

# 1. Basic Concepts

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# Agenda

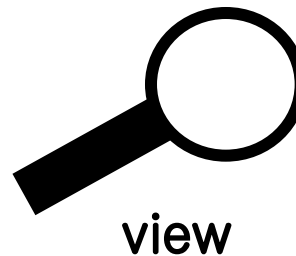
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- **System life cycle**
- **Pointers and dynamic memory allocation**
- **Algorithm specification**
- **Recursion**
- **Data abstraction**
- **Performance analysis**

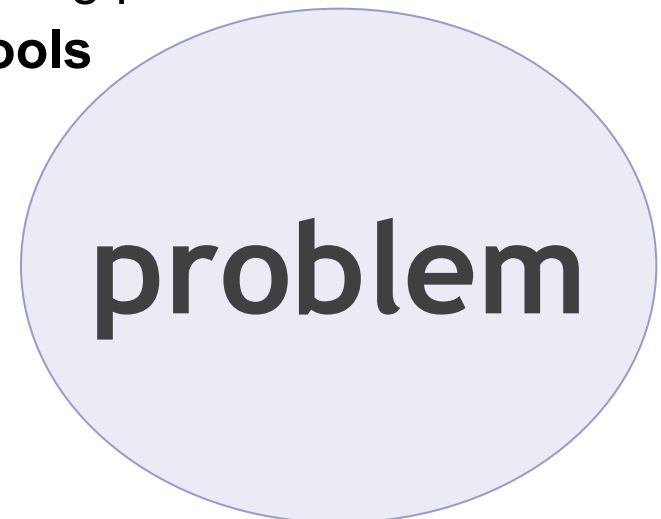
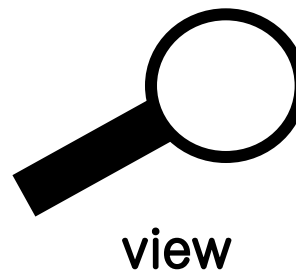
# Overview

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- Building a small program
  - Just do it !

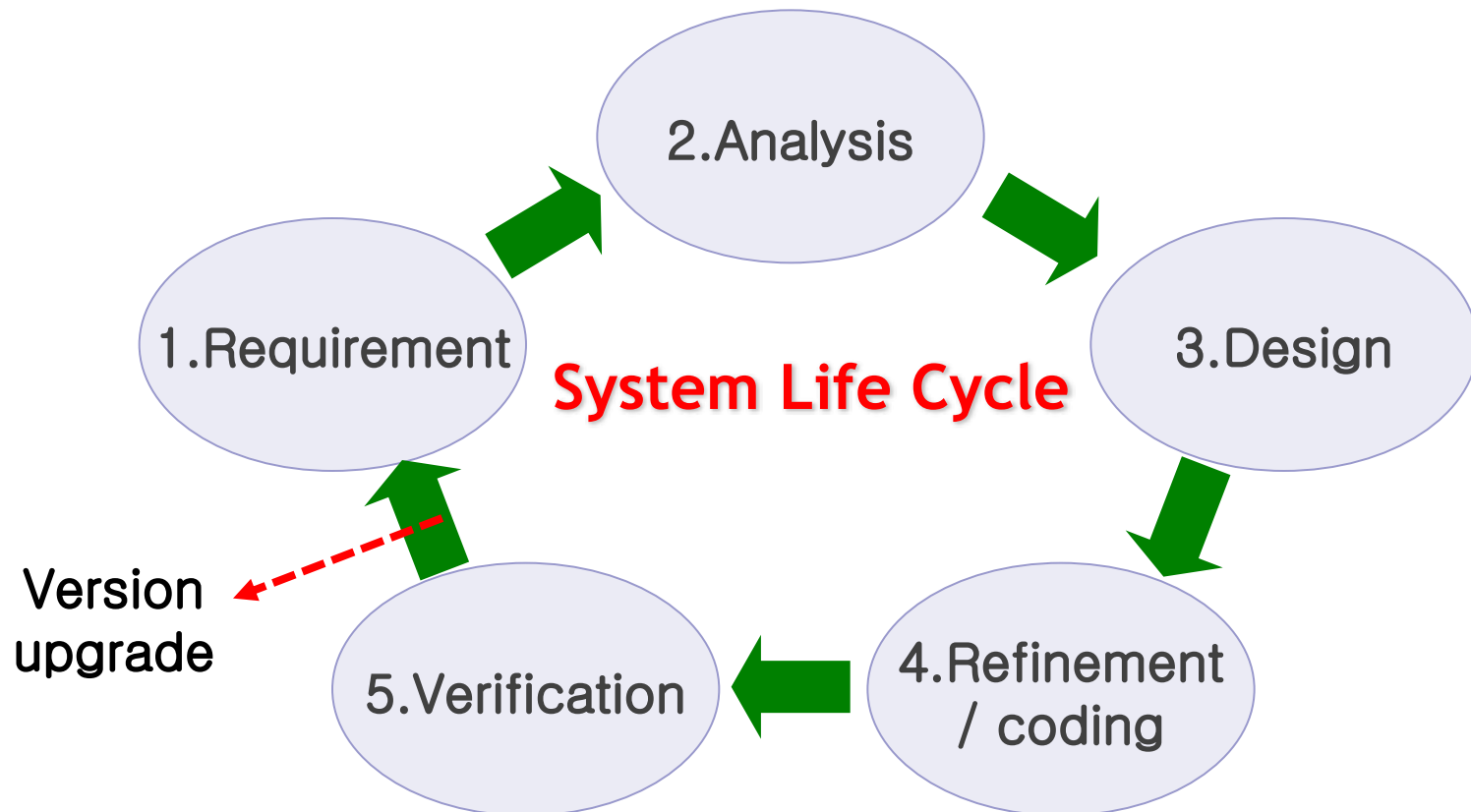


- Building a large-scale system
  - A system composed of complex interacting parts
  - Requires **systematic approach** and **tools**
    - Objective of Data Structures



# System Life Cycle

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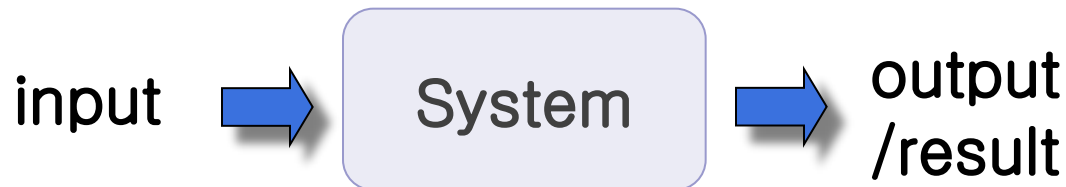
# 1. Requirement

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- Define **purpose/goal** of system



