# Complexity

Ch 1

## Two algorithms. Which one has better performance?

#### Selection sort

```
for (i = 0; i < n-1; i++)
     { /*Find the minimum element in a[i:n-1]*/
        int t = i;
       for (j = i+1; j < n; j++)
          if (a[j] < a[t])
            t = j;
        /* Swap the minimum element with the a[i]*/
        if(t != i)
                                     Select the minimum
          swap(&a[t], &a[i]);
                                     element in unsorted
                         exchange
                                     part
a[0]
            a[i-1] a[i]
                                      a[t]
                                               a[n-1]
                             Unsorted part
   Sorted part
```

#### Insertion sort

```
for (i = 1; i < n; i++)
{/* insert a[i] into a[0:i-1] */
  int t = a[i];
  int j;
  for (j = i - 1; j >= 0 \&\& t < a[j]; j--)
    a[j + 1] = a[j];
                              Insert a[i] into the
  a[j + 1] = t;
                              sorted part
           a[0]
                         a[i-1] a[i]
                                               a[n-1]
              Sorted part
                                  Unsorted part
          a[0] a[j+1] a[i-1] a[i]
                                               a[n-1]
              Sorted part
                                     Unsorted part
```

# Performance analysis

Obtaining estimates of time and space that are machine independent.

Space complexity:
 The amount of memory that it needs to run to completion.

Time complexity:
 The amount of computer time that it needs to run to completion.

## Space complexity

$$S(P) = c + S_p(I)$$

### Total space requirement



= Fixed space requirement + Variable space requirements

Do **not** depend on the number and size of the program *P*'s input and output. Depends on the number and size of input and output associated with the instance *I*.

## Example 1

```
float abc(float a, float b, float c)
{
    return a+b+c*c+(a-b+c)/c+5.00;
}
```

This function has only fixed space requirements. The variable space requirement  $S_{abc}(I) = 0$ .

## Example 2

In C, the address of the first element is passed.

```
float sum(float list[], int n)
{
    float tempsum=0;
    int i;
    for (i = 0; i < n; i++)
        tempsum += list[i];
    return tempsum;
}</pre>
```

C passes all parameters by value. In this case, C does not copy the array list. The variable space requirement  $S_{sum}(I) = 0$ .

## Example 3

```
float rsum(float list[], int n)
{
   if (n) return rsum(list,n-1)+list[n-1];
   return 0;
}
```

One recursive call requires K bytes for 2 parameters and the return address. If the initial length of list[] is N,

the variable space requirement  $S_{rsum}(N) = N*K$  bytes.

## Exercise: factorial function *n!*

A function f to compute the factorial of a number n

$$f(n) = n * f(n-1)$$
 for  $n > 1$   
 $f(0) = 1$  for  $n <= 1$ 

Given that n=N, the size of the input parameter=K, and the size of return address=M.

• Q1: What is the variable space requirement for iterative factorial function?

• Q2: What is the variable space requirement for recursive factorial function?



## Exercise: factorial function *n!*

```
    Iterative

                 double iterFact(int n)
                 { int i;
                    double answer;
                    if ((n == 0) | | (n == 1)) return 1.0;
                    answer = 1.0;
                    for (i = n; i > 1; i--)
                       answer *= i;
                    return answer;

    Recursive

                 double recurFact(int n)
                   if ((n==0) || (n==1)) return 1.0;
                   return n*recurFact(n-1);
```

## Time complexity

Total time requirement T(P)



= Compile time + Execution time

Do **not** depend on the instance characteristics.

Depends on the characteristics of instance *I*.

- Use system clock to time the program or
- Count the number of operations

# Count program step (1)

Program step: A segment with execution time independent from the instance characteristics

```
float sum(float list[], int n)
{
    float tempsum=0;
    int i;
    for (i = 0; i < n; i++)
        tempsum += list[i];

return tempsum;
}

step ++; for assignment

step ++; the for loop (i=0~n-1)
    step ++; for assignment

step ++; to return value
</pre>
```

**Total number of steps: 2\*n+3** 

# Count program step (2)

 Program step: A segment with execution time independent from the instance characteristics

```
float rsum(float list[], int n)
{
    if (n)
        return rsum(list,n-1)+list[n-1];

return 0;
}

for n=0
    step ++; for if condition
    step ++; for return

Number of steps: 2

Number of steps: 2n
```

**Total number of steps: 2\*n+2** 

# For given parameters, computing time may be different.

#### Best-case count

Minimum number of steps that can be executed.

