9,2]$ and $key=6$...
... should return the value 3.

Design an algorithm using divide and conquer technique

Whiteboard

Count a specific key in an array



Count a specific key in an array

Algorithm CountKey(A[], L, R, Key)

//Input: A[] is an array A[0..n-1] from indices L and R ($L \leq R$)

//Output: A count of the number of time Key exists in A[L..R]

```
1.  if L = R
2.      if (A[L] = Key) return 1
3.      else return 0
4.  else
5.      lCount = CountKey(A[], L, ⌊(L+R)/2⌋, Key)
6.      rCount = CountKey(A[], ⌊(L+R)/2⌋+1, R, Key)
7.      return lCount + rCount
```

Big O? All in good time...

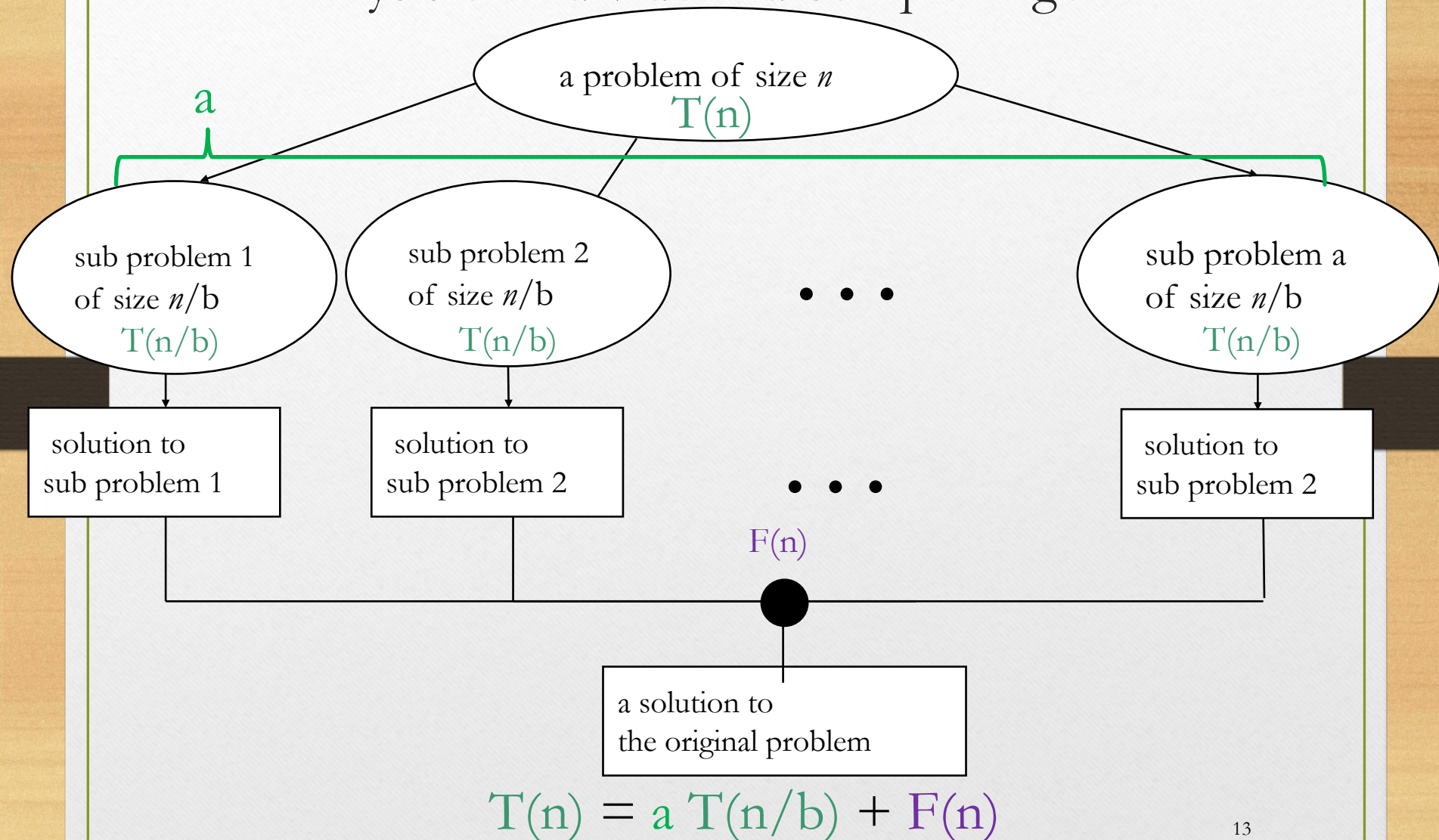
Count a specific key in an array

- *CountKey* looks familiar...
 - ▶ What's the difference between *Binary Search* and *CountKey*?
- We have to searched both sides
 - In Binary Search, one half gets ignored if out of bounds
 - In CountKey, both sides are searched to get sum

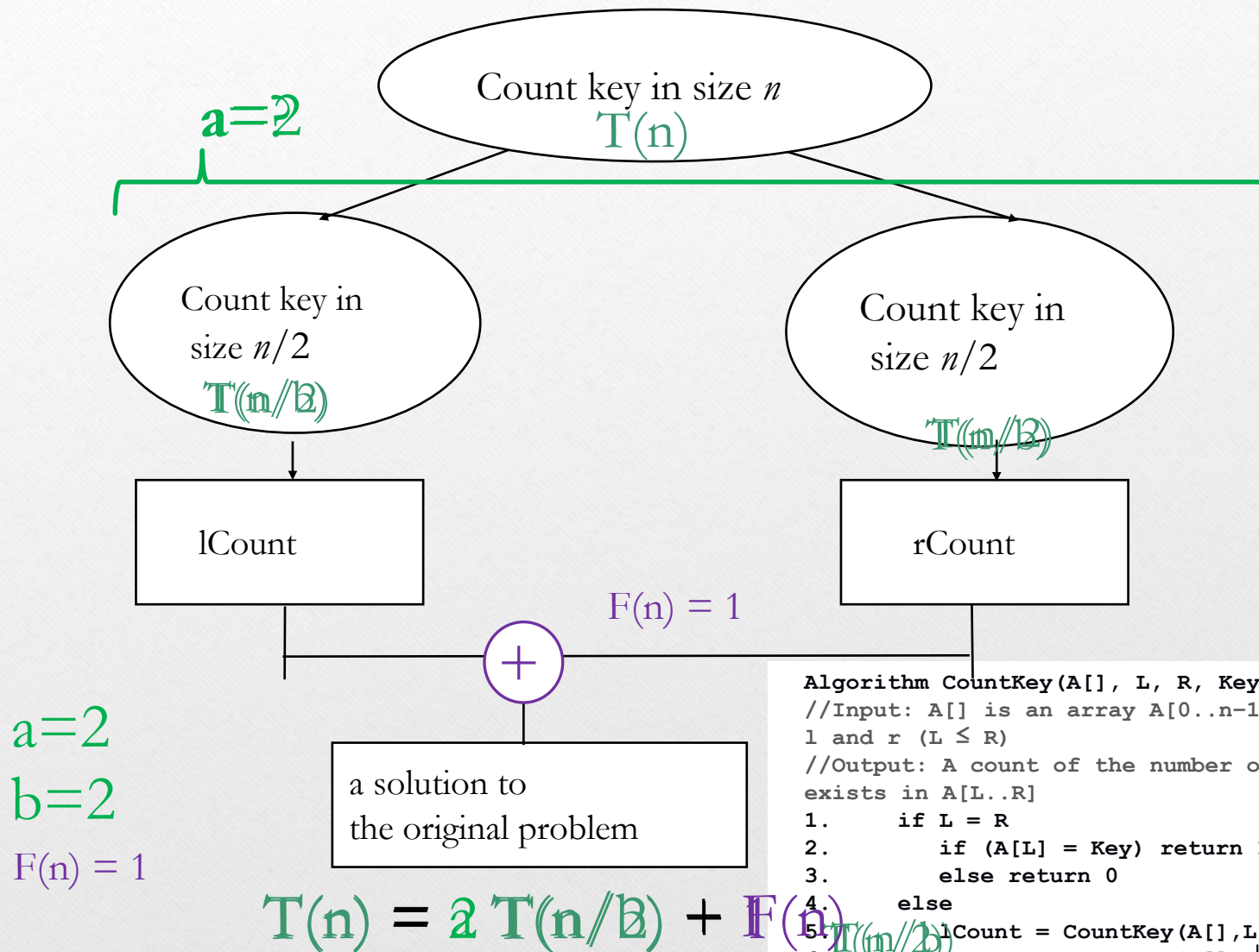
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Analysis of a divide and conquer algorithm