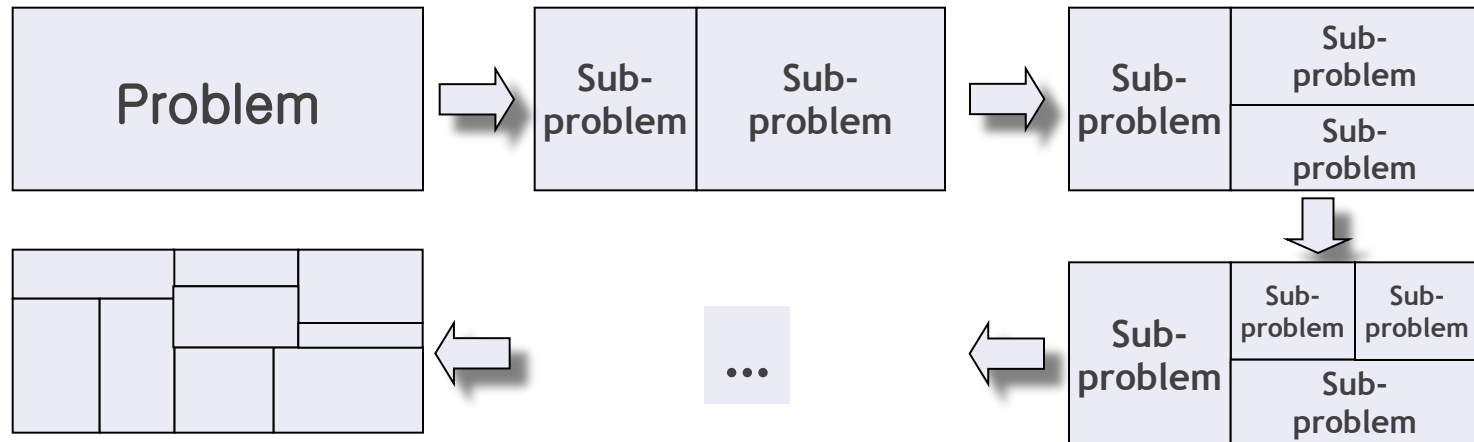
- Define input, output of system
  - Covers all cases
  - Definite/detailed description



## 2. Analysis

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- Break down a problem into manageable pieces
  - Top-down approach: desirable
    - Purpose-driven approach



cf. Bottom-up approach: old, unstructured strategy

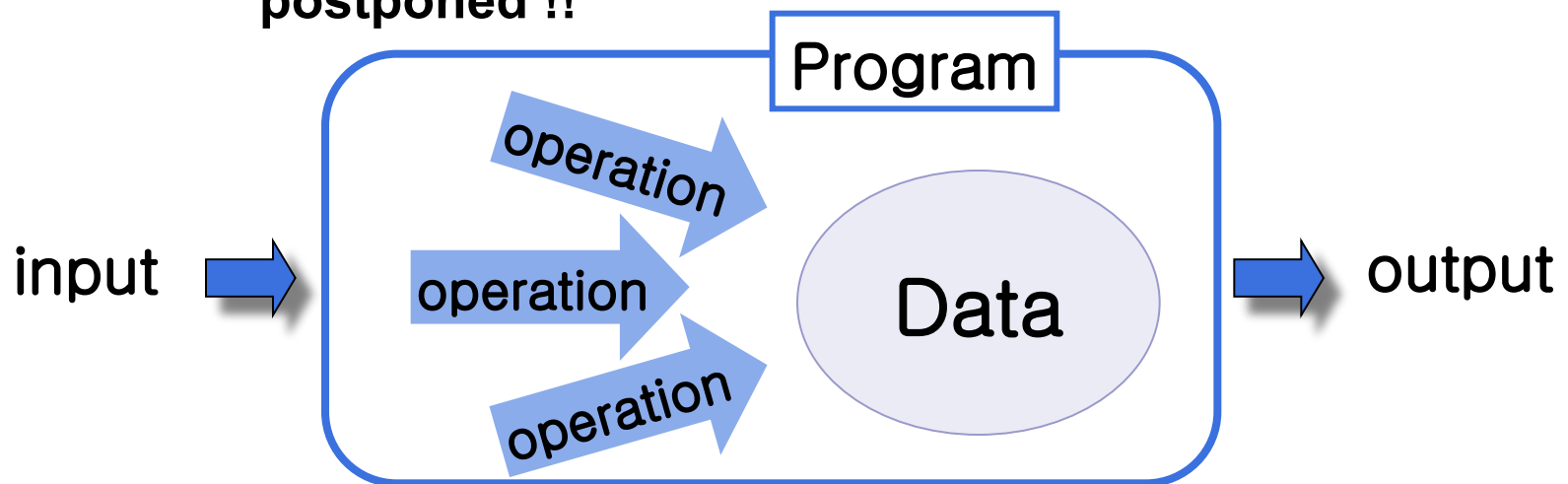
- Building from general walls, roof, plumbing, heating

### 3. Design

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- Find solution from perspective of **data objects** and **operations** on them
    - Data objects: ADT (Abstract Data Type)
    - Operations: **specification** of algorithm
- If the problem is broken-down into manageable pieces, design is easy.

**Note: Language dependent, implementation decisions are postponed !!**



# 4. Refinement and Coding

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- Implementation
    - Actual representations for data objects
    - Algorithms for each operation

→ data representation first, and then algorithms
  - Refinement with data representations and algorithms
  - Coding
- Backup, and backup!
  - If possible, use a version control program like github.
  - Leave some comments and documentations



# 5. Verification

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- Checking validity of system
  - Correctness proof (mathematical) → usually, very difficult
    - Employing proved algorithms can reduce the number of errors
  - Testing
    - with working code and **well-developed test data**
    - running time is another issue.
  - Error removal
    - Documentation, well-divided structure are very helpful
    - it is hard to find a tiny little thing from spaghetti
- Verification is very important for industry-strong code
  - TDD (Test-Driven Development)
    - Define test data before start to develop a system
    - A new trend of software development

“In 1946, ... Operators traced an error in the Mark II to a moth trapped in a relay, coining the term *bug*.

This bug was carefully removed and taped to the log book. “

“Stemming from the first bug,  
today we call errors or glitches in a program a *bug*”

(from wiki)

# Agenda

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- System life cycle
- **Pointers and dynamic memory allocation**
- Algorithm specification
- Recursion
- Data abstraction
- Performance analysis

# Memory and Variables

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- Main memory
  - **List of cells** to store data or instruction
  - Each cell is identified by its **address**.

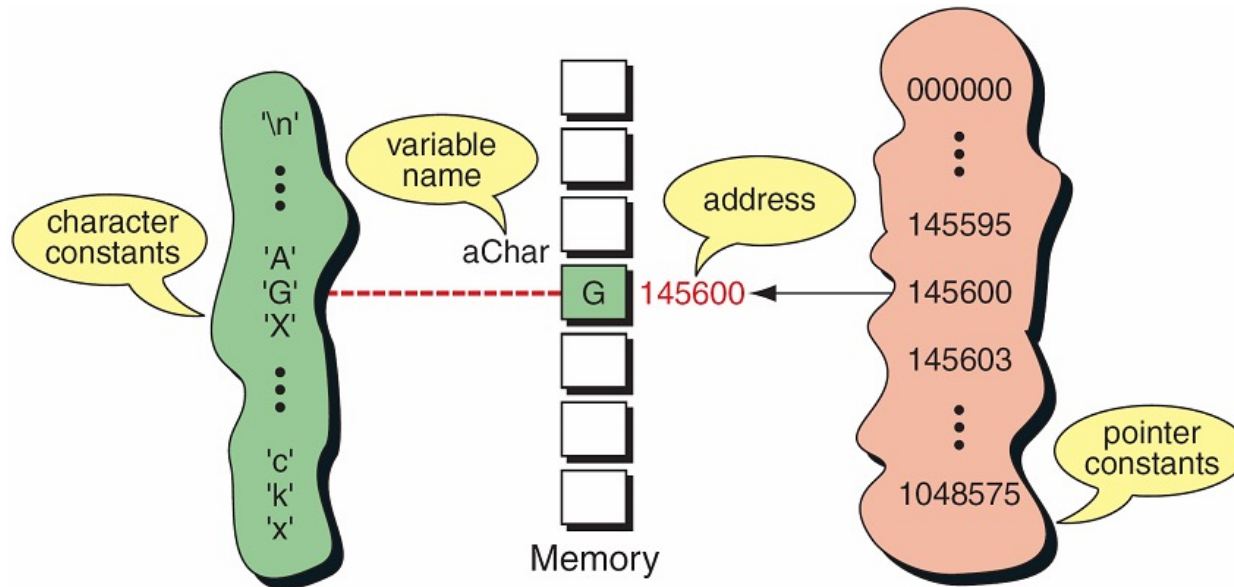
Main memory

0				...
4				...
8				...
12				...
16				...
...	...			

