CS302 Design and Analysis of Algorithm Week-3b

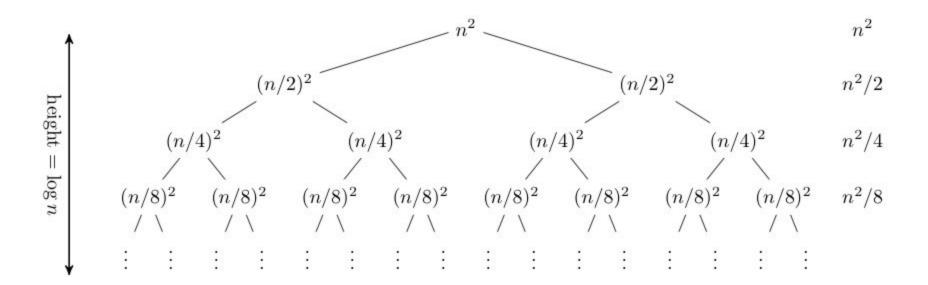
Recurrence Relations

Home Task: Do it yourself

,Solve $T(n) = 3T(n/4) + cn^2$.where c > 0 is constant

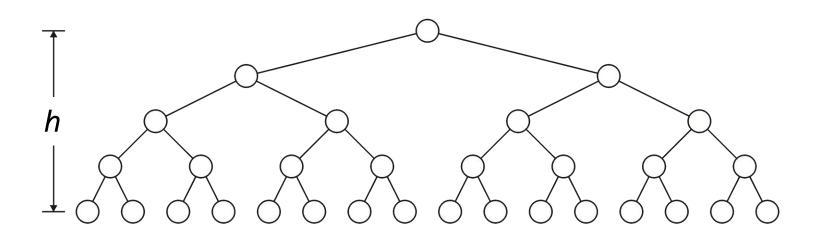
Home Task: Do it yourself

Solve
$$T(n) = 2T(n/2) + n^2$$



Binary Tree

- A perfect binary tree with height h has 2h leaf nodes
- A perfect binary tree of height h has 2^{h + 1} 1 nodes
 - Number of leaf nodes: $L = 2^h$, and with n branch factor = n^h
 - Number of internal nodes: 2^h 1
 - Total number of nodes: $2L-1 = 2^{h+1} 1$



Master Method

(Save our effort)

When the recurrence is in a special form, we can apply the Master Theorem to solve the recurrence immediately

The Master Theorem has 3 cases ...

When not to use

- You cannot use the Master Theorem if
 - •T(n) is not monotone, e.g. T(n) = sin(x)
 - f(n) is not a polynomial, e.g.,T(n)=2T(n/2)+2^n
 - •b cannot be expressed as a constant.

Why to use

- measure of algorithm efficiency
- •has a big impact on running time.
- •Big-O notation is used.
- •To deal with n items, time complexity can be O(1), O(log n), O(n), O(n log n), O(n²), O(n³), O(2ⁿ), even O(nⁿ).

Master Theorem Simplest version

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

$$T(n) = a T(n/b) + f(n)$$
$$T(1) = c$$

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

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

Master Theorem

The Master Theorem applies to recurrences of the following form:

$$T(n) = aT(n/b) + f(n)$$

where $a \ge 1$ and b > 1 are constants and f(n) is an asymptotically positive function. There are 3 cases:

- 1. If f (n) = $O(n^{\log b \text{ a-s}})$ for some constant s > 0, then T (n) = $O(n^{\log b \text{ a}})$.
- 2. If $f(n) = \Theta(n^{\log b \text{ a}} \log^k n)$ with $k \ge 0$, then $T(n) = \Theta(n^{\log b \text{ a}} \log^{k+1} n)$.
- 3. If $f(n) = \Omega(n^{\log b \text{ a+s}})$ with s > 0, and f(n) satisfies the regularity condition, then $T(n) = \Theta(f(n))$. Regularity condition: af $(n/b) \le cf(n)$ for some constant c < 1 and all sufficiently large n.