Analysis of Count a specific key in an array



```

Algorithm CountKey(A[], L, R, Key)
//Input: A[] is an array A[0..n-1] from indices
1 and r ( $L \leq R$ )
//Output: A count of the number of time Key
exists in A[L..R]
1.   if  $L = R$ 
2.       if ( $A[L] = \text{Key}$ ) return 1
3.       else return 0
4.   else
5.       lCount = CountKey(A[], L,  $\lfloor (L+R)/2 \rfloor$ , Key)
6.       rCount = CountKey(A[],  $\lfloor (L+R)/2 \rfloor + 1$ , R,
           Key)
7.       return lCount + rCount
    
```

$F(n) = 1$

Master Theorem

$$T(n) = a T(n/b) + F(n)$$

- 1) If $n^{\log_b a} < F(n)$, $T(n) \in O(F(n))$
- 2) If $n^{\log_b a} > F(n)$, $T(n) \in O(n^{\log_b a})$
- 3) If $n^{\log_b a} = F(n)$, $T(n) \in O(n^{\log_b a} \log n)$

Add to Cheat Sheet

Count Keys: $T(n) = 2T(n/2) + 1 \Rightarrow T(n) \in ?$

$$\begin{aligned} a &= 2 \\ b &= 2 \end{aligned}$$



$$n^{\log_b a}$$



$$n^{\log_2 2}$$



$$n$$

$$F(n) = 1$$



$$T(n) \in O(n^{\log_b a})$$

$$T(n) \in O(n^1)$$

Master Theorem

$$T(n) = a T(n/b) + F(n)$$

- 1) If $n^{\log_b a} < F(n)$, $T(n) \in O(F(n))$
- 2) If $n^{\log_b a} > F(n)$, $T(n) \in O(n^{\log_b a})$
- 3) If $n^{\log_b a} = F(n)$, $T(n) \in O(n^{\log_b a} \log n)$

Binary Search: $T(n) = T(n/2) + 1 \Rightarrow T(n) \in ?$

$$\begin{aligned} a &= 1 \\ b &= 2 \end{aligned}$$



$$n^{\log_b a}$$



$$n^{\log_2 1}$$



$$1^2$$

$$F(n) = 1$$



$$T(n) \in O(n^{\log_b a} \log n)$$

$$T(n) \in O(\log n)$$

```

binarySearch(a[], k, start, end)
    if end < start
        return not found
    middle ← floor((start+end)/2)
    if k = a[middle]
        return found
    if k > a[middle]
        return binarySearch(a[], k, middle+1, end)
    else if k < a[middle]
        return binarySearch(a[], k, start, middle-1)
    
```


Master Theorem

$$T(n) = a T(n/b) + F(n)$$

- 1) If $n^{\log_b a} < F(n)$, $T(n) \in O(F(n))$
- 2) If $n^{\log_b a} > F(n)$, $T(n) \in O(n^{\log_b a})$
- 3) If $n^{\log_b a} = F(n)$, $T(n) \in O(n^{\log_b a} \log n)$

Random Algorithm: $T(n) = 4T(n/2) + n^3 \Rightarrow T(n) \in ?$

$$\begin{aligned} a &= 4 \\ b &= 2 \end{aligned}$$



$$n^{\log_b a}$$



$$n^{\log_2 4}$$



$$n^2$$

$$F(n) = n^3$$



$$T(n) \in O(n^3)$$

Alg(n)

Alg(n/2)

Alg(n/2)

Alg(n/2)

Alg(n/2)

for each i in n

for each j in n

for each k in n

do something

Master theorem

Example 2: $T(n) = 4T(n/2) + n \Rightarrow T(n) \in ?$

$$\begin{array}{l} a = 4 \\ b = 2 \end{array} \quad \begin{array}{c} \longrightarrow \\ \longrightarrow \\ \longrightarrow \end{array} \begin{array}{c} n^{\log_b a} \\ n^{\log_2 4} \\ n^2 \end{array} \quad \left. \begin{array}{c} \\ \\ F(n) = n \end{array} \right\} \longrightarrow T(n) \in O(n^2)$$