- Variable
  - A variable is a block of memory with a symbolic name  
Ex) `int a = 100;`
  - A variable has name, value(content), and **address**
  - A variable can store a piece of data of a particular type.

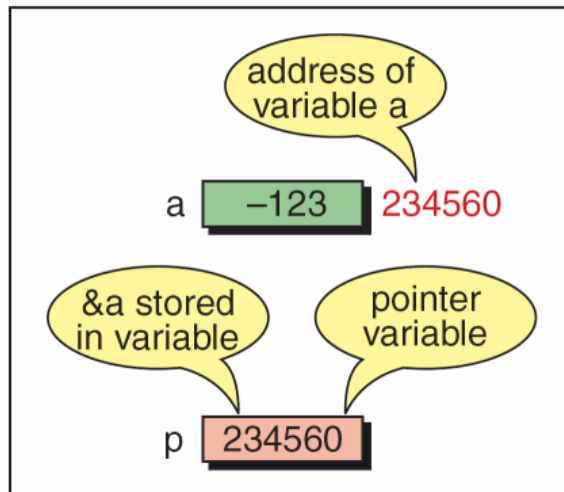
# Pointers

- **Pointer**: constant or variable that contains an address that can be used to access data
  - Range of pointer: address space of computer  
Ex) `char aChar = 'G';`      // assume `&aChar == 1465600`

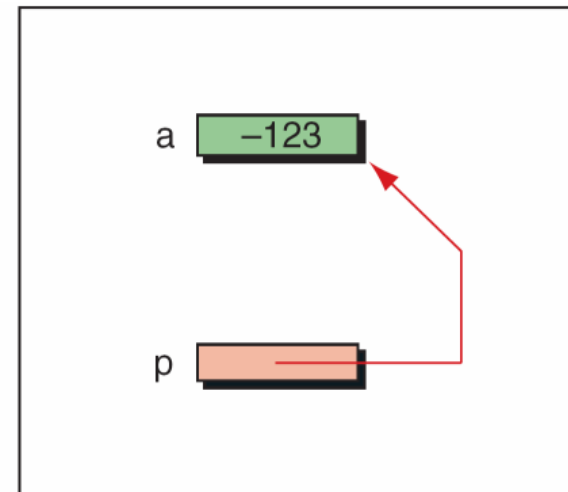


# Pointer Variables

- **Pointer variable**: a variable to store an address



Physical representation



Logical representation

# Using Pointer Variables

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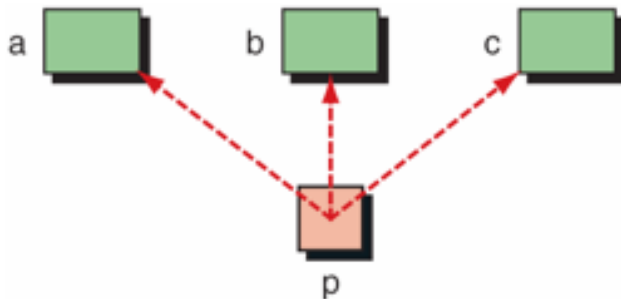
- Declaration
  - `int *pa;`
- Extracting address of a variable (address operator `&`)
  - `pa = &a;`
- Dereferencing (dereferencing operator `*`)
  - `*pa = 89;`
  - `c = *pa * 2;`
- Address operator vs. dereferencing operator
  - `&` is inverse of `*`  
Ex) `*&a ≡ a;`      `// * and & cancel each other`  
cf. How about `&*a` ?
- Arithmetic operations on pointers
  - addition, subtraction, multiplication, and division

# Flexibility of Pointer

- Pointing different variables

```
int a = 10, b = 20, c = 30;  
int *p;
```

```
p = &a;  
printf("*p = %d\n", *p);  
p = &b;  
printf("*p = %d\n", *p);  
p = &c;  
printf("*p = %d\n", *p);
```

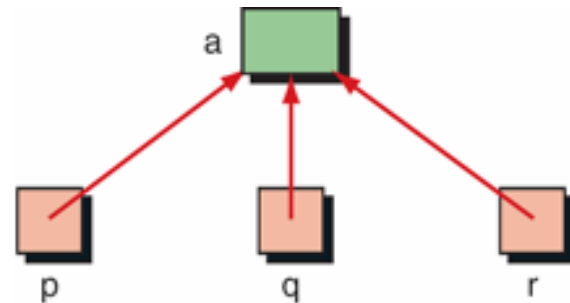


- Multiple pointers for a variables

```
int a = 10;  
int *p, *q, *r;
```

```
p = q = r = &a;
```

```
printf("*p = %d\n", *p);  
printf("*q = %d\n", *q);  
printf("*r = %d\n", *r);
```





# Example

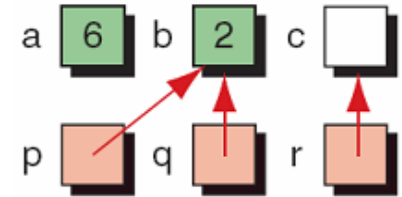
int a, b, c;    a  b  c 

int \*p, \*q, \*r;    p  q  r 

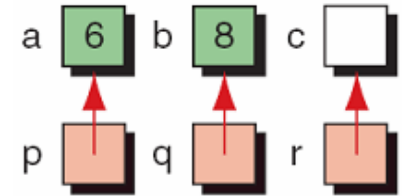
a = 6; b = 2;    a  b  c 

p = &b;    p  q  r 

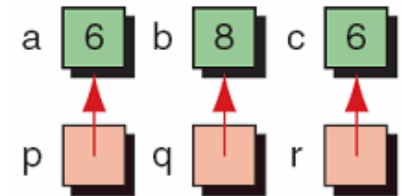
q = p; r = &c;



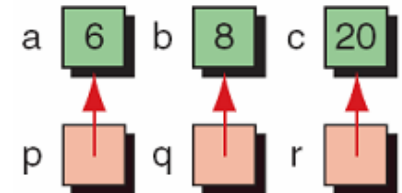
p = &a; \*q = 8;



\*r = \*p;



\*r = a + \*q + \*&c;



