



DESIGN AND ANALYSIS OF ALGORITHMS (DAA)

MASTER THEOREM

MERGE SORT

Course Instructor: Dr. Shreya Ghosh

MASTER THEOREM

- When analyzing algorithms, recall that we only care about the **asymptotic behavior**.
- Recursive algorithms are no different. Rather than solve exactly the recurrence relation associated with the cost of an algorithm, it is enough to give an asymptotic characterization.
- The main tool for doing this is the master theorem

Theorem (Master Theorem)

Let $T(n)$ be a monotonically increasing function that satisfies

$$\begin{aligned}T(n) &= aT\left(\frac{n}{b}\right) + f(n) \\T(1) &= c\end{aligned}$$

where $a \geq 1, b \geq 2, c > 0$. If $f(n) \in \Theta(n^d)$ where $d \geq 0$, then

$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

MASTER THEOREM

MASTER THEOREM - PITFALLS

You *cannot* use the Master Theorem if

- ▶ $T(n)$ is not monotone, ex: $T(n) = \sin n$
- ▶ $f(n)$ is not a polynomial, ex: $T(n) = 2T(\frac{n}{2}) + 2^n$
- ▶ b cannot be expressed as a constant, ex: $T(n) = T(\sqrt{n})$

Note here, that the Master Theorem does *not* solve a recurrence relation.

MASTER THEOREM - EXAMPLES

Let $T(n) = T\left(\frac{n}{2}\right) + \frac{1}{2}n^2 + n$. What are the parameters?

$$a = 1$$

$$b = 2$$

$$d = 2$$

Therefore which condition?

Since $1 < 2^2$, case 1 applies.

Thus we conclude that

$$T(n) \in \Theta(n^d) = \Theta(n^2)$$

MASTER THEOREM - EXAMPLES

Let $T(n) = 2T\left(\frac{n}{4}\right) + \sqrt{n} + 42$. What are the parameters?

$$a = 2$$

$$b = 4$$

$$d = \frac{1}{2}$$

Therefore which condition?

Since $2 = 4^{\frac{1}{2}}$, case 2 applies.

Thus we conclude that

$$T(n) \in \Theta(n^d \log n) = \Theta(\sqrt{n} \log n)$$

MASTER THEOREM - EXAMPLES

Let $T(n) = 3T\left(\frac{n}{2}\right) + \frac{3}{4}n + 1$. What are the parameters?

$$a = 3$$

$$b = 2$$

$$d = 1$$

Therefore which condition?

Since $3 > 2^1$, case 3 applies. Thus we conclude that

$$T(n) \in \Theta(n^{\log_b a}) = \Theta(n^{\log_2 3})$$

FOURTH CONDITION

Recall that we cannot use the Master Theorem if $f(n)$ (the non-recursive cost) is not polynomial.

There is a limited 4-th condition of the Master Theorem that allows us to consider polylogarithmic functions.

Corollary

If $f(n) \in \Theta(n^{\log_b a} \log^k n)$ for some $k \geq 0$ then

$$T(n) \in \Theta(n^{\log_b a} \log^{k+1} n)$$

“FOURTH” CONDITION EXAMPLE

Say that we have the following recurrence relation:

$$T(n) = 2T\left(\frac{n}{2}\right) + n \log n$$

Clearly, $a = 2, b = 2$ but $f(n)$ is not a polynomial. However,

$$f(n) \in \Theta(n \log n)$$

for $k = 1$, therefore, by the 4-th case of the Master Theorem we can say that

$$T(n) \in \Theta(n \log^2 n)$$

ASIDE UNDERSTANDING THE MASTER THEOREM

Theorem (Master Theorem)

Let $T(n)$ be a monotonically increasing function that satisfies

$$\begin{aligned} T(n) &= aT\left(\frac{n}{b}\right) + f(n) \\ T(1) &= c \end{aligned}$$

where $a \geq 1, b \geq 2, c > 0$. If $f(n) \in \Theta(n^d)$ where $d \geq 0$, then

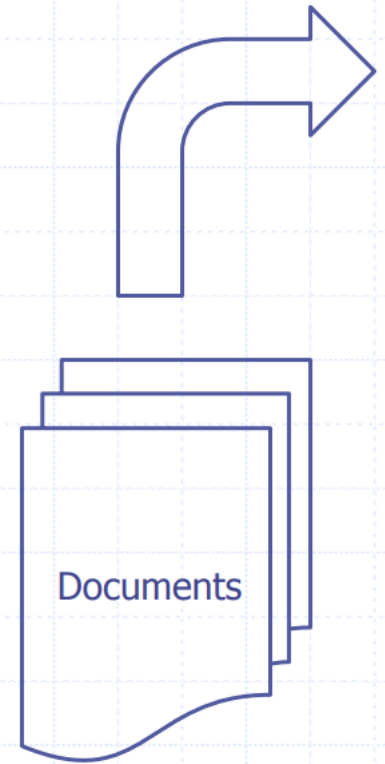
$$T(n) = \begin{cases} \Theta(n^d) & \text{if } a < b^d \\ \Theta(n^d \log n) & \text{if } a = b^d \\ \Theta(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

- a measures how many recursive calls are triggered by each method instance
- b measures the rate of change for input
- d measures the dominating term of the non recursive work within the recursive method
- c measures the work done in the base case

- The $\log_b a < d$ case
 - Recursive case does a lot of non recursive work in comparison to how quickly it divides the input size
 - Most work happens in beginning of call stack
 - Non recursive work in recursive case dominates growth, n^d term
- The $\log_b a = d$ case
 - Recursive case evenly splits work between non recursive work and passing along inputs to subsequent recursive calls
 - Work is distributed across call stack
- The $\log_b a > d$ case
 - Recursive case breaks inputs apart quickly and doesn't do much non recursive work
 - Most work happens near bottom of call stack

Application: Internet Search Engines

- Sorting has a lot of applications, including uses in Internet search engines.
- Sorting arises in the steps needed to build a data structure, known as the **inverted file** or **inverted index**, that allows a search engine to quickly return a list of the documents that contain a given keyword.