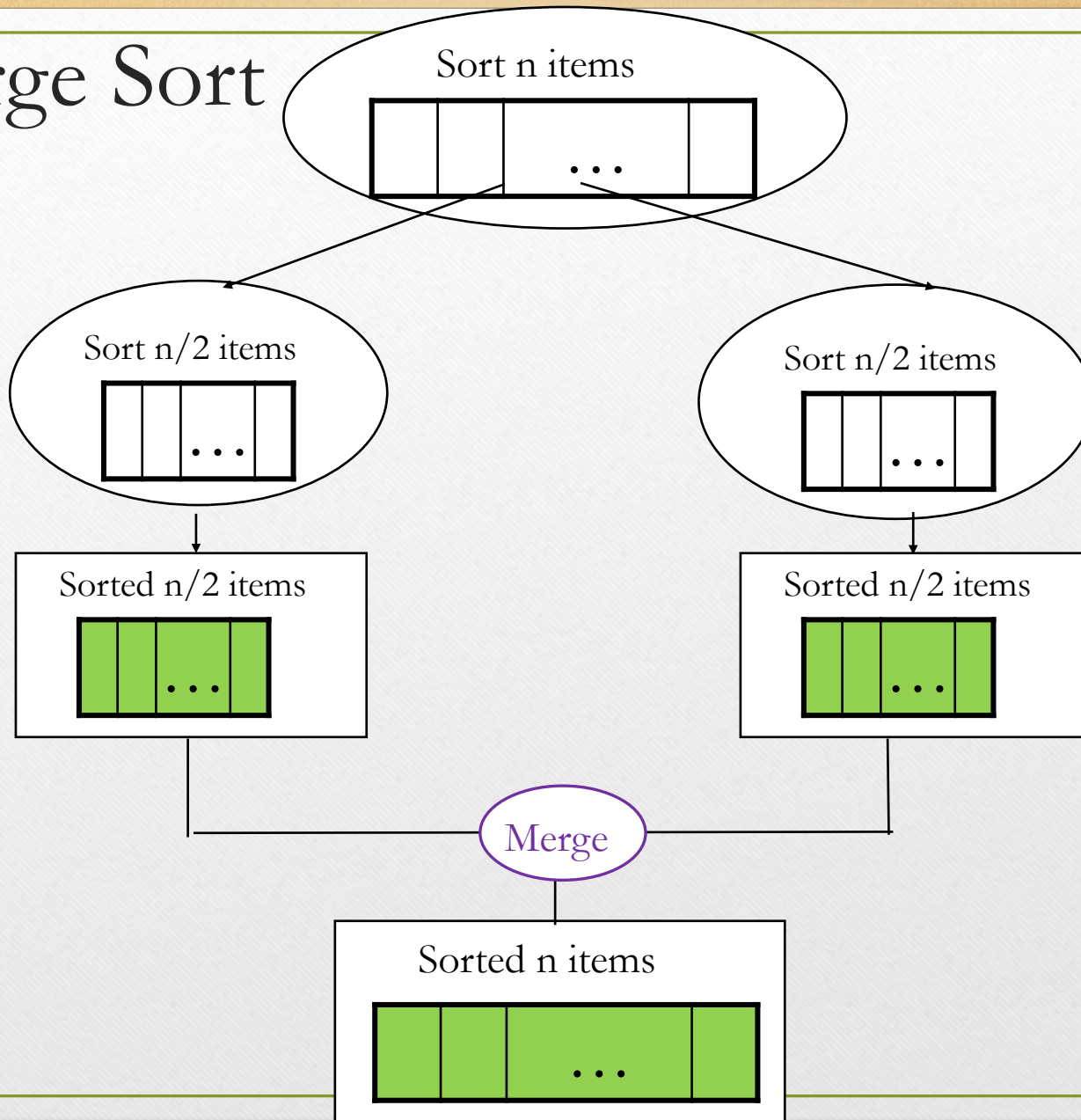
Example 3: $T(n) = 4T(n/2) + n^2 \Rightarrow T(n) \in ?$

$$\begin{array}{l} a = 4 \\ b = 2 \end{array} \quad \begin{array}{c} \longrightarrow \\ \longrightarrow \\ \longrightarrow \end{array} \begin{array}{c} n^{\log_b a} \\ n^{\log_2 4} \\ n^2 \end{array} \quad \left. \begin{array}{c} \\ \\ F(n) = n^2 \end{array} \right\} \longrightarrow T(n) \in O(n^2 \log n)$$

This week:

- Divide and Conquer technique
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- Merge sort

Merge Sort



Pseudocode of Mergesort

ALGORITHM *Mergesort*($A[0..n - 1]$)

//Sorts array $A[0..n - 1]$ by recursive mergesort

//Input: An array $A[0..n - 1]$ of orderable elements

//Output: Array $A[0..n - 1]$ sorted in nondecreasing order

if $n > 1$

 copy $A[0..\lfloor n/2 \rfloor - 1]$ to $B[0..\lfloor n/2 \rfloor - 1]$

 copy $A[\lfloor n/2 \rfloor..n - 1]$ to $C[0..\lceil n/2 \rceil - 1]$

Mergesort($B[0..\lfloor n/2 \rfloor - 1]$)

Mergesort($C[0..\lceil n/2 \rceil - 1]$)

Merge(B, C, A)

Merging

Implementation of Merge(B,C,A)

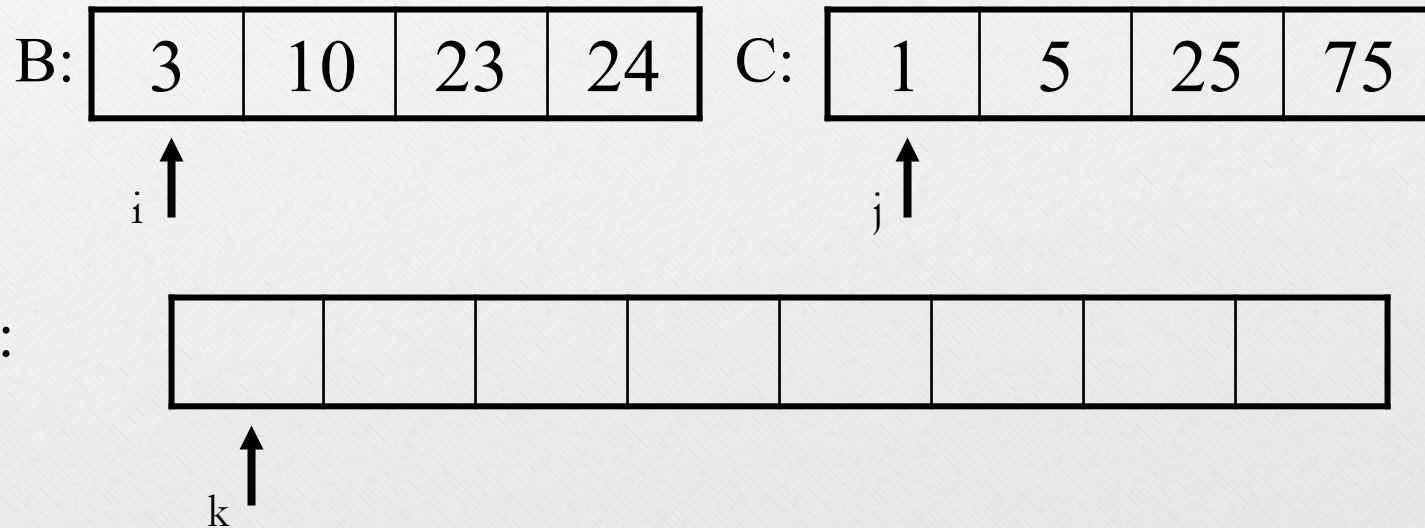
- Important two sorted arrays (B & C) and merge it into an empty array (A)

- Example:

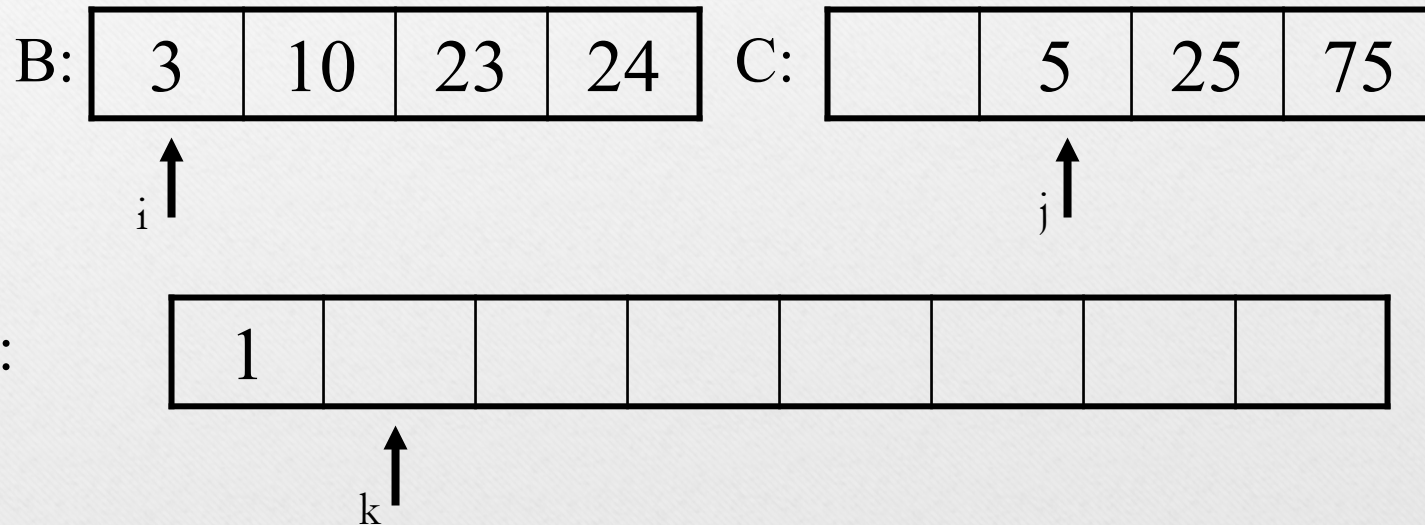
$B = \{ 3 \ 8 \ 9 \}$ $C = \{ 1 \ 5 \ 7 \}$

$\text{merge}(B, C, A) = A = \{ 1 \ 3 \ 5 \ 7 \ 8 \ 9 \}$

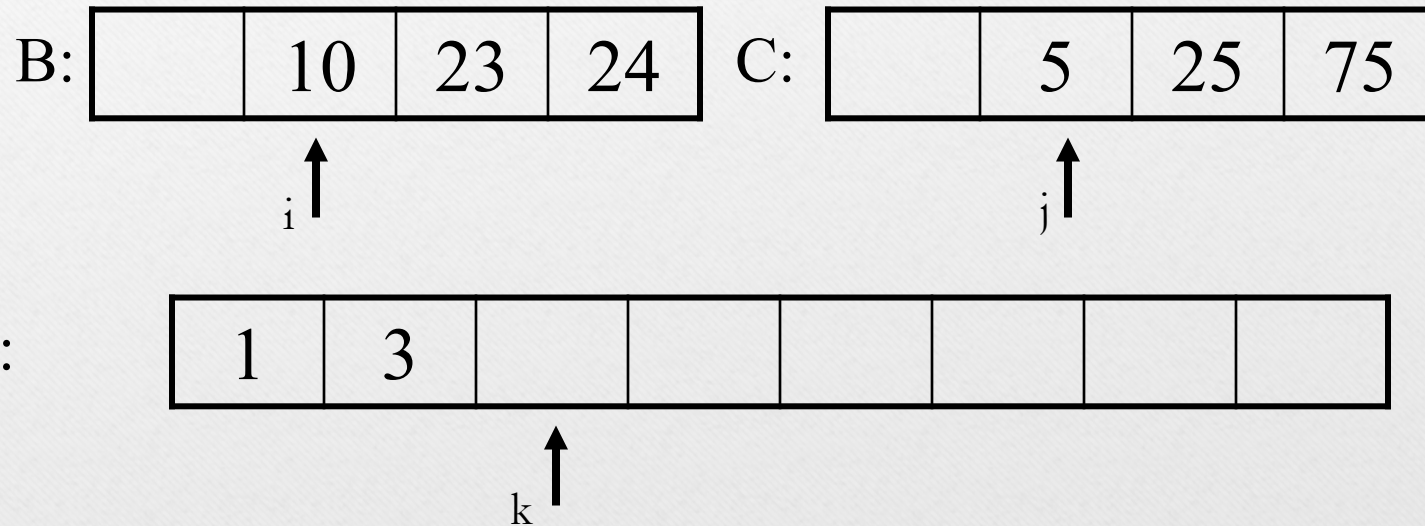
Merging (cont.)



Merging (cont.)



Merging (cont.)



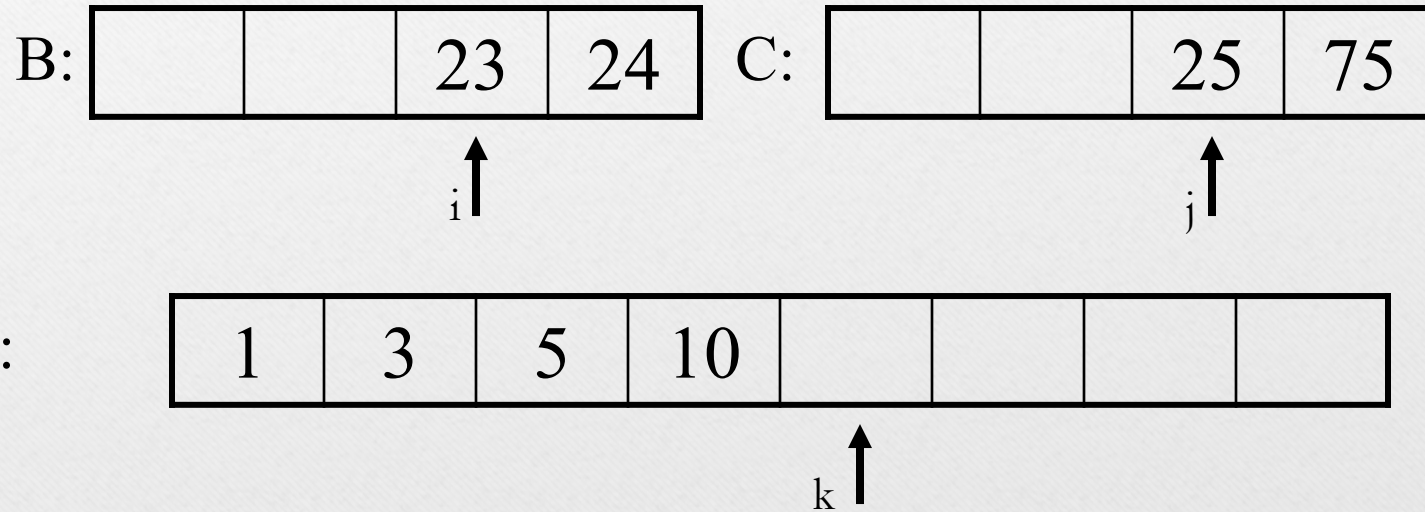
Merging (cont.)