# Pointers for Inter-Function Communication

- Passing addresses

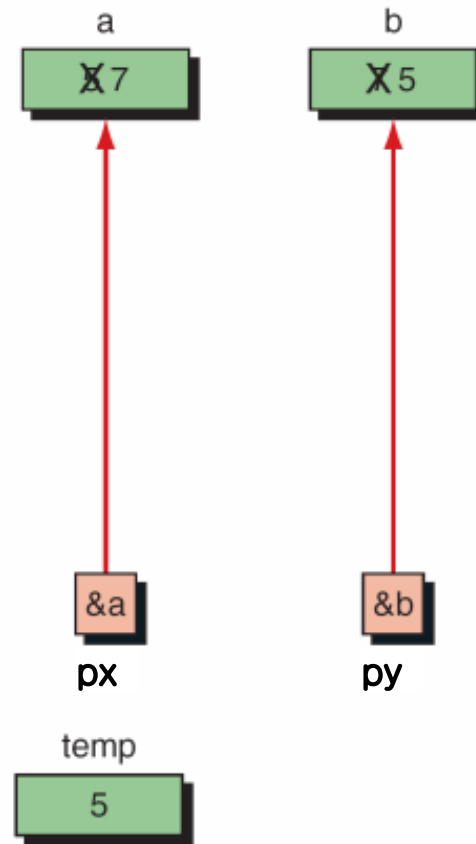
```
// Function Declaration
void exchange (int*, int*);

int main (void)
{
    int a = 5;
    int b = 7;

    exchange (&a, &b);
    printf("%d %d\n", a, b);
    return 0;
} // main
```

```
void exchange (int* px, int* py)
{
    int temp;

    temp = *px;
    *px = *py;
    *py = temp;
    return;
} // exchange
```



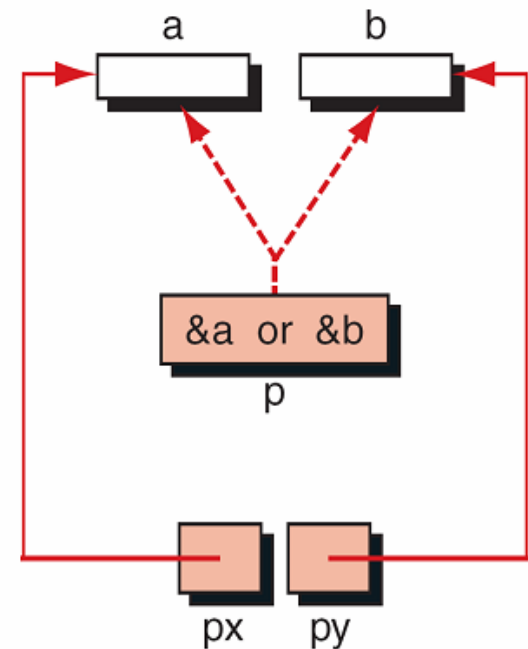
# Pointers for Inter-Function Communication

- Functions returning pointers

```
// Prototype Declarations
int* smaller (int* p1, int* p2);

int main (void)
...
int a;
int b;
int* p;
...
scanf ( "%d %d", &a, &b );
p = smaller (&a, &b);
...
```

```
int* smaller (int* px, int* py)
{
    return (*px < *py ? px : py);
} // smaller
```



# example

---

- what is the output of the following code segment?  
assume that the keyboard inputs are "1 2 3 4 5"

```
int main()
{
    int i, a[5];
    int *p=a;

    for(i=0; i<5; i++, p++)
        scanf("%d", p);

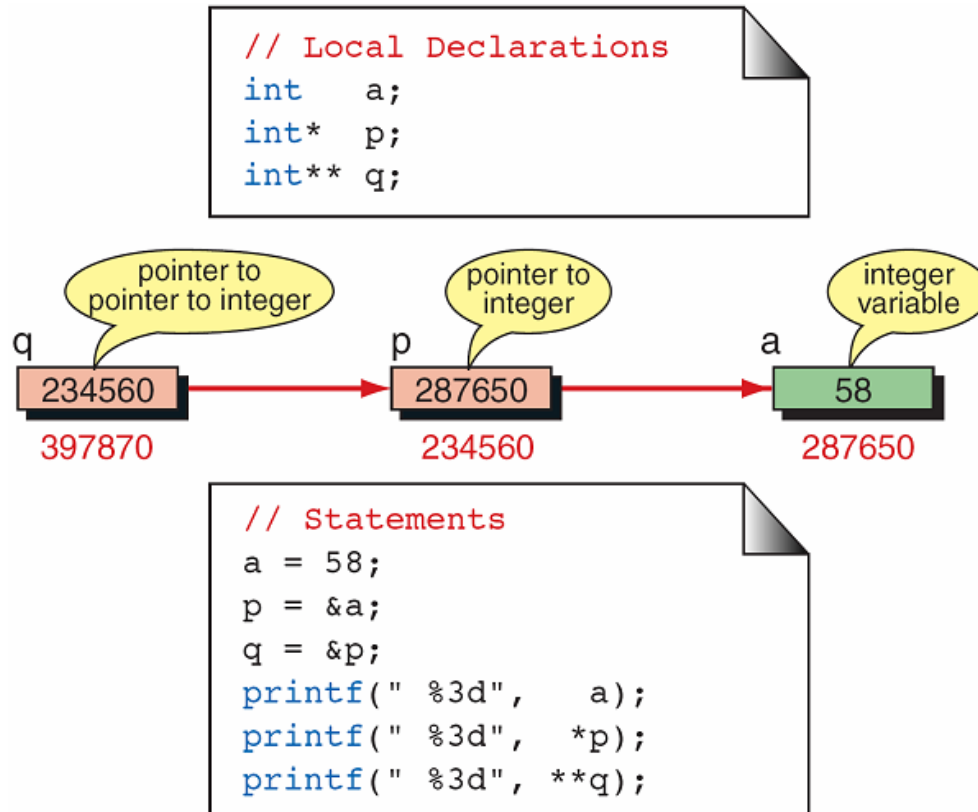
    for(i=0; i<5; i++)
        printf("a[%d]: %d\n", i, a[i]);

    return 0;
}
```

**practice**

# Pointers to Pointers

- **Pointer to pointer (double pointer):** a pointer that points a pointer variable
  - Note! Pointer variable itself occupies memory space



# Example: Double Pointers

---

Exchange pointer variables

```
void ExchangePointers(int **pa, int **pb){  
    int *temp = *pa;  
    *pa = *pb;  
    *pb = temp;  
}
```

```
int main(){  
    int a = 10, b = 20;  
    int *p1 = &a, *p2 = &b;  
  
    ExchangePointers(&p1, &p2);  
    printf("*p1 = %d, *p2 = %d\n", *p1, *p2);  
}
```

**practice**

example (value, addr)

a: 10, 160      b: 20, 180

p1: 160, 240      p2: 180, 260

pa: 240, 400      pb: 260, 404

# Pointers to Pointers

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- Triple pointer

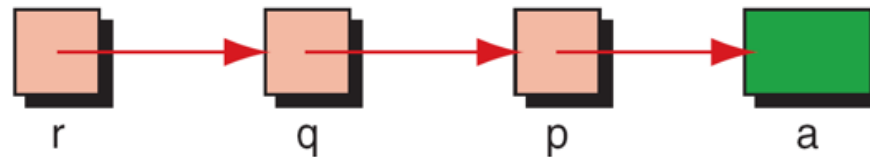
```
int a = 0;
```

```
int *p = &a; // same with int *p; p = &a;
```

```
int **q = &p;
```

```
int ***r = &q;
```

```
// Note  $a \equiv *p \equiv **q \equiv ***r$ 
```



# Compatibility

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- Pointer type compatibility

- A pointer variable can store a pointer of the same type.