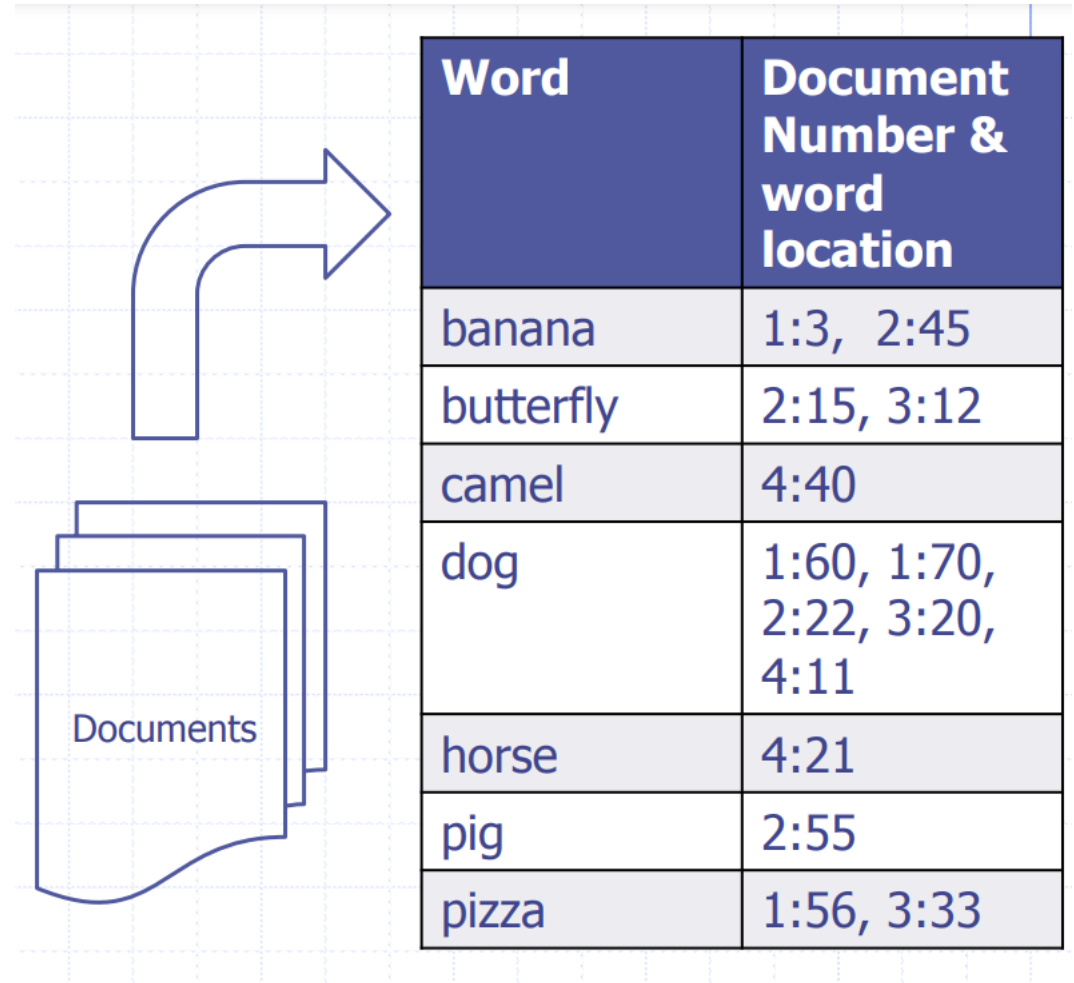


The diagram illustrates the process of building an inverted index. On the left, a stack of documents is shown with the label 'Documents'. A large, curved arrow points from this stack towards a table on the right, representing the transformation of document data into an inverted index structure.

Word	Document Number & word location
banana	1:3, 2:45
butterfly	2:15, 3:12
camel	4:40
dog	1:60, 1:70, 2:22, 3:20, 4:11
horse	4:21
pig	2:55
pizza	1:56, 3:33

Application: How Sorting Builds an Internet Search Engine

- To build an inverted file we need to identify, for each keyword, k , the documents containing k .
- Bringing all such documents together can be done simply by sorting the set of keyword document pairs by keywords.
- This places all the (k, d) pairs with the same keyword, k , right next to one another.
- From this sorted list, it is then a simple computation to scan the list and build a lookup table of documents for each keyword that appears in this sorted list.

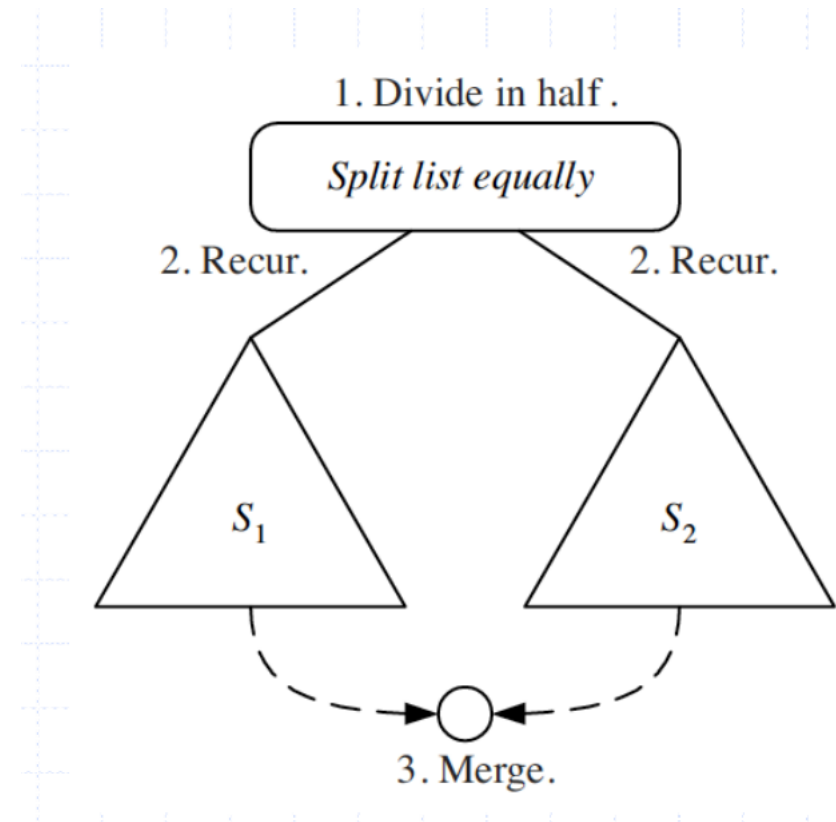


The diagram illustrates the process of building an inverted index. On the left, a stack of documents is shown with the label 'Documents'. A large, curved arrow points from this stack towards a table on the right. The table has two columns: 'Word' and 'Document Number & word location'. The table contains the following data:

Word	Document Number & word location
banana	1:3, 2:45
butterfly	2:15, 3:12
camel	4:40
dog	1:60, 1:70, 2:22, 3:20, 4:11
horse	4:21
pig	2:55
pizza	1:56, 3:33

Divide-and-Conquer

- Divide-and conquer is a general algorithm design paradigm:
 - Divide: divide the input data S in two disjoint subsets S_1 and S_2
 - Recur: solve the subproblems associated with S_1 and S_2
 - Combine the solutions for S_1 and S_2 into a solution for S
- The base case for the recursion are subproblems of size 0 or 1



The MERGE-SORT Algorithm

Merge-sort on an input sequence S with n elements consists of three steps:

- **Divide**: partition S into two sequences S_1 and S_2 of about $n/2$ elements each
- **Recur**: recursively sort S_1 and S_2
- **Conquer**: merge S_1 and S_2 into a unique sorted sequence

Algorithm *mergeSort*(S)

Input sequence S with n elements

Output sequence S sorted according to C

if $S.size() > 1$

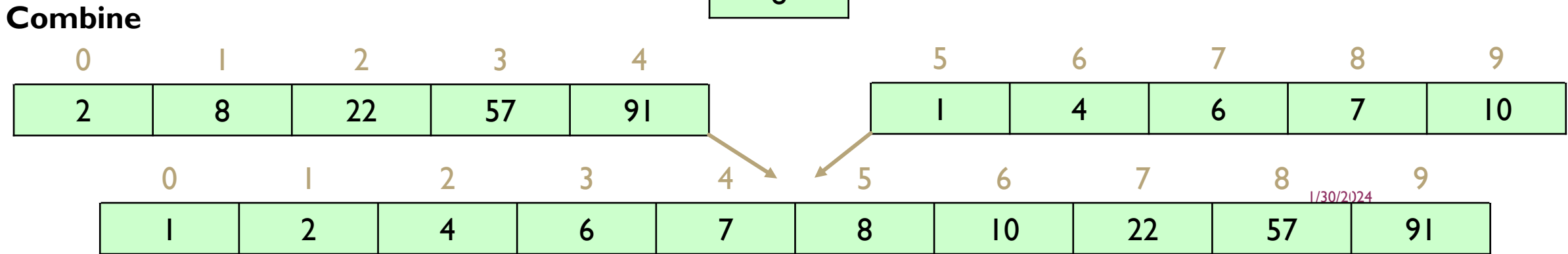
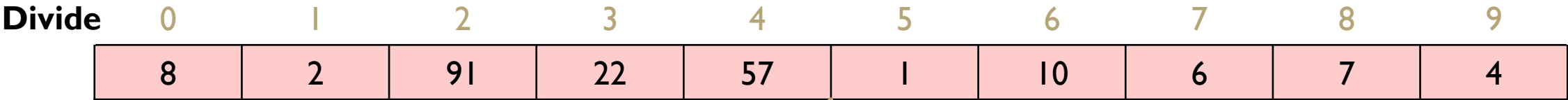
$(S_1, S_2) \leftarrow partition(S, n/2)$

mergeSort(S_1)

mergeSort(S_2)

$S \leftarrow merge(S_1, S_2)$

Merge sort



Algorithm *mergeSort(S)*

Input sequence S with n elements

Output sequence S sorted according to C

if $S.size() > 1$

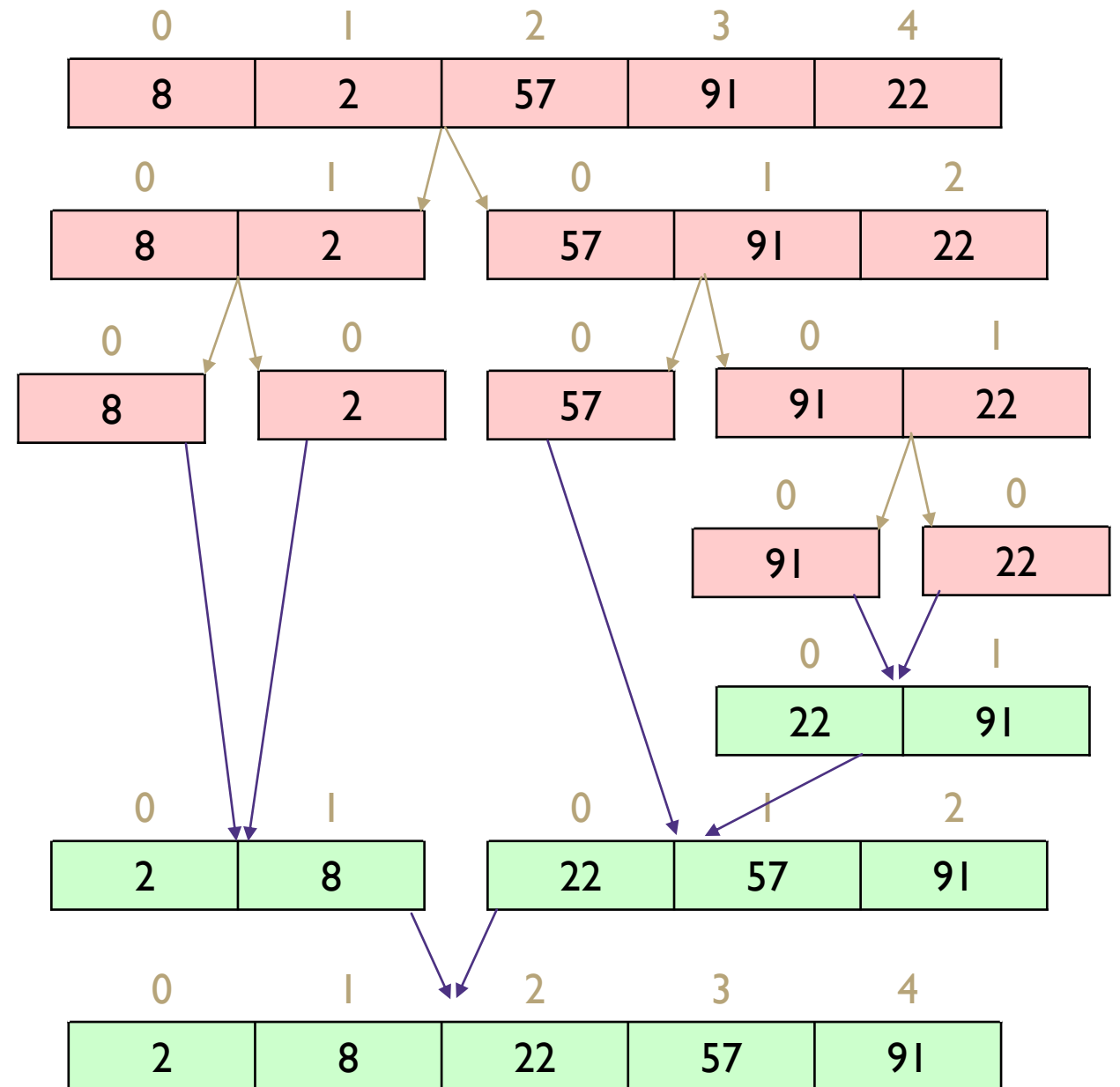
$(S_1, S_2) \leftarrow \text{partition}(S, n/2)$