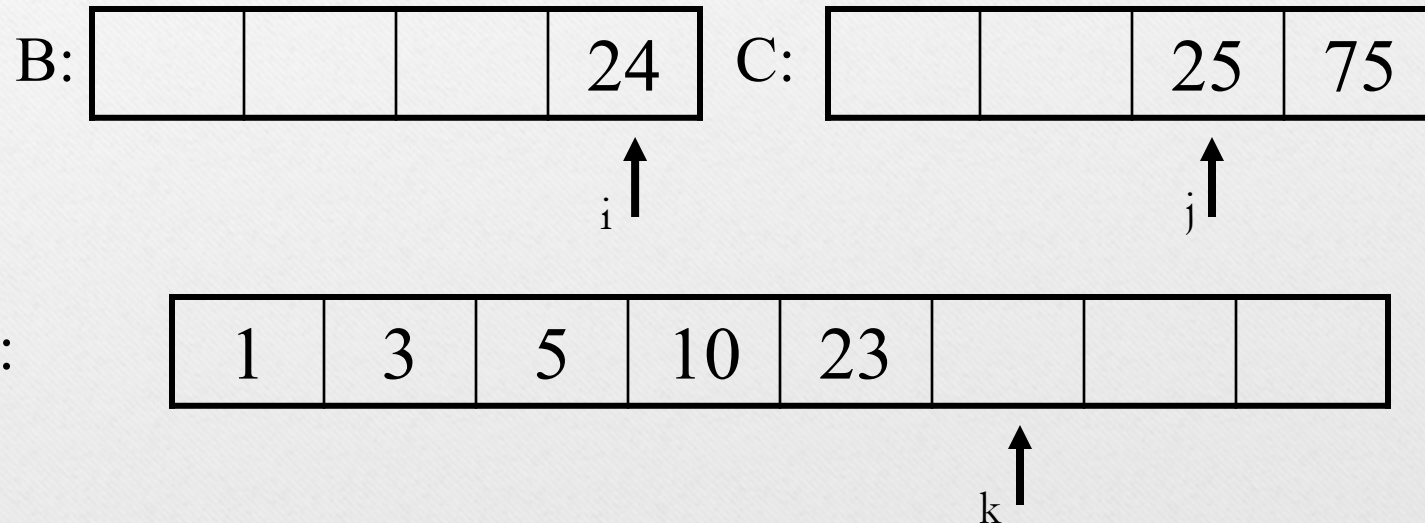
A:



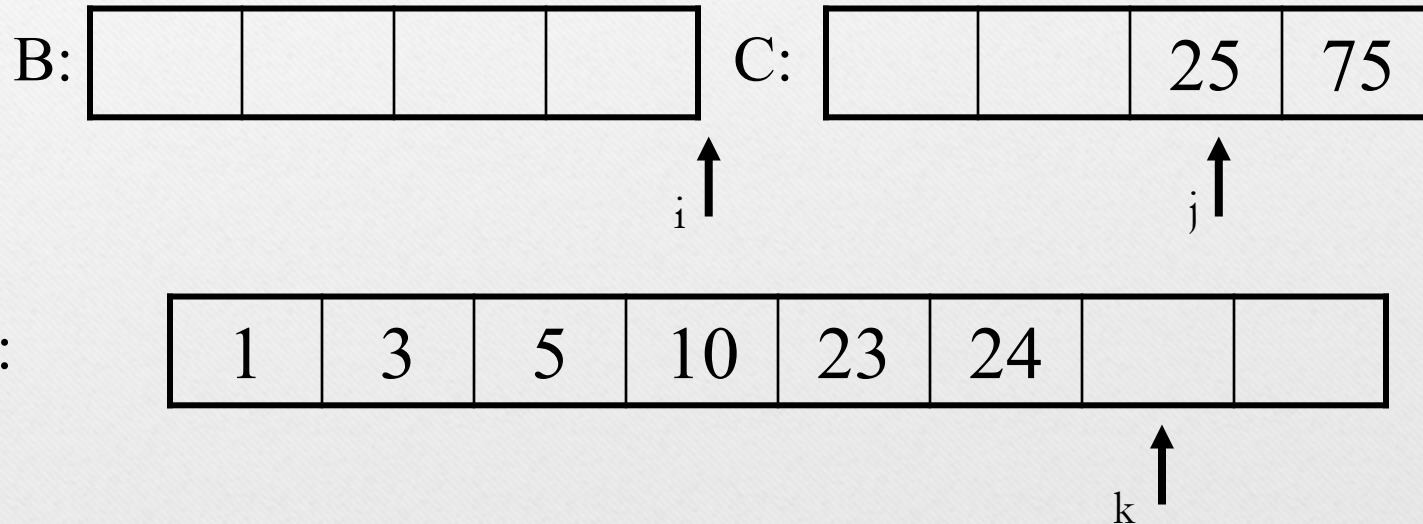
Merging (cont.)



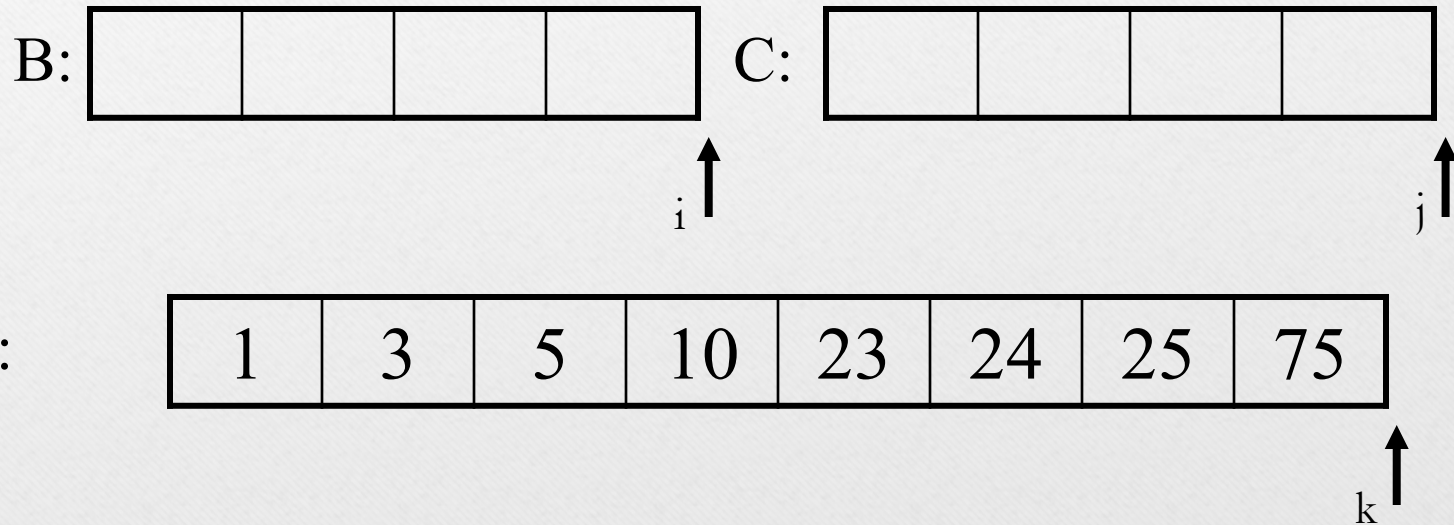
Merging (cont.)



