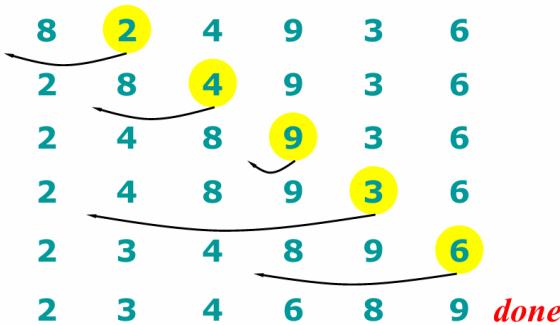


Data Structure Lecture 1: Algorithm Complexity Analysis

Part 1. Sort Examples

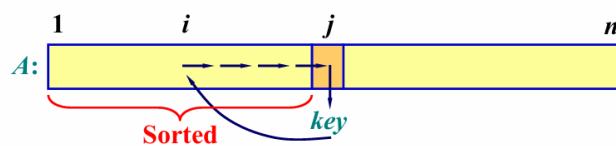
1. Insertion sort

Example of insert sort



1. 伪代码：

```
INSERTION-SORT( $A$ )
1 for  $j \leftarrow 2$  to  $\text{length}[A]$ 
2   do  $\text{key} \leftarrow A[j]$ 
3     // Insert  $A[j]$  into the sorted sequence  $A[1 \dots j - 1]$ 
4      $i \leftarrow j - 1$ 
5     while  $i > 0$  and  $A[i] > \text{key}$ 
6       do  $A[i + 1] \leftarrow A[i]$ 
7        $i \leftarrow i - 1$ 
8    $A[i + 1] \leftarrow \text{key}$ 
```



2. C++实现：

代码块

```
1 template<typename T>
2 void Insertion_sort(std::vector<T>& arr){
3     int length = arr.size();
4     for(int j = 1; j < length; j++){
5         T key = arr[j];
6         int i = j - 1;
7         while(i >= 0 && arr[i] > key){
8             arr[i + 1] = arr[i];
9             i--;
10        }
11    }
12}
```

```

10      }
11      arr[i + 1] = key;
12  }
13 }
```

3. Time Complexity Analysis

INSERTION-SORT(A)	cost	times
1 for $j \leftarrow 2$ to $\text{length}[A]$	c_1	n
2 do $\text{key} \leftarrow A[j]$	c_2	$n - 1$
3 // Insert $A[j]$ into the sorted sequence $A[1 .. j - 1]$	0	$n - 1$
4 $i \leftarrow j - 1$	c_4	$n - 1$
5 while $i > 0$ and $A[i] > \text{key}$	c_5	$\sum_{j=2}^n t_j$
6 do $A[i + 1] \leftarrow A[i]$	c_6	$\sum_{j=2}^n (t_j - 1)$
7 $i \leftarrow i - 1$	c_7	$\sum_{j=2}^n (t_j - 1)$
8 $A[i + 1] \leftarrow \text{key}$	c_8	$n - 1$

$$T(n) = c_1 n + c_2(n-1) + c_4(n-1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n-1)$$

Best case:

- The best case occurs if the array is already sorted

$$\begin{aligned} T(n) &= c_1 n + c_2(n-1) + c_4(n-1) + c_5(n-1) + c_8(n-1) \\ &= (c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8) \quad \text{an + b (linear function)} \end{aligned}$$

Worst case:

- The worst case occurs if the array is in reverse sorted order

$$\begin{aligned} T(n) &= c_1 n + c_2(n-1) + c_4(n-1) + c_5\left(\frac{n(n+1)}{2} - 1\right) + c_6\left(\frac{n(n-1)}{2}\right) + c_7\left(\frac{n(n-1)}{2}\right) + c_8(n-1) \\ &= \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2}\right)n^2 + (c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8)n - (c_2 + c_4 + c_5 + c_8) \\ &\quad \text{an}^2 + bn + c \text{ (quadratic function)} \end{aligned}$$

The a , b , c coefficients are related to particular machines. When $n \rightarrow \infty$, we don't care about them.

Insertion sort works best with small amounts of data, since the coef of insert sort is the smallest (only compare and valuation), while other sort algorithms need more operations.