$\text{mergeSort}(S_1)$

$\text{mergeSort}(S_2)$

$S \leftarrow \text{merge}(S_1, S_2)$

$$T(n) = \begin{cases} 1 & \text{if } n \leq 2 \\ 2T\left(\frac{n}{2}\right) + n & \text{otherwise} \end{cases}$$



The MERGE-SORT Algorithm

Merge-sort on an input sequence S with n elements consists of three steps:

- **Divide**: partition S into two sequences S_1 and S_2 of about $n/2$ elements each
- **Recur**: recursively sort S_1 and S_2
- **Conquer**: merge S_1 and S_2 into a unique sorted sequence

Algorithm *mergeSort*(S)

Input sequence S with n elements

Output sequence S sorted according to C

if $S.size() > 1$

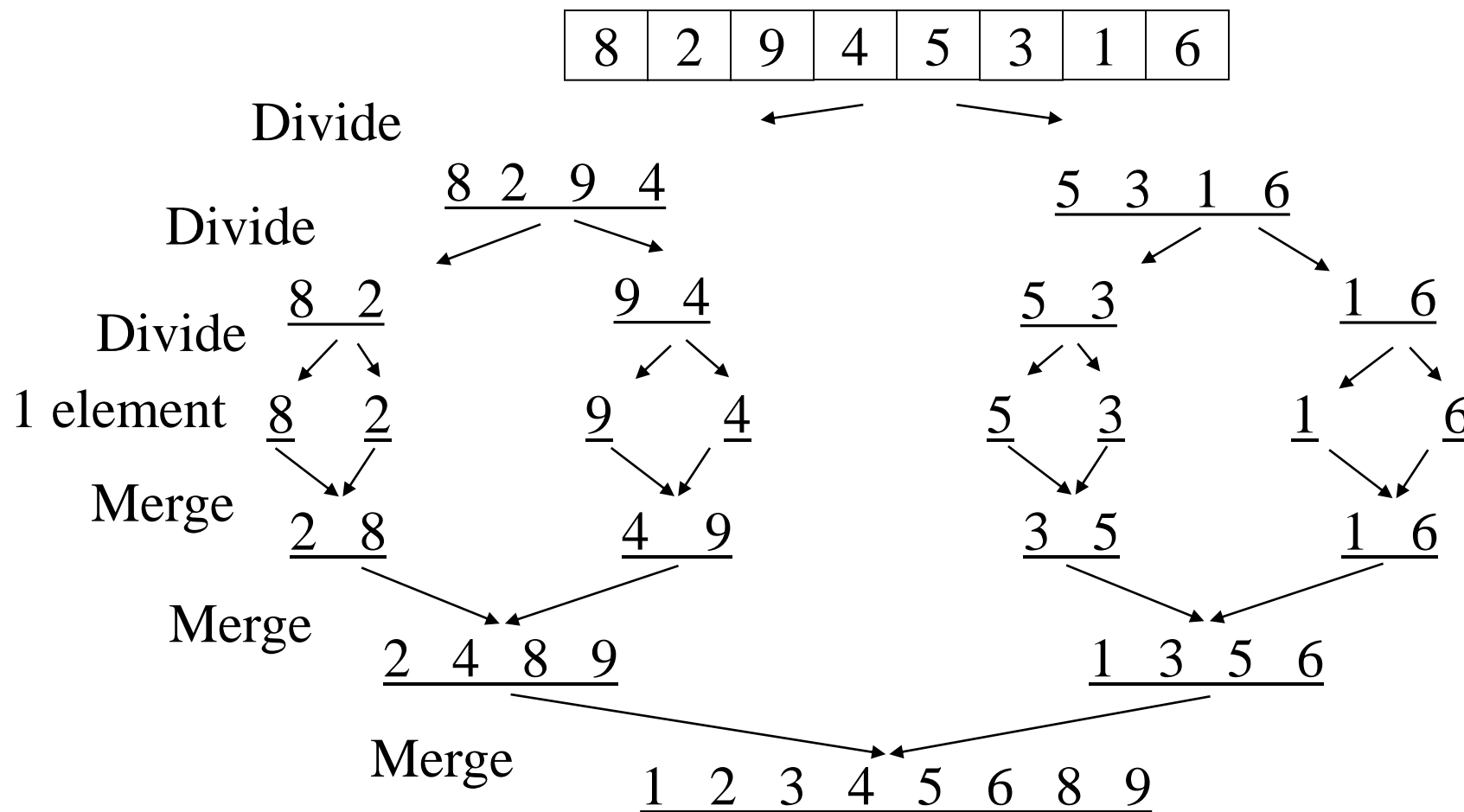
$(S_1, S_2) \leftarrow partition(S, n/2)$

mergeSort(S_1)