#### Worst-case count

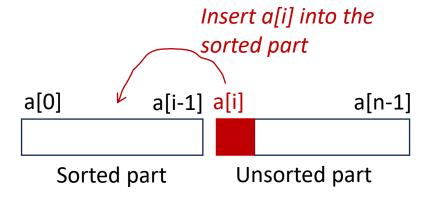
Maximum number of steps that can be executed.

#### Average count

Average number of steps executed.

## Worst-case comparison count

## Insertion sort



for 
$$(j = i - 1; j >= 0 \&\& t < a[j]; j--)$$
  
 $a[j + 1] = a[j];$ 

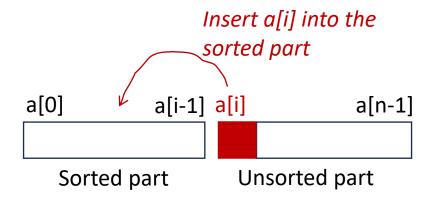
- a[0:i-1] = [1,2,3,4] and t = 0  $\rightarrow$  4 compares
- a[0:i-1] = [1,2,3,...,i] and t = 0  $\rightarrow$  i compares

## For a list in decreasing order:

total compares =  $1 + 2 + 3 + ... + (n-1) = (n-1)n/2 = \frac{1}{2}(n^2) - \frac{1}{2}(n)$ 

## Best-case comparison count

### Insertion sort



for 
$$(j = i - 1; j >= 0 && t < a[j]; j--)$$
  
  $a[j + 1] = a[j];$ 

- a[0:i-1] = [1,2,3,4] and t = 5  $\rightarrow$  1 compare
- a[0:i-1] = [1,2,3,...,i] and t = i+1  $\rightarrow$  1 compare

## For a list in increasing order: total compares = 1 + 1 + 1 + ... + 1 = n-1

# Asymptotic complexity

 Sometimes determining exact step counts is difficult and not useful.

Using big-O to represent space and time complexity

 $\frac{1}{2}(n^2) - \frac{1}{2}(n)$  vs.  $n^2 + 3n - 4$ : which one is better?

- Both are  $O(n^2)$
- There performance are similar.

# Time complexity of insertion sort

Worst case

When the list is in decreasing order

$$\frac{1}{2}(n^2) - \frac{1}{2}(n) \rightarrow O(n^2)$$

Best case

When the list is in increasing order

$$n-1 \rightarrow O(n)$$

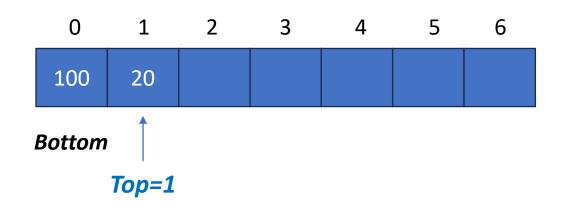
# How do various functions grow with *n*?

	linear		quadratic	cubic	exponential
log n	n	n log n	n²	n³	<b>2</b> <sup>n</sup>
0	1	0	1	1	2
1	2	2	4	8	4
2	4	8	16	64	6
3	8	24	64	512	256
4	16	64	256	4096	65536
5	32	160	1024	32768	4294967296

On a computer performing 1 billion (10<sup>9</sup>) steps/second 2<sup>40</sup> steps require 18.3 mins, 2<sup>50</sup> steps require 13 days, and 2<sup>60</sup> steps require 310 years.

Sometimes, improving algorithm may be more useful than improving hardware.

## Stacks: Using variable top to implement operations



- IsEmpty(): check whether top >= 0
- IsFull(): check whether top == MAX\_STACK\_SIZE-1
- Top(): if not empty, return stack[top]

Time complexity

O(1)

O(1)

O(1)

## Exercise

#### Selection sort

```
for (i = 0; i < n-1; i++)
     { /*Find the minimum element in a[i:n-1]*/
       int t = i;
       for (j = i+1; j < n; j++)
          if (a[j] < a[t])
            t = j;
        /* Swap the minimum element with the a[i]*/
        if(t != i)
                                     Select the minimum
          swap(&a[t], &a[i]);
                                     element in unsorted
                         exchange
                                     part
a[0]
            a[i-1] a[i]
                                     a[t]
                                              a[n-1]
                             Unsorted part
   Sorted part
```



- Q4: Given a list in <u>increasing</u> order, what is the time complexity of selection sort?
- Q5: Given a list in <u>decreasing</u> order, what is the time complexity of selection sort?