The Master Theorem

- o if T(n) = aT(n/b) + f(n) where $a \ge 1 & b > 1$
- o then

$$T(n) = \begin{cases} \Theta(n^{\log_b a}) & f(n) = O(n^{\log_b a - \varepsilon}) \\ \Theta(n^{\log_b a} \lg n) & f(n) = \Theta(n^{\log_b a}) \end{cases}$$

$$\Theta(f(n)) & f(n) = \Omega(n^{\log_b a + \varepsilon}) & af(n/b) < cf(n) \text{ for large } n \end{cases}$$

Master Theorem Updated

The Master Theorem applies to recurrences of the following form:

$$T(n) = aT(n/b) + f(n)$$

where $a \ge 1$ and b > 1 and $k \ge 0$ and p can be any real number and f (n) is an asymptotically positive function and $f(n) = \Theta(n^k \log^p n)$. There are 3 cases:

- 1. If $a > b^k$, then T (n) = $\Theta(n^{\log b a})$.
- 2. If $a = b^k$, then
 - 1. If p > -1 then $T(n) = \Theta(n^{\log b} \circ \log^{p+1} n)$.
 - 2. If p = -1 then $T(n) = \Theta(n^{\log b} \circ \log \log n)$.
 - 3. If P < -1 then $T(n) = \Theta(n^{\log b})$.
- a. If $a < b^k$, then
 - a. If $p \ge 0$, then $T(n) = \Theta(n^k \log^p n)$.
 - b. If p < 0 then $T(n) = \Theta(n^k)$.



•The form:
$$T(n) = \begin{cases} c & \text{if } n < d \\ aT(n/b) + f(n) & \text{if } n \ge d \end{cases}$$

Example

$$T(n) = 4T(n/2) + n$$

Solution:

```
a=4, b=2

f(n)=n

log_b a=2

so case 1 says T(n) is \Theta(n^2).
```



$$T(n) = \begin{cases} c & \text{if } n < d \\ aT(n/b) + f(n) & \text{if } n \ge d \end{cases}$$

•The form:

•Example:
$$T(n) = 2T(n/2) + n$$

Solution: $a=2, b=2$
 $log_b a=1$
 $f(n)= n$
 $N \ power \ matched$
so case 2 says $T(n)$ is $\Theta(n \log n)$.



•The form:
$$T(n) = \begin{cases} c & \text{if } n < d \\ aT(n/b) + f(n) & \text{if } n \ge d \end{cases}$$

•Example:

$$T(n) = T(n/3) + n \log n$$

Solution:

a=1 b=3 $\log_b a=0$ f(n)=nlogn so case 3 says T(n) is Θ (n log n).



•The form:
$$T(n) = \begin{cases} c & \text{if } n < d \\ aT(n/b) + f(n) & \text{if } n \ge d \end{cases}$$

•Example:
$$T(n) = 8T(n/2) + n^2$$

Solution:

```
a=8
b=2
f(n)=n^2
log_b a=3
so case 1 says T(n) is \Theta(n^3).
```



•The form:
$$T(n) = \begin{cases} c & \text{if } n < d \\ aT(n/b) + f(n) & \text{if } n \ge d \end{cases}$$

•Example:

$$T(n) = 9T(n/3) + n^3$$

Solution:

```
a=9
b=3
f(n)=n^3
log_b a=2
so case 3 says T(n) is \Theta(n^3).
```



•The form:
$$T(n) = \begin{cases} c & \text{if } n < d \\ aT(n/b) + f(n) & \text{if } n \ge d \end{cases}$$

•Example: T(n) = T(n/2) + 1

• Solution:

```
a=1
b=2
f(n)=1
log_b a=0
so case 2 says T(n) is \Theta(log n).
```



Merge sort

MERGE-SORT A[1 ... n]

- 1. If n = 1, done.
- 2. Recursively sort $A[1..\lceil n/2\rceil]$ and $A[\lceil n/2\rceil+1..n]$.
- 3. "Merge" the 2 sorted lists.