2. Merge Sort

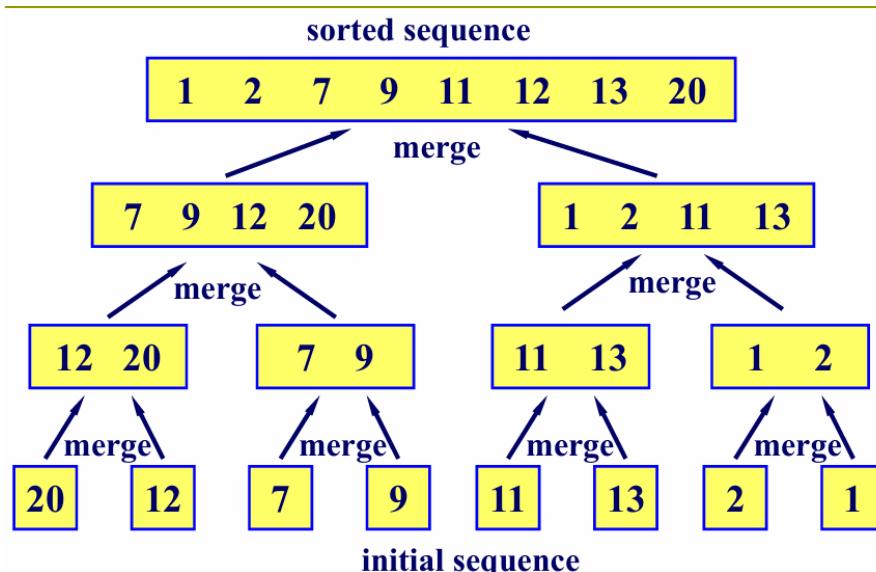
Idea : divide and conquer

1. pseudocode

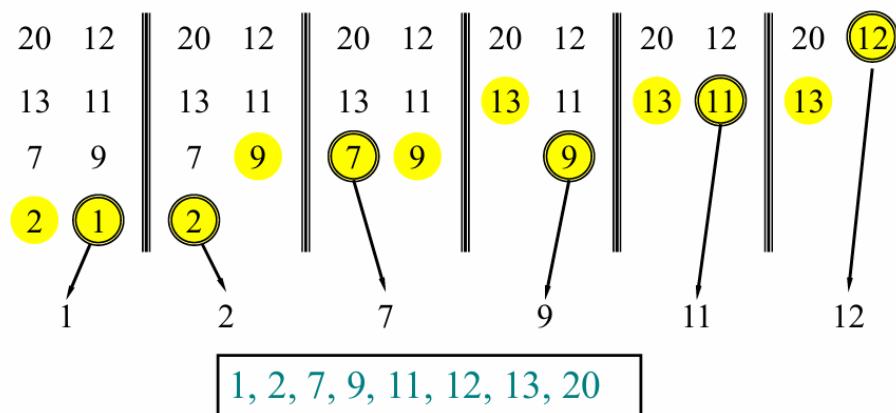
MERGE-SORT $A[1..n]$

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

Key subroutine: MERGE



Merging two sorted arrays



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

2. C++ Implementation

代码块

```
1 template<typename T>
2 void Merge_sort_1(std::vector<T>& arr){
3     std::function<void(int, int)> helper = [&](int left, int right) -> void{
4         if(left >= right - 1){
5             return;
6         }
7
8         std::vector<T> temp(arr.size()); // temporary vector to store the
element
```

```

9
10         int middle = (left + right) / 2;
11         // divide
12         helper(left, middle);
13         helper(middle, right);
14
15         for(int i = left; i < right; i++){
16             temp[i] = arr[i];
17         }
18
19         // conquer
20         int p1 = 0, p2 = 0, p = left;
21         while(p1 < middle - left || p2 < right - middle){
22             // left used up
23             if(p1 == middle - left){
24                 arr[p] = temp[p2 + middle];
25                 p2++;
26             }
27             // right used up
28             else if(p2 == right - middle){
29                 arr[p] = temp[p1 + left];
30                 p1++;
31             }
32             // both not used up
33             else if(temp[p1 + left] < temp[p2 + middle]){
34                 arr[p] = temp[p1 + left];
35                 p1++;
36             }
37             else{
38                 arr[p] = temp[p2 + middle];
39                 p2++;
40             }
41             p++;
42         }
43     };
44     helper(0, arr.size());
45 }
```

Occupy a lot of memory!

Complexity analysis:

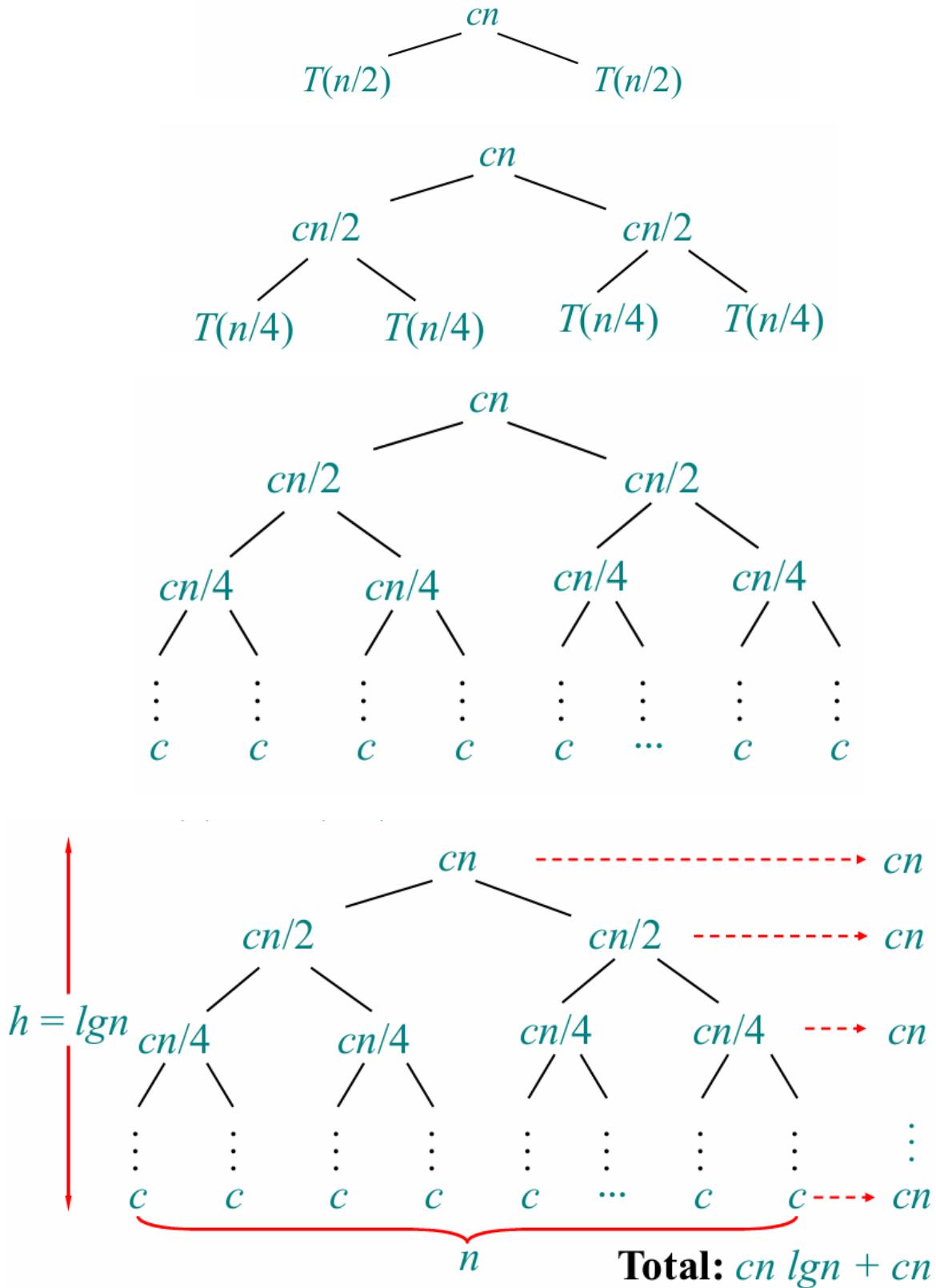
Let's discuss the time complexity of the merge sort:

$$T(N) = 2T\left(\frac{N}{2}\right) + cn$$

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

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

We draw the recursive tree:



Part 2. Complexity Theory

1. Kind of analyses

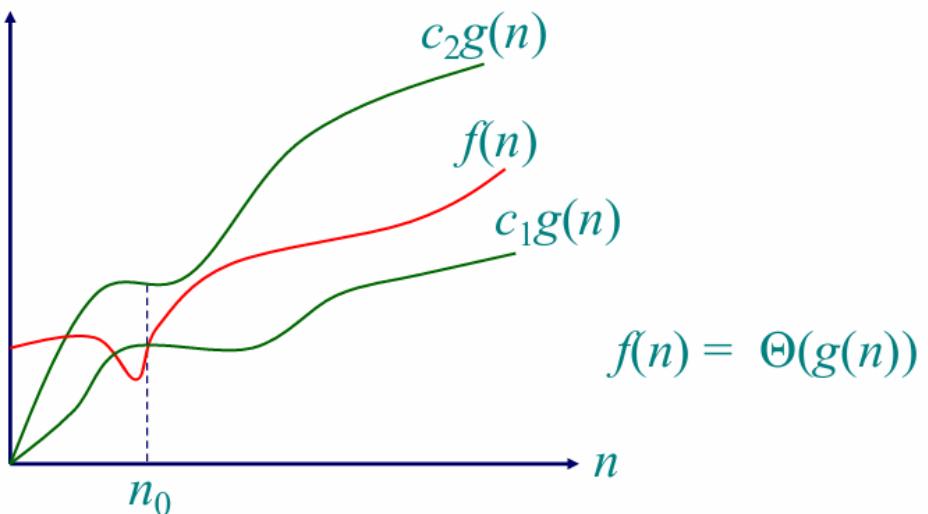
- Worst case (usually)
- Average case (sometimes)
- Best case (bogus, no meaning)