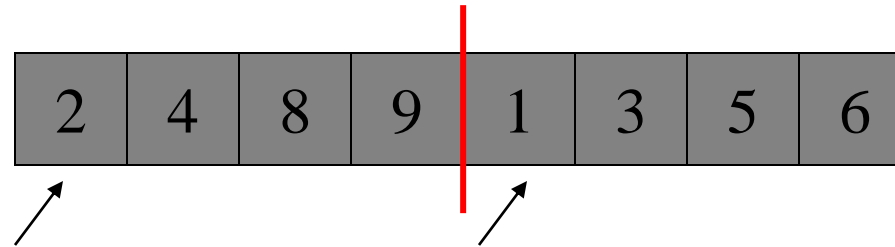
mergeSort(S_2)

$S \leftarrow merge(S_1, S_2)$

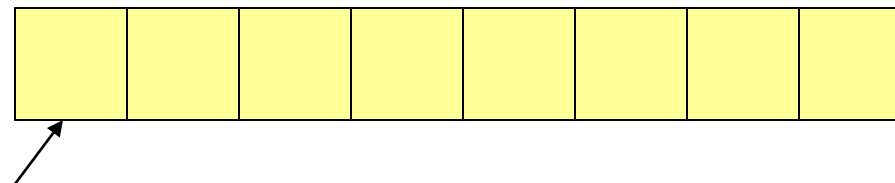
MERGE SORT EXAMPLE



AUXILIARY ARRAY

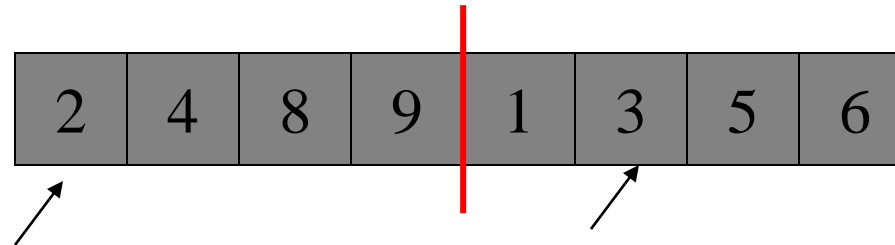


- The merging requires an auxiliary array.

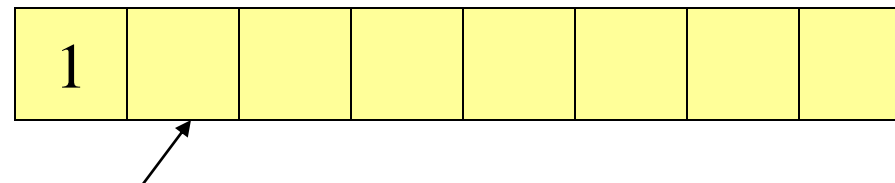


Auxiliary array

AUXILIARY ARRAY

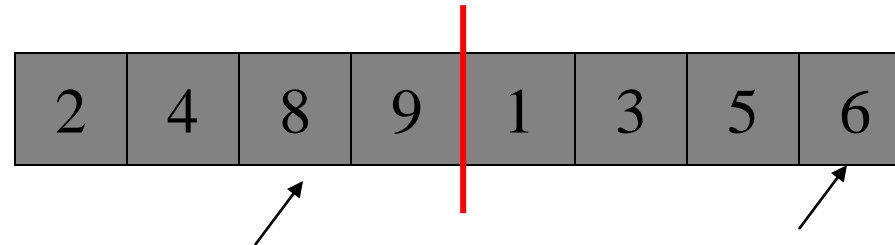


- The merging requires an auxiliary array.

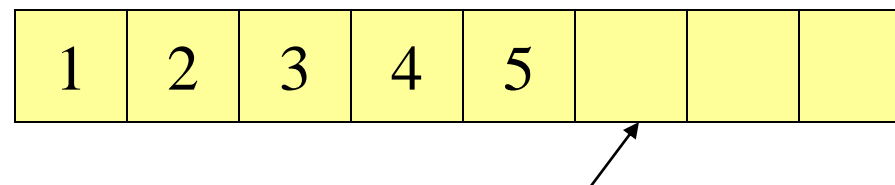


Auxiliary array

AUXILIARY ARRAY

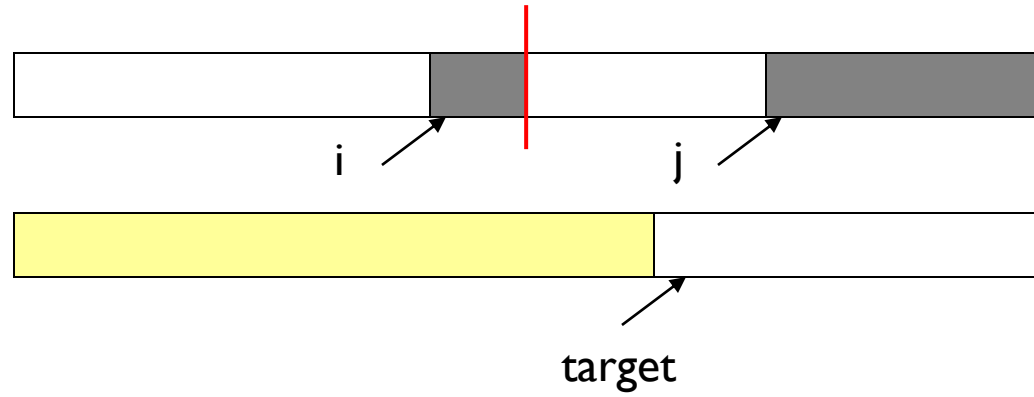


- The merging requires an auxiliary array.