Key subroutine: MERGE

Analyzing merge sort

```
T(n)
              Merge-Sort A[1 ... n]
     \Theta(1) 1. If n = 1, done.
     2T(n/2) 2. Recursively sort A[1...[n/2]]
                  and A[n/2]+1... n .
     \Theta(n)
               3. "Merge" the 2 sorted
               lists
T(n)=2T(n/2)+cn
                        (recall previous lecture)
The best and worst case running time will be O(n
logn).
```



20 12

13 1

7 ¹9

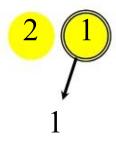
2 1



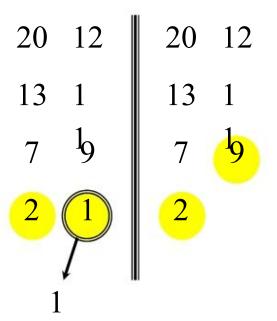
```
20 12
```

13 1

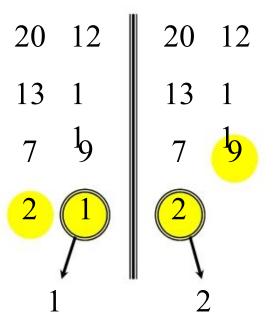
7 19



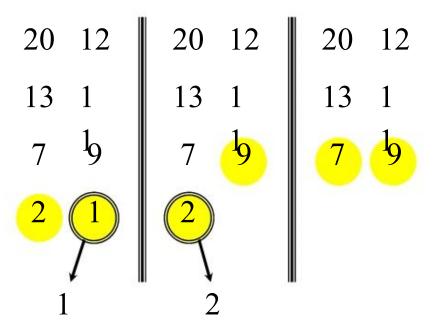




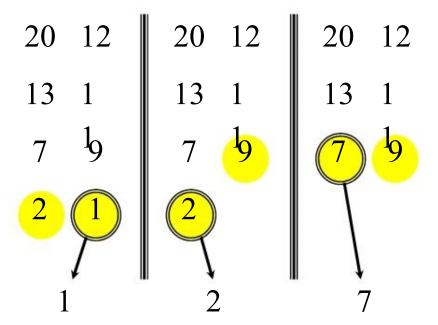




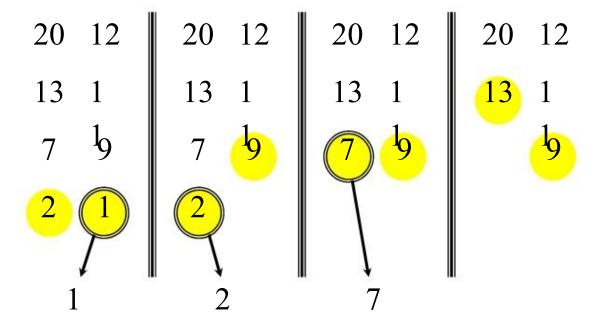




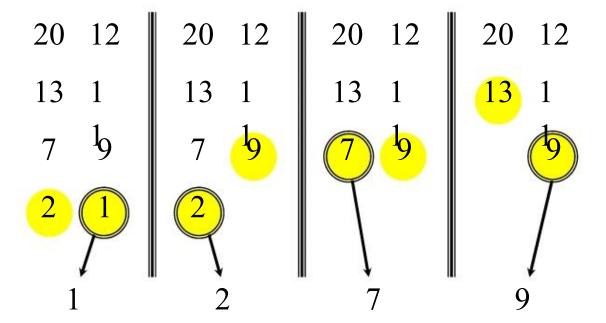




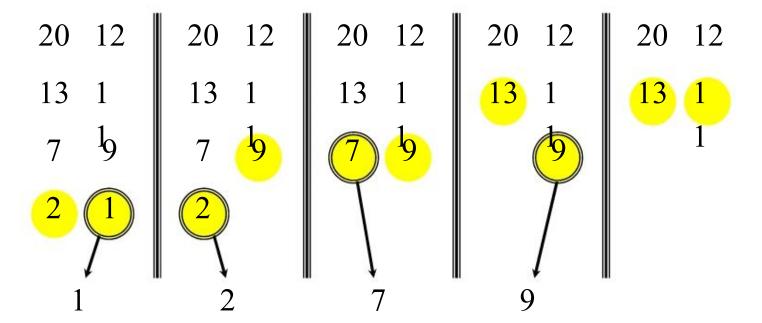




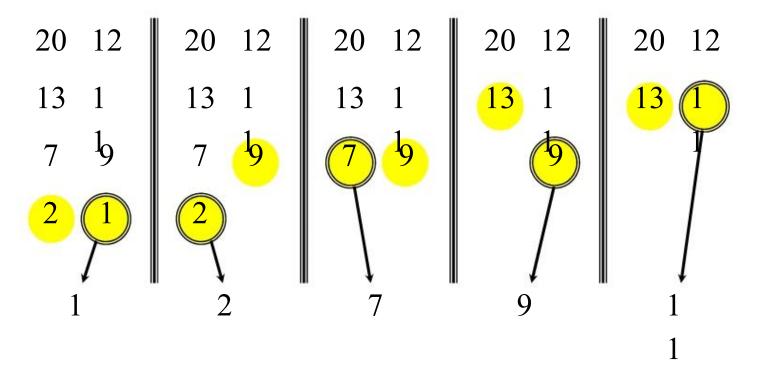




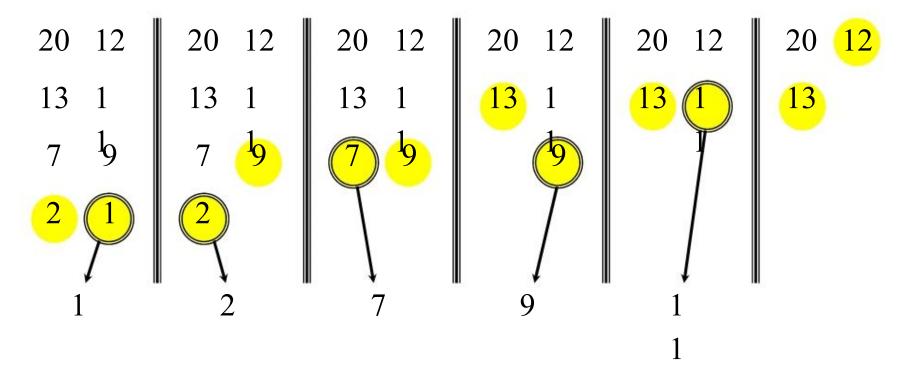




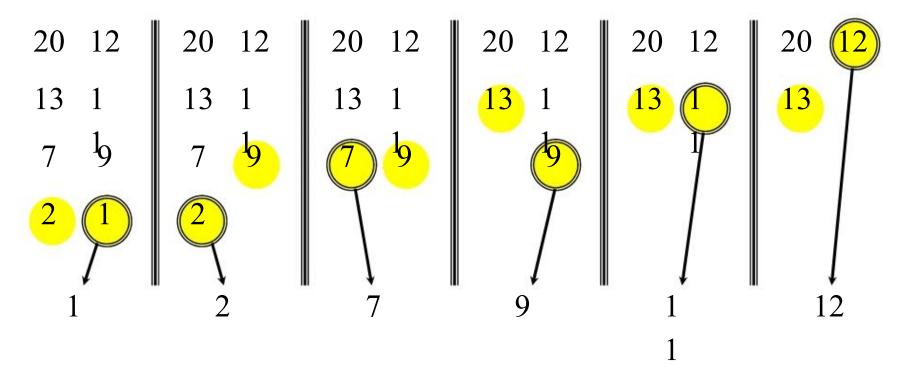




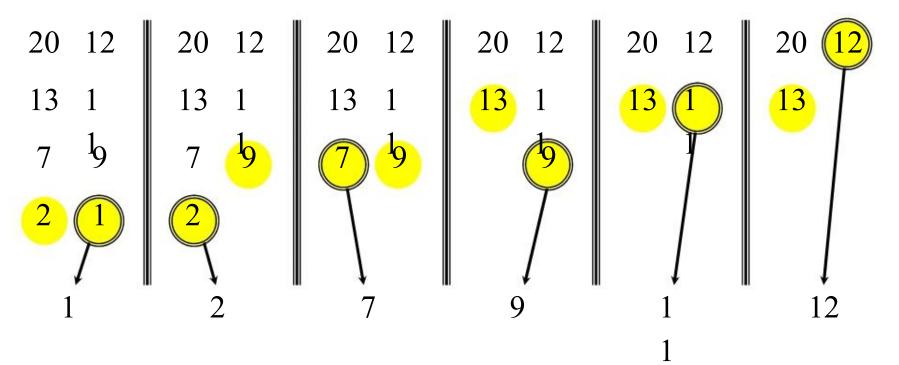












Time = $\Theta(n)$ to merge a total of n elements (linear time).

Merge Sort