Merging (cont.)



Merging (cont.)



Merging (cont.)

- Must put the next-smallest element into the merged list at each point
- Each next-smallest could come from either list

Pseudocode of Merge

ALGORITHM *Merge*($B[0..p-1]$, $C[0..q-1]$, $A[0..p+q-1]$)

//Merges two sorted arrays into one sorted array

//Input: Arrays $B[0..p-1]$ and $C[0..q-1]$ both sorted

//Output: Sorted array $A[0..p+q-1]$ of the elements of B and C

$i \leftarrow 0$; $j \leftarrow 0$; $k \leftarrow 0$

while $i < p$ or $j < q$ **do**

if $B[i] \leq C[j]$

$A[k] \leftarrow B[i]$; $i \leftarrow i + 1$

else $A[k] \leftarrow C[j]$; $j \leftarrow j + 1$

$k \leftarrow k + 1$

Can this be
optimized?
(Hint: i or j
maxes out)

Pseudocode of Merge

ALGORITHM *Merge*($B[0..p-1]$, $C[0..q-1]$, $A[0..p+q-1]$)

//Merges two sorted arrays into one sorted array

//Input: Arrays $B[0..p-1]$ and $C[0..q-1]$ both sorted

//Output: Sorted array $A[0..p+q-1]$ of the elements of B and C

$i \leftarrow 0$; $j \leftarrow 0$; $k \leftarrow 0$

while $i < p$ **and** $j < q$ **do**

if $B[i] \leq C[j]$

$A[k] \leftarrow B[i]$; $i \leftarrow i + 1$

else $A[k] \leftarrow C[j]$; $j \leftarrow j + 1$

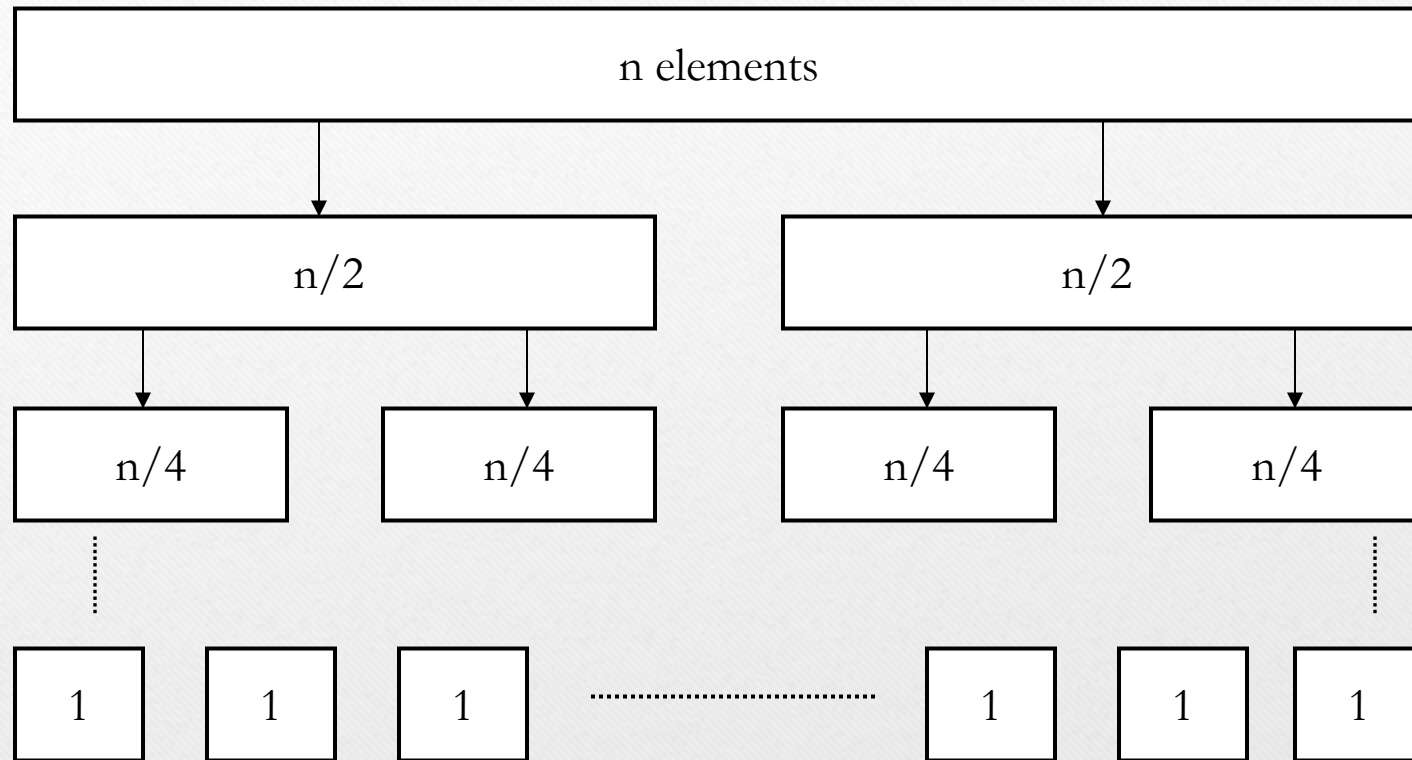
$k \leftarrow k + 1$

if $i = p$

 copy $C[j..q-1]$ to $A[k..p+q-1]$

else copy $B[i..p-1]$ to $A[k..p+q-1]$

Is this
needed?