Ex) char c, \*pc;

int a;

pc = &c;        // no problem

pc = &a;        // prohibited

- Pointer size compatibility

- Although size of a variable vary with types, **size of all pointers are the same.**

- int i, \*pi;

- char c, \*pc;

- float f, \*pf;

sizeof(i) ≠ sizeof(c) ≠ sizeof(f)

**sizeof(pi) = sizeof(pc) = sizeof(pf)**



# Pointer to Void

---

- **void type pointer** (`void *`) is just to store a **generic address**
  - A generic type that is not associated with a reference type
- void pointer can store any type of pointers

```
void *vp;  
int a;  
char c;  
vp = &a;    // assigning integer pointer to vp  
vp = &c;    // assigning character pointer to vp
```
- **NULL** pointer
  - **NULL** is defined by **(void\*)0**, in `stdio.h`
  - Frequently used to initialize pointer variables

# Pointer to Void

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- void pointer cannot be dereferenced as it is

```
int a = 10;
```

```
void *pVoid = &a;
```

```
*pVoid = 10; // illegal
```

To be dereferenced, void pointer should be casted.

- void pointer can be dereferenced by **casting**

```
int a = 10;
```

```
void *pVoid = &a;
```

```
printf("(*(int)pVoid = %d\n", *(int*)pVoid);
```

- Arithmetic operations are not available (+, -, [], ...) **why not?**

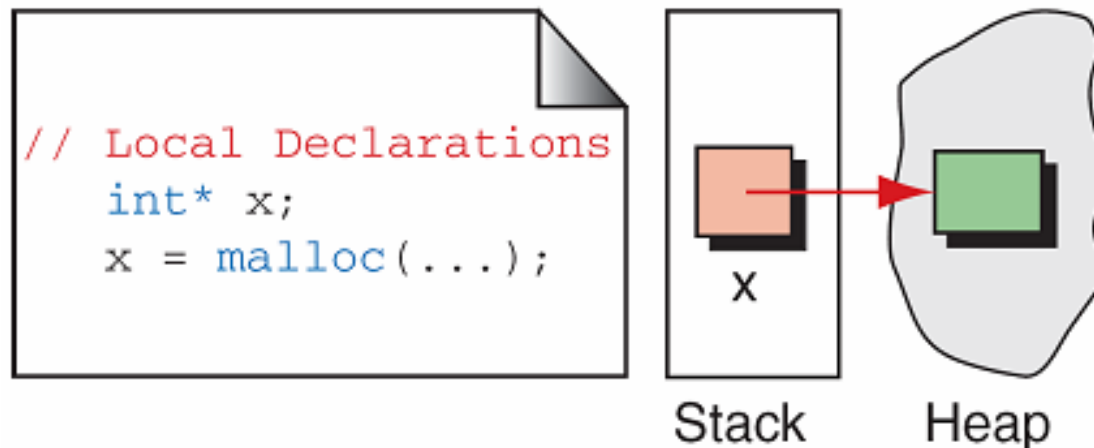
```
vp = vp + 1; // error
```

```
vp[2] = 0;    // error
```

# Dynamic Memory Allocation

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- Dynamic memory allocation: acquiring memory space to store information.
  - Memory allocation using predefined allocation functions
    - Size is dynamically determined
    - Allocated from heap



# Dynamic Memory Allocation

---

- Heap
  - Allocating memory
    - **malloc()** function
      - Size of memory block
  - Using memory
    - Address (pointer)
    - **\* or [] operator**
  - Releasing memory
    - **free()** function
      - Address of the memory block
- Bank
  - Getting a loan
    - Loan application form
      - Amount of money
  - Using money
    - Account number
    - cash card
  - Repaying loan
    - Repayment application form
      - Borrowed money

# Dynamic Memory Allocation

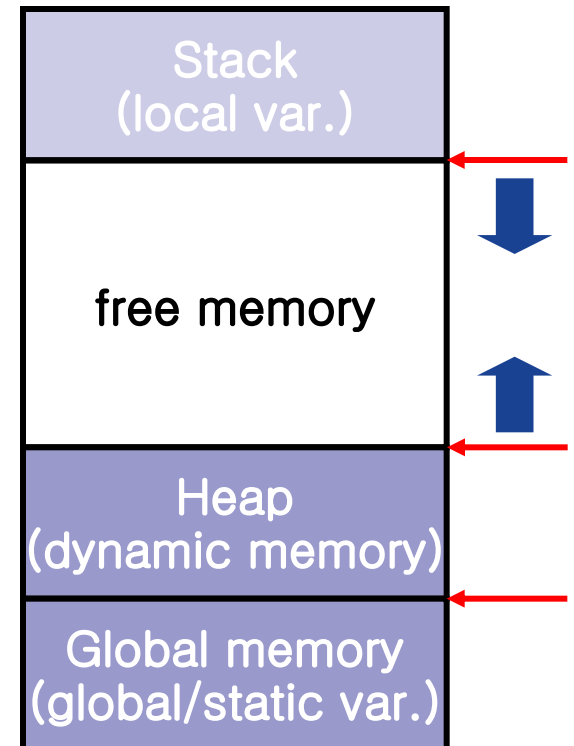
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- Dynamic memory allocation: request for memory space
  - Declared in malloc.h
  - Allocation: `void *malloc(size_t NBYTES);`
    - Usually, `size_t` is defined as `unsigned int`
  - Deallocation: `void free(void *APTR);`

```
int *pi;  
float *pf;  
pi = (int*)malloc(sizeof(int));  
pf = (float*)malloc(sizeof(float));
```

```
*pi = 1024;    // use of pi, pf  
*pf = 3.14;
```

```
free(pi);  
free(pf);
```



# Static vs. Dynamic Memory Allocation

---

- Singleton variable

```
int a = 0;  
int *pi = &a;
```

- Array

```
float fa[100];  
float *pfa = fa;
```

- Singleton variable

```
int *pi = (int*)malloc(sizeof(int));
```

- Array

```
float *pfa =  
(float*)malloc(size*sizeof(float));
```

# Static vs. Dynamic Memory Allocation

---

- Structure

```
struct Time {  
    int hour, min, sec;  
};
```

```
struct Time curTime;  
struct Time *pCurTime =  
    &curTime;
```

```
pCurTime->hour = 10;
```

- Structure

```
struct Time {  
    int hour, min, sec;  
};
```

```
struct Time *pCurTime = (struct Time*)  
    malloc(sizeof(struct Time));
```

```
pCurTime->hour = 10;
```

# Dynamic Memory Allocation

---

- Dynamically allocated memory can be used for any purpose.
  - Size and type should be specified properly

Ex) A singleton variable

```
int *pi = (int*)malloc(sizeof(int));
```

Ex) An array

```
float *pfa = (float*)malloc(size*sizeof(float));  
// similar to "float pfa[size];", but not the same
```

Ex) A structure variable

```
struct Time {                                // structure definition  
    int hour, min, sec;  
};  
struct Time *pCurTime = (struct Time*)malloc(sizeof(struct Time));  
pCurTime->hour = 10;                        // same with (*pCurTime).hour = 10;
```