- (Divide:) split A down the middle into two subsequences, each of size roughly n/2
- (Conquer:) sort each subsequence by calling merge sort recursively on each.
- (Combine:) merge the two sorted subsequences into a single sorted list

Merge Sort

```
MERGE-SORT( array A, int p, int r)

1 if (p < r)

2 then

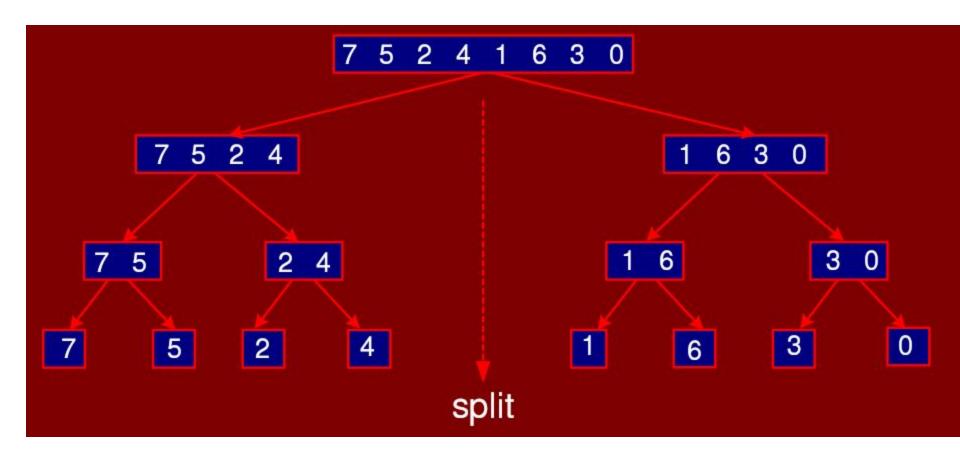
3 q \leftarrow (p+r)/2

4 MERGE-SORT(A, p, q) // sort A[p..q]

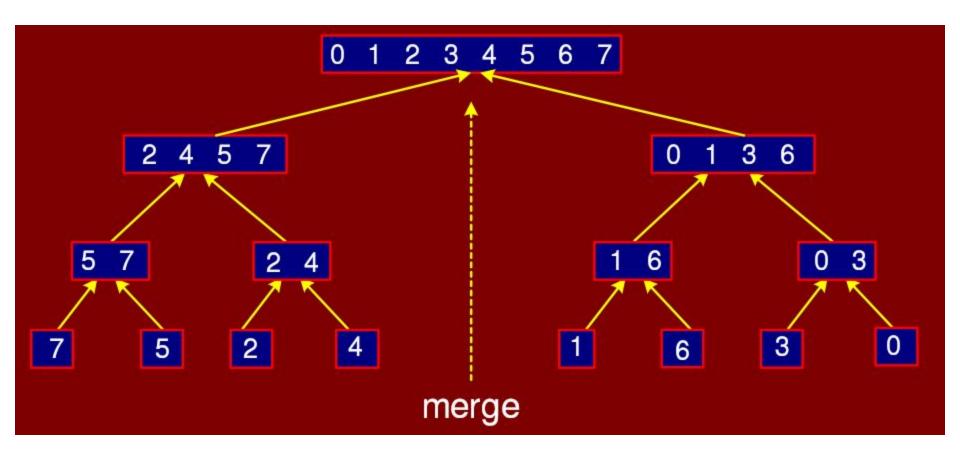
5 MERGE-SORT(A, q+1, r) // sort A[q+1..r]

6 MERGE(A, p, q, r) // merge the two pieces
```