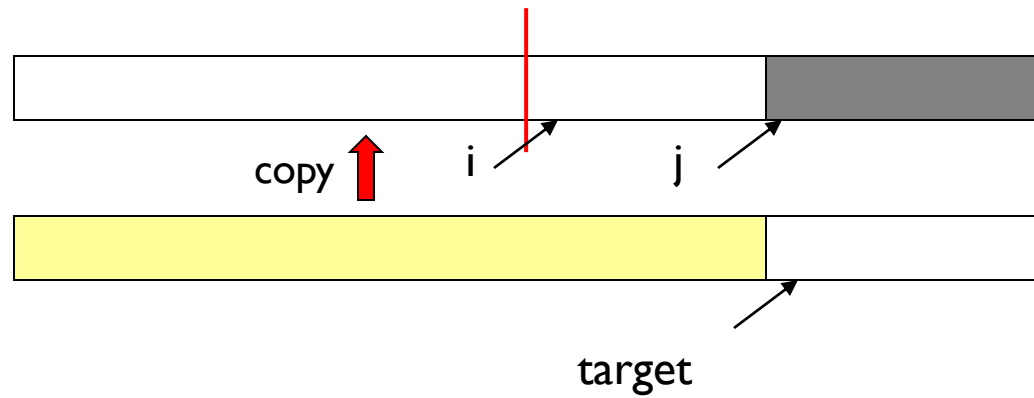


Auxiliary array

MERGING

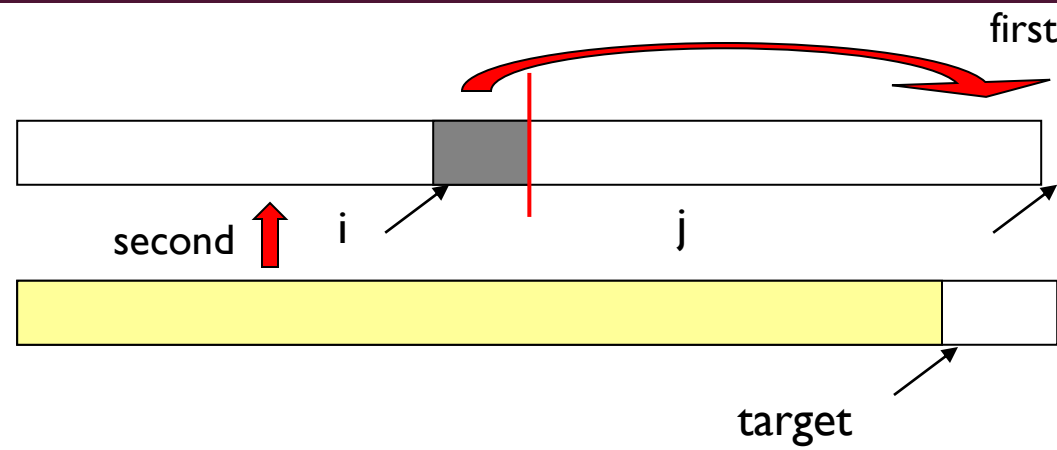


normal



Left completed
first

MERGING



Right completed
first

MERGING

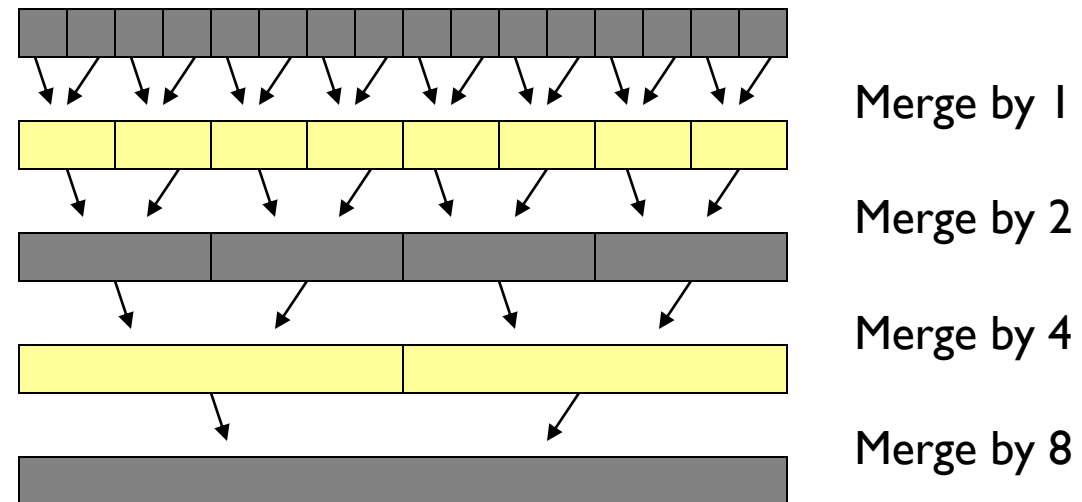
```
Merge(A[], T[] : integer array, left, right : integer) : {  
    mid, i, j, k, l, target : integer;  
    mid := (right + left)/2;  
  
    i := left; j := mid + 1; target := left;  
  
    while i ≤ mid and j ≤ right do  
        if A[i] ≤ A[j] then T[target] := A[i] ; i:= i + 1;  
        else T[target] := A[j]; j := j + 1;  
        target := target + 1;  
  
    if i > mid then //left completed//  
        for k := left to target-1 do A[k] := T[k];  
    if j > right then //right completed//  
        k := mid; l := right;  
        while k ≥ i do A[l] := A[k]; k := k-1; l := l-1;  
        for k := left to target-1 do A[k] := T[k];  
}
```