- **Dynamically allocated memory block must be deallocated eventually**



# Invalid Use of Pointer

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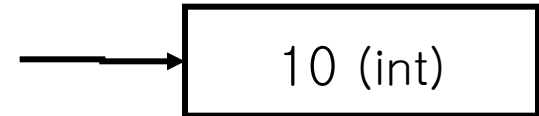
- Invalid type casting

Ex) `int i = 10;`

`int *pi = &i;`

`float *pf = (float*) pi;`      // semantic error

float  
pointer

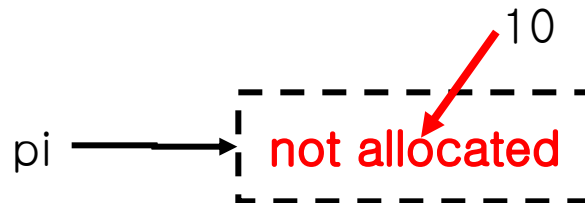


- Unassigned pointer

`int *pi;`

`// pi = (int*) malloc(sizeof(int));`      // forgot

`*pi = 10;`      // error



# Invalid Use of Pointer

- **Dangling pointer** a pointer that points to dynamic memory that has been deallocated

```
int *pi = (int *)malloc(sizeof(int));
```

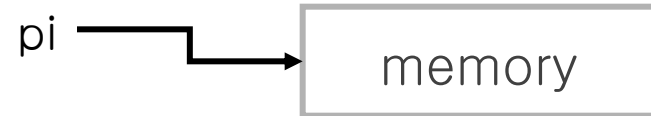
```
*pi = 10;    // valid use
```

```
...
```

```
free(pi);
```

```
...
```

```
free(pi);    // error: pi is already deallocated
```



- **Memory leak** the loss of available memory space that occurs when dynamic data is allocated but never deallocated

```
{
```

```
int *pi;
```

```
pi = func(10);
```

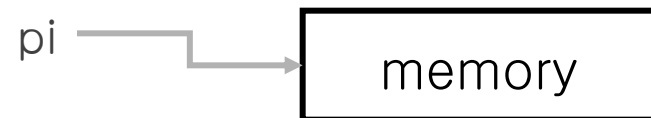
```
pi[0] = 10;
```

```
...
```

```
// free(pi); // forgot
```

```
}
```

```
int *func(int len)
{
    int *a = malloc(len*sizeof(int));
    return a;
}
```



# Safe Coding Practices

---

- Initialize every pointer at declaration

Ex)

```
int *pi;           // bad
```

```
int *pi = NULL;    // good
```

- Set deallocated pointer variable by NULL

```
free(pi);
```

```
pi = NULL;         // free(NULL) is safe
```

# string conversion

---

implement a function 'convert\_case' to convert letter cases  
ex) "Hello World" → "hELLO wORLD"

```
#include <string.h>
#include <stdio.h>
#include <ctype.h>
#include <stdlib.h>

#define MAX_LEN 1024
char* convert_case(char*);
int main()
{
    char aLine[MAX_LEN];
    char *p;

    fgets(aLine, MAX_LEN, stdin);

    p = convert_case(aLine);
    printf("%s\n", p);

    free(p);
    return 0;
}
```

**practice**

do not use any character array  
in convert\_case()  
→ use malloc()

# Agenda

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- System life cycle
- Pointers and dynamic memory allocation
- **Algorithm specification**
- Recursion
- Data abstraction
- Performance analysis

# Algorithm Specification

---

- **Algorithm:** a finite set of instruction that accomplishes a particular task.
  - **Input:** zero or more quantities
  - **Output:** at least one quantity
  - **Definiteness:** clear and unambiguous instructions
  - **Finiteness:** for all cases, the algorithm terminates after a finite number of steps
    - Difference from program (but in this class, they are interchangeable)
  - **Effectiveness:** basic and feasible instructions
- Description of algorithm
  - Natural language
  - Programming language (C source code)
  - Etc.
    - Flow chart, pseudo code, ...

# Example

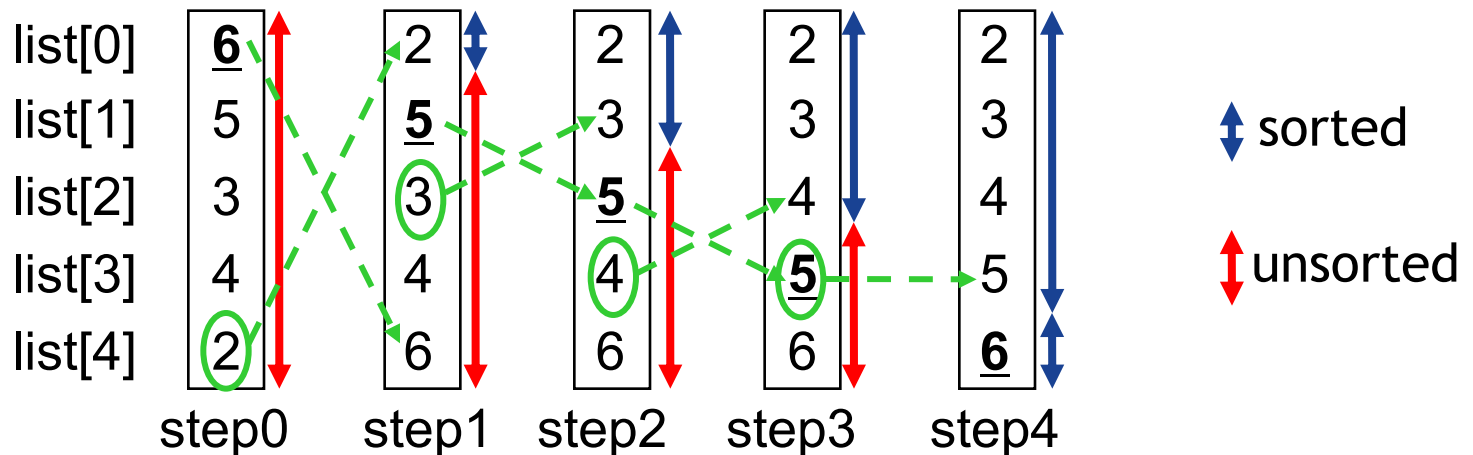
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- How can one put an elephant in a refrigerator?
  - Open the fridge.
  - Push the elephant into the fridge. (effective?)
  - Close the door.

# Example: Selection Sort

- Problem definition: sort  $n$  integers
  - Simple solution
    - From those integers currently unsorted, find the smallest and place it next in the sorted list.
- Not clearly described.