Split part



Merge Part



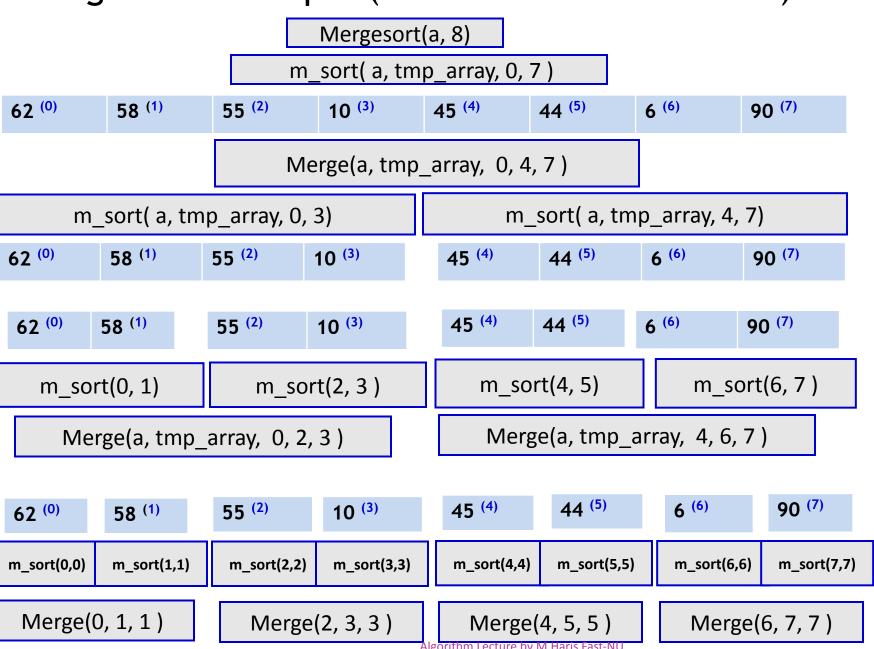
Merge Sort

- The fundamental operation in this algorithm is merging two sorted lists.
- Because the lists are sorted, this can be done in one pass through the input, if the output is put in a third list.
- The basic merging algorithm takes
 - two input arrays: a and b,
 - an output array: c
 - three counters: *aptr, bptr,* and *cptr,*
 - which are initially set to the beginning of their respective arrays.
- The smaller of a[aptr] and b[bptr] is copied to the next entry in c, and the appropriate counters are advanced.
- When either input list is exhausted, the remainder of the other list is copied to *c*.