RECURSIVE MERGESORT

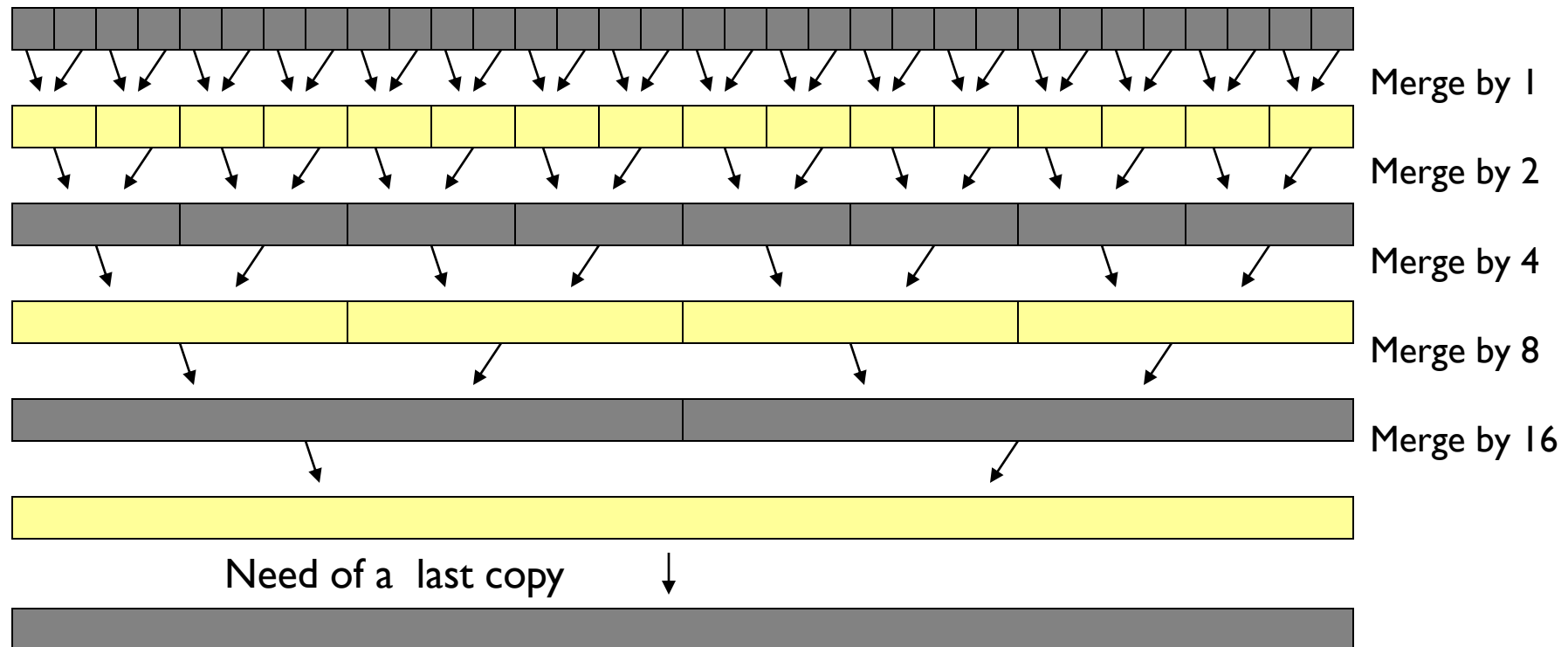
```
Mergesort(A[], T[] : integer array, left, right : integer) : {  
  if left < right then  
    mid := (left + right)/2;  
    Mergesort(A,T,left,mid);  
    Mergesort(A,T,mid+1,right);  
    Merge(A,T,left,right);  
}
```

```
MainMergesort(A[1..n]: integer array, n : integer) : {  
  T[1..n]: integer array;  
  Mergesort[A,T,1,n];  
}
```

ITERATIVE MERGESORT



ITERATIVE MERGESORT



What is the Big-Theta of worst-case Merge Sort?

$$T(n) = \begin{cases} 1 & \text{if } n \leq 2 \\ 2T\left(\frac{n}{2}\right) + n & \text{otherwise} \end{cases}$$

MASTER THEOREM

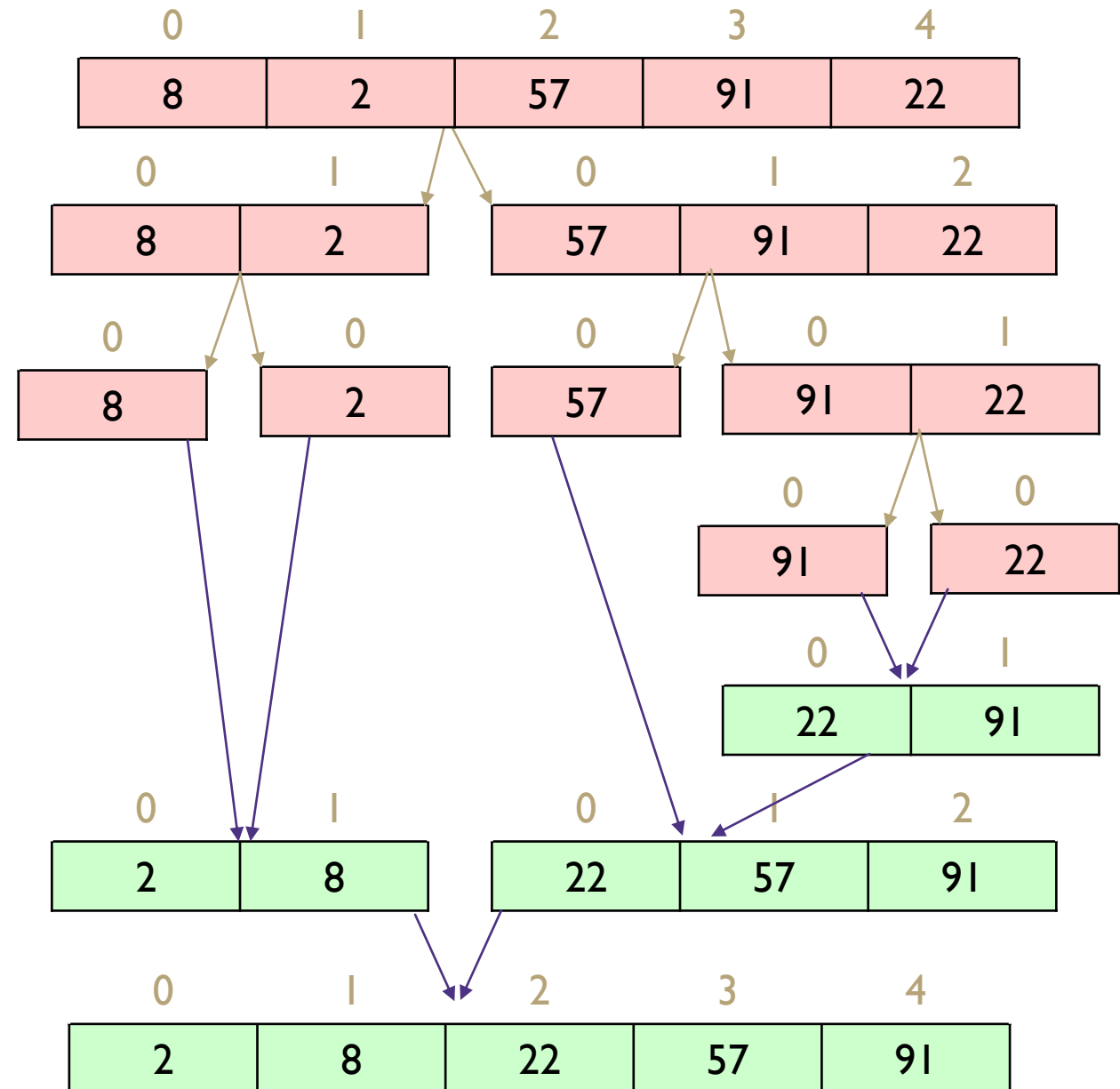
$$T(n) = \begin{cases} d & \text{if } n \text{ is at most some constant} \\ aT\left(\frac{n}{b}\right) + f(n) & \text{otherwise} \end{cases}$$