Merge Sort Example

99	6	86	15	58	35	86	4	0
----	---	----	----	----	----	----	---	---

99	6	86	15
----	---	----	----

58	35	86	4	0
----	----	----	---	---

99	6
----	---

86	15
----	----

58	35
----	----

86	4	0
----	---	---

99

6

86

15

58

35

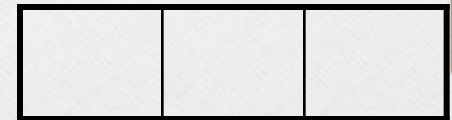
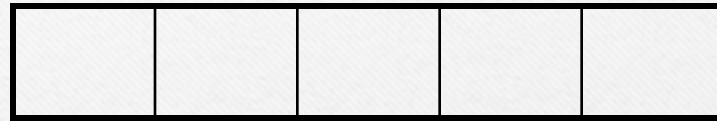
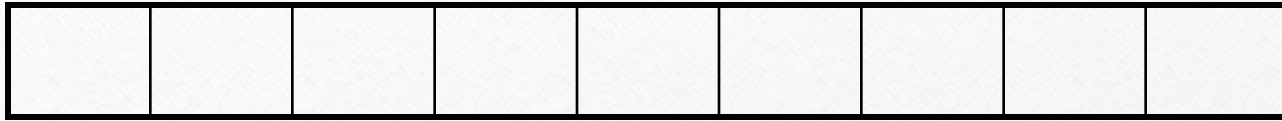
86

4	0
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4

0

Merge Sort Example



99

6

86

15

58

35

86

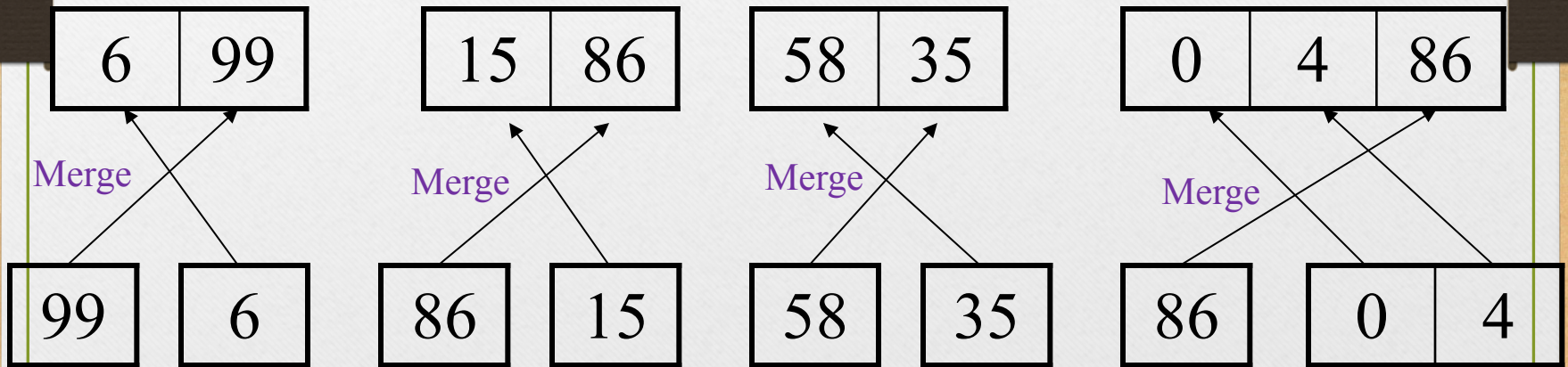
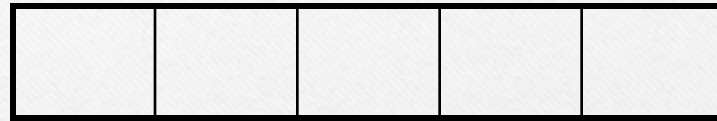
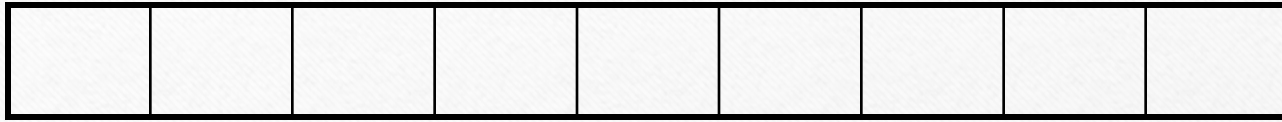
0 4

Merge

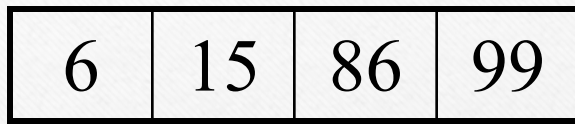
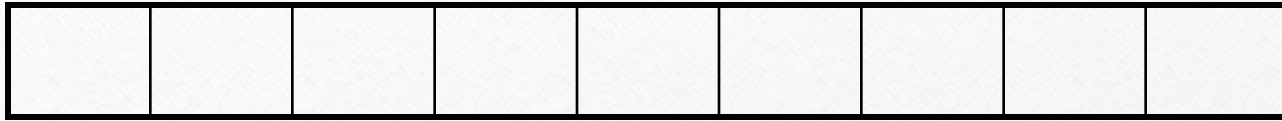
4

0

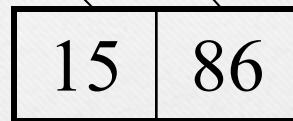
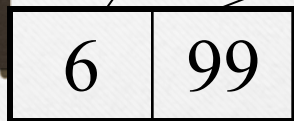
Merge Sort Example