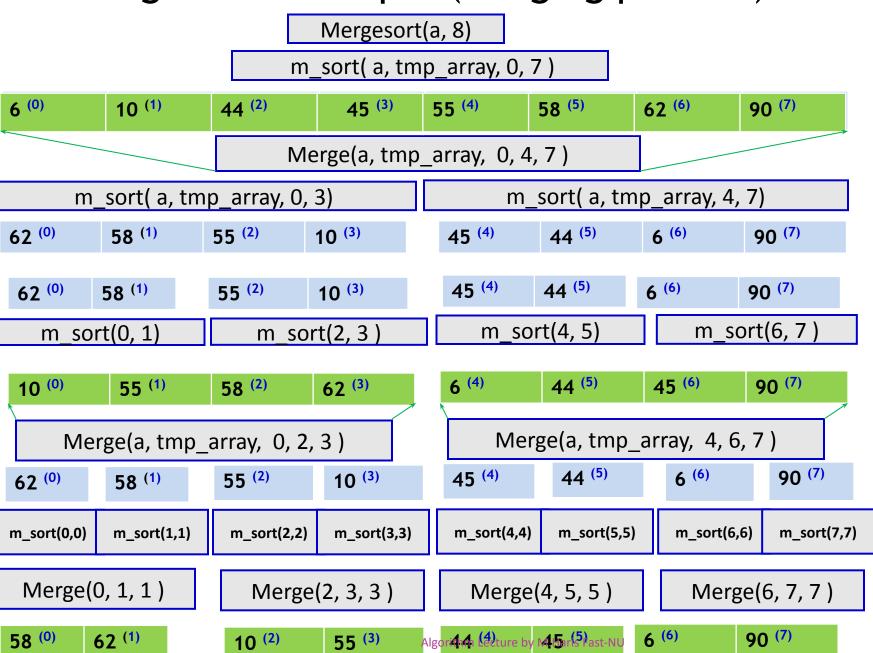
Merge Sort

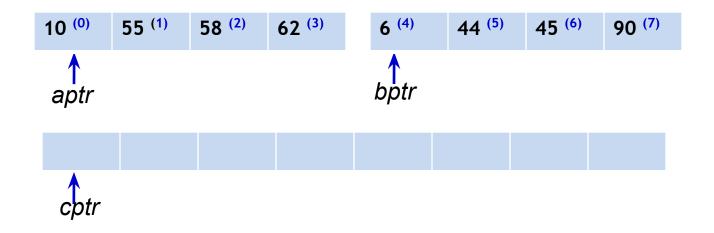
```
void m_sort( input_type a[], input_type tmp_array[ ], int left, int right )
int center;
if( left < right )</pre>
 center = (left + right) / 2;
                                   Calculate the centre index of the input list
                                            Recursively call the m_sort procedure
 m_sort( a, tmp_array, left, center );
                                            for t Recursively call the m sort procedure
 m_sort( a, tmp_array, center+1, right ),
                                                 for the right-half of the input data
 merge( a, tmp_array, left, center+1, right );
                                                      Merge the two sorted lists
```

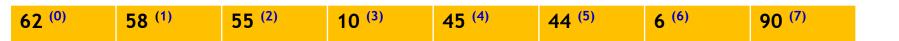
Merge Sort Example (recursive Function Calls)

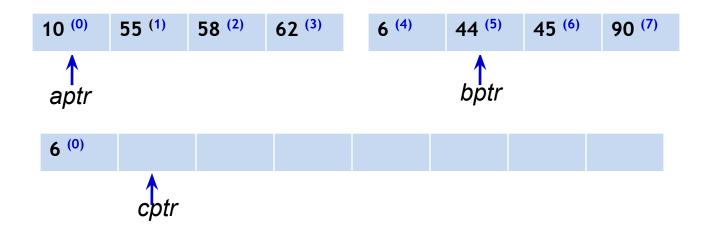


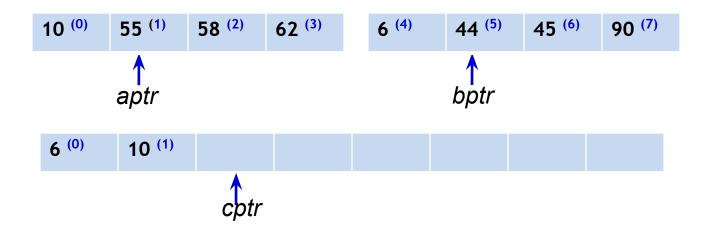
Merge Sort Example (Merging process)