Where $f(n)$ is $\Theta(n^c)$

If $\log_b a < c$ then $T(n) \in \Theta(n^c)$

If $\log_b a = c$ then $T(n) \in \Theta(n^c \log n)$

If $\log_b a > c$ then $T(n) \in \Theta(n^{\log_b a})$



$$T(n) = \begin{cases} 1 & \text{if } n \leq 2 \\ 2T\left(\frac{n}{2}\right) + n & \text{otherwise} \end{cases}$$

MASTER THEOREM

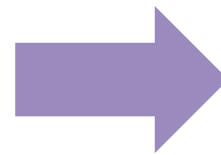
$$T(n) = \begin{cases} d & \text{if } n \text{ is at most some constant} \\ aT\left(\frac{n}{b}\right) + f(n) & \text{otherwise} \end{cases}$$

Where $f(n)$ is $\Theta(n^c)$

If $\log_b a < c$ then $T(n) \in \Theta(n^c)$

If $\log_b a = c$ then $T(n) \in \Theta(n^c \log n)$

If $\log_b a > c$ then $T(n) \in \Theta(n^{\log_b a})$

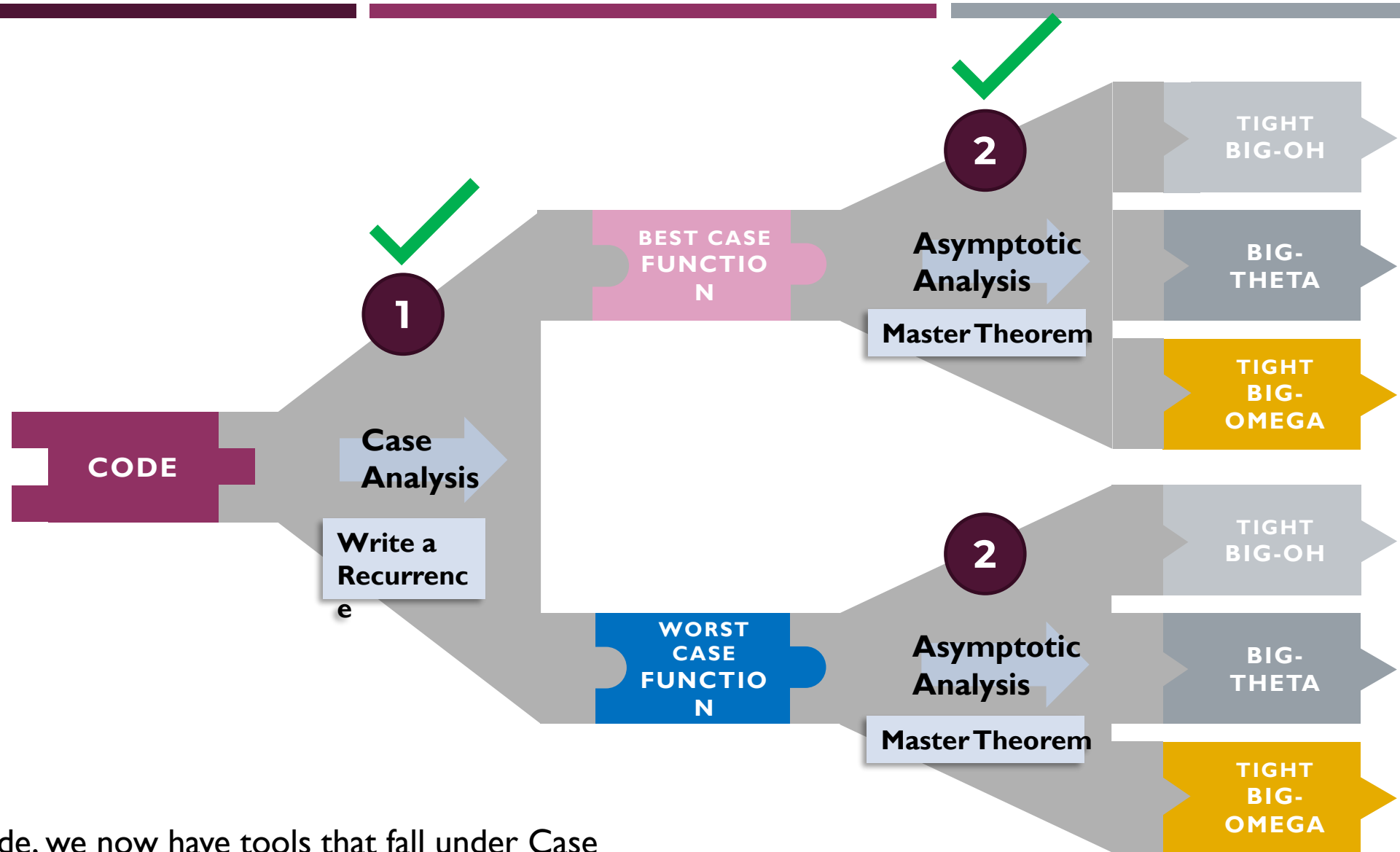


$a=2$ $b=2$ and $c=1$

$\log_2 2 = 1$

We're in case 2

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



For recursive code, we now have tools that fall under Case Analysis (Writing Recurrences) and Asymptotic Analysis (The Master Theorem).