- Selection sorting





# Example: Selection Sort

---

- Algorithm of selection sort

```
for(i = 0; i < n; i++){  
    Examine list[i] to list[n-1] and suppose the smallest integer is at  
    list[min];  
    Exchange list[i] and list[min];  
}
```

→ *Each step is clearly defined.*

# Example: Selection Sort

---

- Implementation of selection sort in C language

```
void sort(int list[], int n)
{
    int i = 0, j = 0, min = 0, temp = 0;

    for(i = 0; i < n - 1; i++){
        // find the minimum of list[i] through list[n-1]
        min = i;
        for(j = i + 1; j < n; j++){
            if(list[j] < list[min])
                min = j;
        }

        // exchange list[i] and list[min]
        temp = list[i];
        list[i] = list[min];
        list[min] = temp;
    }
}
```

**practice**

```
swap(&list[i], &list[min]);
```

```
void swap(int *x, int *y)
```

```
{
    int temp = *x;
    *x = *y;
    *y = temp;
}
```

# Example: Selection Sort

---

- how to practice?

```
#include <stdio.h>

void sort(int [], int);
int main()
{
    int aList[] = {0, 3, 1, 5, 7, 9, 2};

    sort(aList, 7);

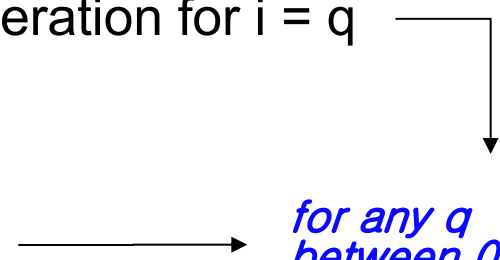
    for(int i=0; i<7; i++)
        printf("%d ", aList[i]);

    return 0;
}
```

**practice**

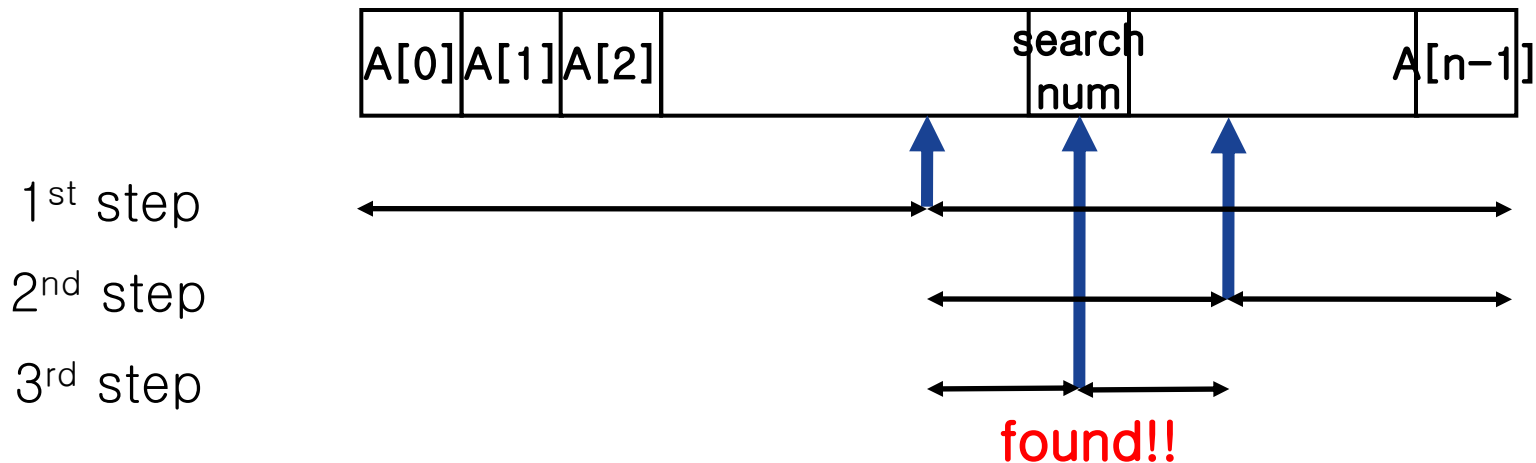
# Example: Selection Sort

---

- Prove that  $\text{sort}(\text{list}, n)$  works correctly.
  1. When the outer loop completes its iteration for  $i = q$  we have  $\text{list}[q] \leq \text{list}[r]$ ,  $q < r < n$ .
  2. On subsequent iterations,  $i > q$ ,  $\text{list}[0]$  through  $\text{list}[q]$  are unchanged. *for any  $q$  between 0 and  $n-2$*
  3. So, on the last iteration of the outer loop ( $i=n-2$ ), we have  $\text{list}[0] \leq \text{list}[1] \leq \dots \leq \text{list}[n-1]$ .

# Example: Binary Search

- Given
  - $n \geq 1$  distinct integers **already sorted** and stored in an array A.
    - $A[0] \leq A[1] \leq \dots \leq A[n-1]$
  - An integer to find (searchnum)
- Problem: find an index, i, such that  $A[i] = \text{searchnum}$ .
  - If searchnum is not present, return -1.



# Example: Binary Search Algorithm

---

1. Let *left* and *right* denote the left and right ends of the list to be searched.
  - Initially,  $left = 0$ ,  $right = n - 1$
2. Let *middle* be the middle point of the list to be searched.
  - $middle = (left + right) / 2$
3. Compare  $A[middle]$  with *searchnum*
  - Case1:  $searchnum < A[middle]$ 
    - If *searchnum* is present, it must be in the left half.  $[left, middle - 1]$
    - Therefore, set *right* to  $middle - 1$
  - Case2:  $searchnum == A[middle]$ 
    - *Searchnum* is found. Return *middle*.
  - Case3:  $searchnum > A[middle]$ 
    - If *searchnum* is present, it must be in the right half.  $[middle + 1, right]$
    - Therefore, set *left* to  $middle + 1$
4. If  $left \leq right$ , go back to 2

# Example: Binary Search Implementation

---

- Binary search in C

practice

```
int binsearch(int list[], int searchnum, int left,
              int right)
{
    /* search list[0] <= list[1] <= . . . <= list[n-1] for
       searchnum. Return its position if found. Otherwise
       return -1 */
    int middle;
    while (left <= right) {
        middle = (left + right)/2;
        switch (COMPARE(list[middle], searchnum)) {
            case -1: left = middle + 1;
                    break;
            case 0 : return middle;
            case 1 : right = middle - 1;
        }
    }
    return -1;
}
```

# Example: Binary Search Implementation

---

```
int main()
{
    int list[] = {1,2,3,4,5,6,7,10,11,17,20,23,25,29,31};
    int target, idx;
    scanf("%d", &target);

    idx = binsearch(list, target, 0, 14);

    if (idx < 0)
        printf("there is no target\n");
    else
        printf("the target is at %d\n", idx);
}
```

**practice**

// 15 items



# Example: Binary Search

---

- COMPARE: a function to compare x and y
  - If  $x < y$ , return -1
  - If  $x == y$ , return 0
  - If  $x > y$ , return +1

```
int COMPARE(int x, int y)
{
    if(x < y)
        return -1;
    else if(x == y)
        return 0;
    else
        return 1;
}
```

# Agenda

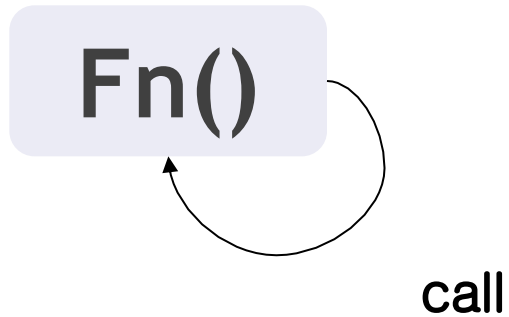
---

- System life cycle
- Pointers and dynamic memory allocation
- Algorithm specification
- **Recursion**
- Data abstraction
- Performance analysis

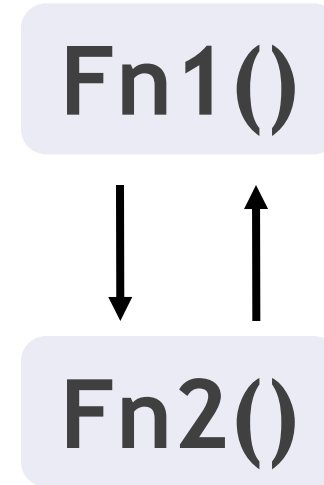
# Recursive Algorithm

---

- Recursion: a function call to itself



< direct recursion >



< indirect recursion >

- “Recursive definition” of recursion
  - Recursion:  
see **Recursion**

# Recursive Algorithm

---

- Why recursion?
  - Extremely powerful
  - Complex processes are often expressed in very clear terms with recursion.
- Example: Recursive implementation of Factorial (x!)

```
int Factorial(int x)
{
    if(x == 1)                // termination condition
        return 1;
    else
        return x * Factorial(x-1);
}
```
- Theoretically, every iteration can be transformed to recursion and vice versa.