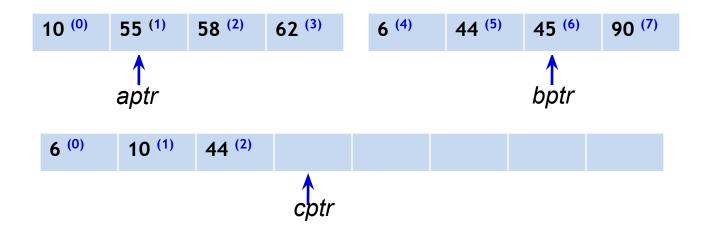


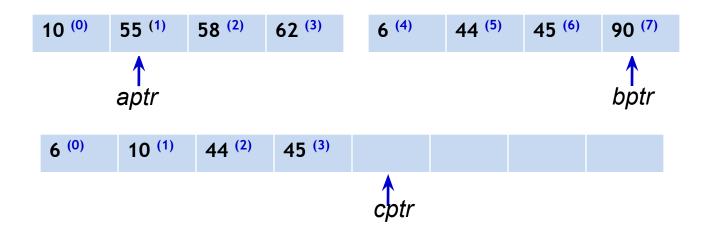


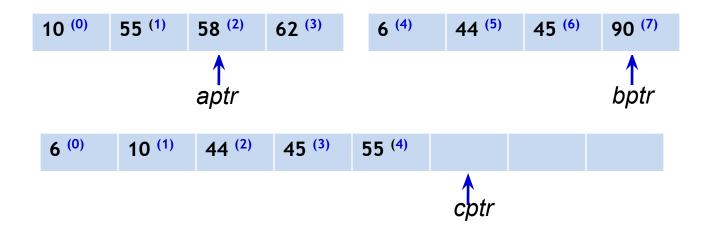


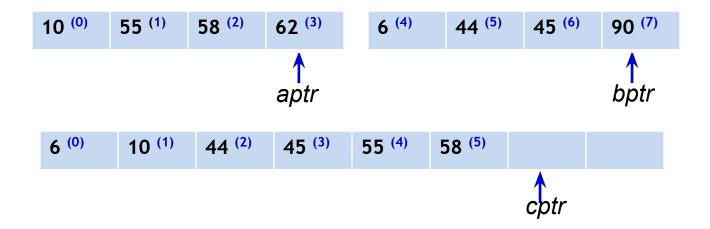


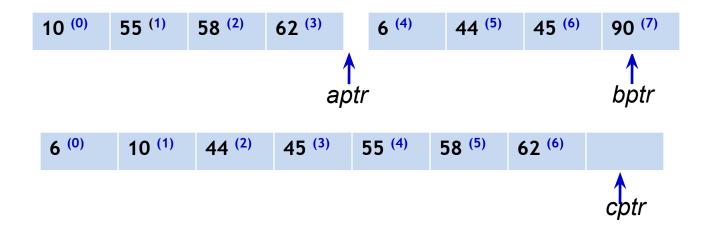


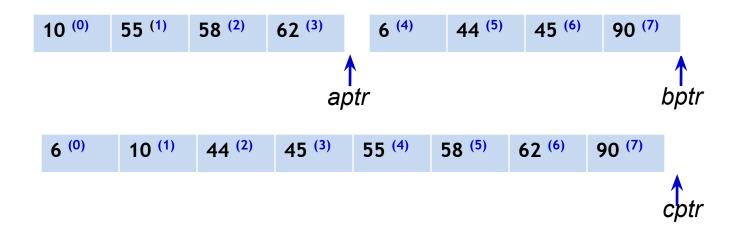












Time Complexity of merge sort using tree method, Master theorem

$$T(n) = 2 T(n/2) + O(n)$$
of sub-problems size of sub-problems work dividing & combining

Merge Sort is an efficient, stable sorting algorithm with an average, best-case, and worst-case time complexity of **O(n log n)**

Assignment ahead

Deadline: 14/02/2023