2. Θ, O, Ω Notation

1. Math definition:

a. Θ Notation

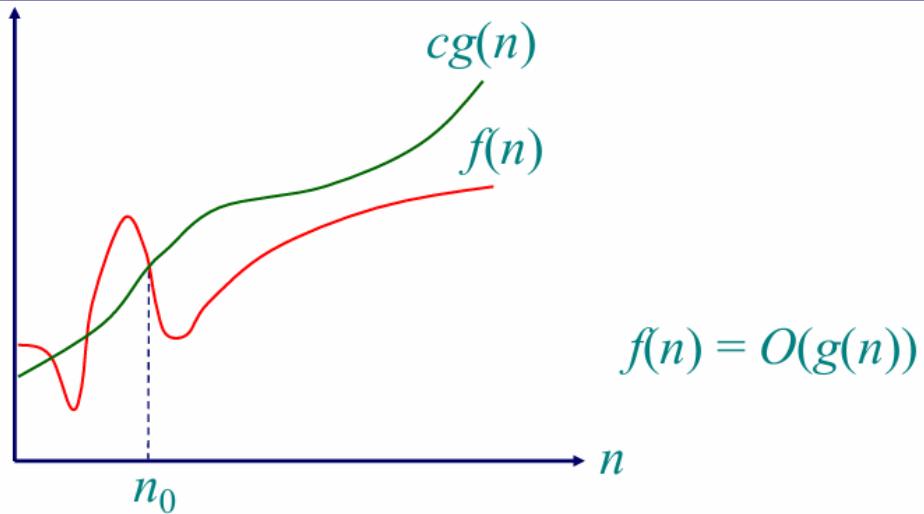
$\Theta(g(n)) = \{f(n) : \text{there exist positive constants } c_1, c_2, \text{ and } n_0 \text{ such that } 0 \leq c_1g(n) \leq f(n) \leq c_2g(n) \text{ for all } n \geq n_0\}$



$f(n) = \Theta(g(n))$ can be thought of $f(n) \approx g(n)$ in their scales

b. O Notation

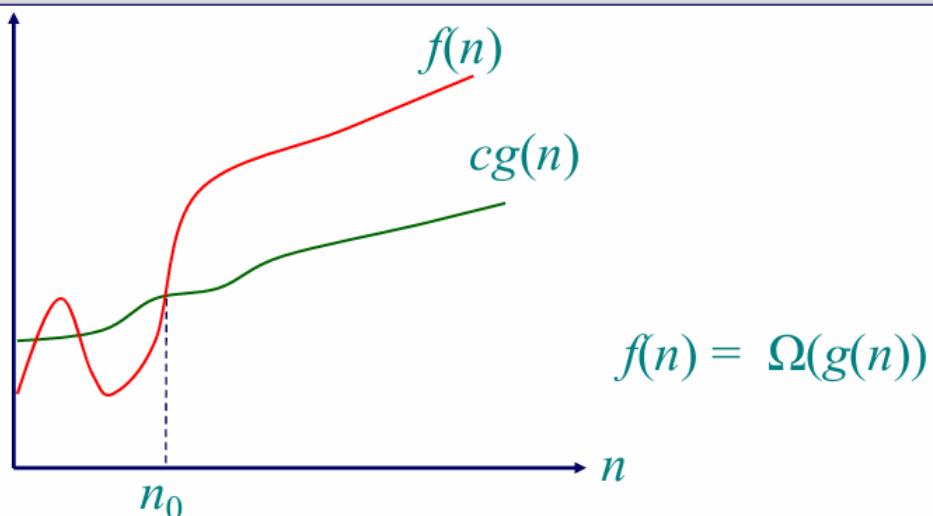
$O(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq cg(n) \text{ for all } n \geq n_0\}$



$f(n) = O(g(n))$ can be thought of $f(n) \leq g(n)$ in their scales

c. Ω Notation

$\Omega(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0\}$



$f(n) = \Omega(g(n))$ can be thought of $f(n) \geq g(n)$ in their scales

2. Engineering meaning:

When we **drop low order terms, ignore leading constants** when $n \rightarrow \infty$

For example: $3n^3 + 90n^2 - 5n + 6046 = \Theta(n^3)$