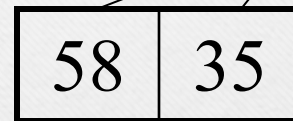
Merge Sort Example



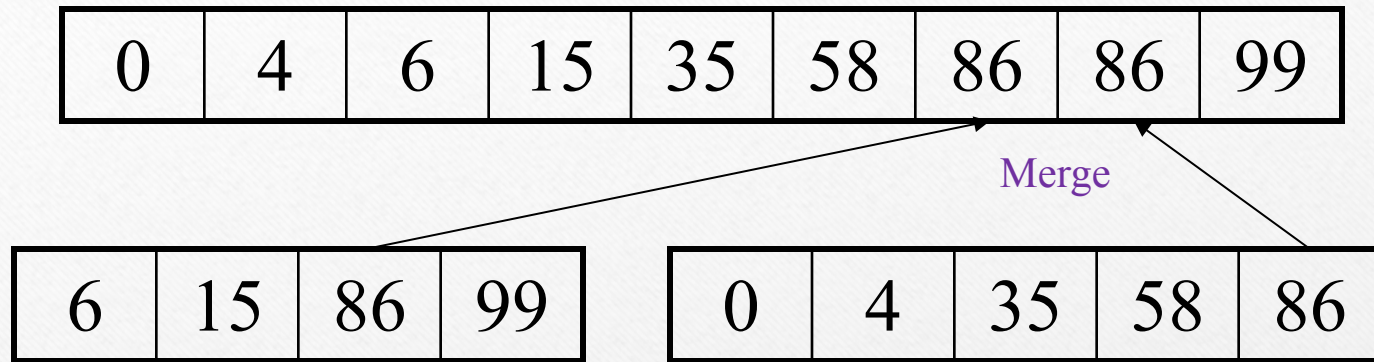
Merge



Merge



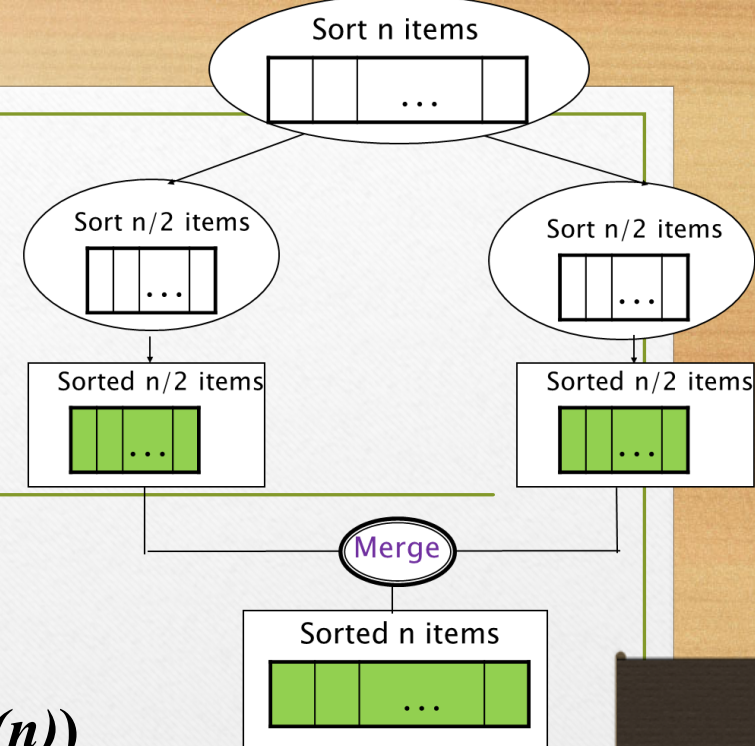
Merge Sort Example



Merge Sort Example

0	4	6	15	35	58	86	86	99
---	---	---	----	----	----	----	----	----

Masters Theorem



$$T(n) = a T(n/b) + F(n)$$

- 1) If $n^{\log_b a} < F(n)$, $T(n) \in O(F(n))$
- 2) If $n^{\log_b a} > F(n)$, $T(n) \in O(n^{\log_b a})$
- 3) If $n^{\log_b a} = F(n)$, $T(n) \in O(n^{\log_b a} \log n)$

Merge Sort: $T(n) = 2T(n/2) + n \Rightarrow T(n) \in ?$

$$\begin{aligned} a &= 2 \\ b &= 2 \end{aligned}$$



$$n^{\log_b a}$$



$$n^{\log_2 2}$$



$$n$$

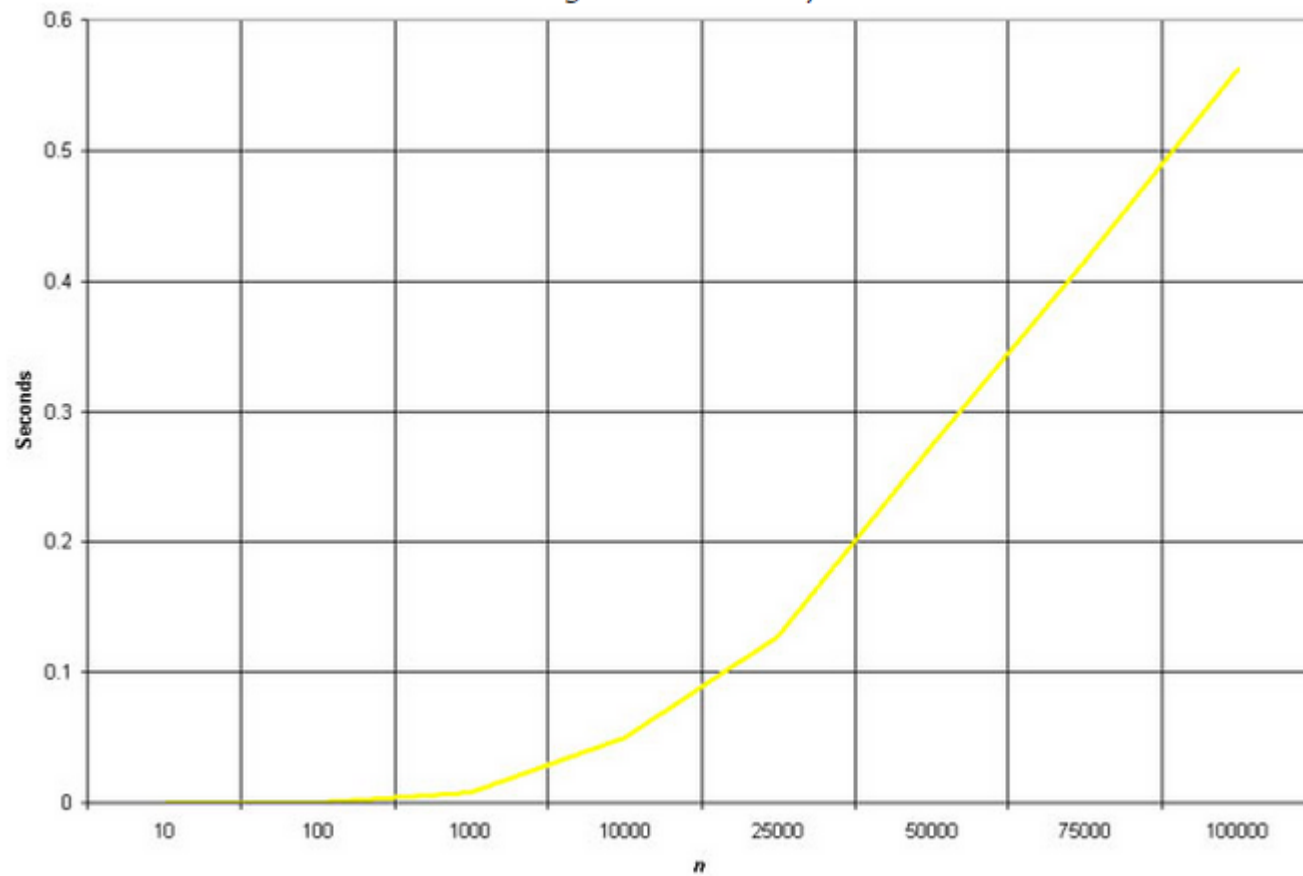
$$F(n) = n$$



$$T(n) \in O(n^{\log_b a} \log n)$$

$$T(n) \in O(n \log n)$$

Merge Sort Efficiency



This week:

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Try it/ homework

1. Chapter 5.1, page 174, questions 1, 6
2. Chapter 5.3, page 185, question 2