# Recursive vs. Iterative Algorithm: Factorial

---

```
int Factorial(int n)
{ // recursive solution
  if (n==0) return 1;
  else
    return n*Factorial(n-1);
}
```

using a selection  
with less local variable.

```
int Factorial(int n)
{ // iterative solution
  int factor, count;
  factor = 1;
  for(count=2; count <=n; count++)
    factor = factor * count;
  return factor;
}
```

using a loop  
with more local variables.

# Recursive Algorithm

---

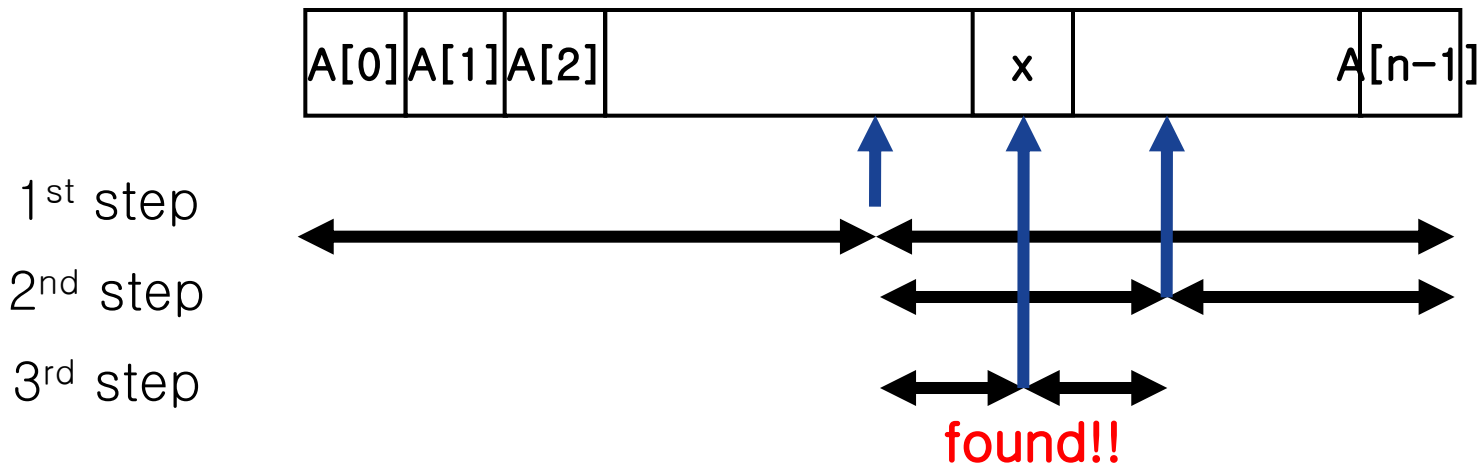
- Problems suitable for recursion
  - Recursively defined problems: problem that can be divided into **the same problem** with **smaller size**
    - Factorial
      - $\text{Factorial}(x) = x * (x-1) * (x-2) * \dots * 1 = x * \text{Factorial}(x-1)$
    - Fibonacci numbers
      - $\text{Fibonacci}(x) = \text{Fibonacci}(x-1) + \text{Fibonacci}(x-2)$
      - $\text{Fibonacci}(0) = \text{Fibonacci}(1) = 1$  → termination condition
    - Binomial coefficients

$$\binom{n}{m} = \binom{n-1}{m} + \binom{n-1}{m-1}$$

$$\binom{n}{k} = \frac{n(n-1) \dots (n-k+1)}{k(k-1) \dots 1}$$

# Example of Recursive Algorithm

- Binary search
  - Problem: find a number  $x$  from a sorted list  $A[]$



# Example of Recursive Algorithm

---

- Recursive algorithm for binary search

```
int binsearch(int A[], int x, int left, int right)
{
    if(left <= right){                                     // termination condition 1
        int mid = (left + right) / 2;
        switch(COMPARE(A[mid], x)){
            case -1: return binsearch(A, x, mid+1, right); // A[mid] < x
            case 0: return mid;                             // termination cond. 2, A[mid] == x
            case +1: return binsearch(A, x, left, mid-1);   // A[mid] > x
        }
    }
    return -1;
}
```

**practice.**  
**compare this to the previous iterative version**



# An Example

---

- Given
  - An array A

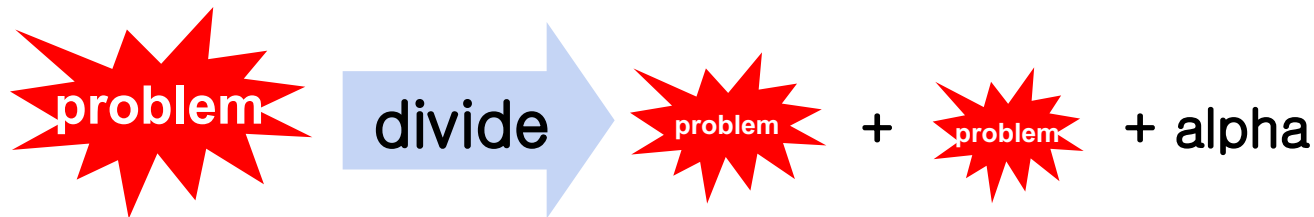
<b>i</b>	0	1	2	3	4	5	6	7	8
<b>A[i]</b>	2	5	6	8	9	13	15	30	54

- A target number  $X = 8$
- Procedure
  - Step1:
  - Step2:
  - Step3:

# Design of Recursive Algorithm

---

1. Find a way to divide a problem into sub-problems whose solution is the same with the original problem with reduced size



2. Describe problem reduction algorithm with recursion general case

Ex)  $F(n) = F(a) + F(b) + \dots + \text{some other part}$

- a, b, ... should be smaller than n

3. Describe non-recursive solution(s) for very simple case(s) base case

Ex)  $F(1)$ ,  $F(0)$

→ termination

# Example: permutation generator

---

- print out all possible permutations of the set (list)

```
void perm(char *list, int i, int n)
{
    int j, temp;
    if (i == n){
        for (j=0; j<=n; j++)
            printf("%c", list[j]);
        printf("\n");
    }
    else{
        for (j=i; j<=n; j++){
            swap(&list[i], &list[j]);
            perm(list, i+1, n);
            swap(&list[j], &list[i]);
        }
    }
}
```

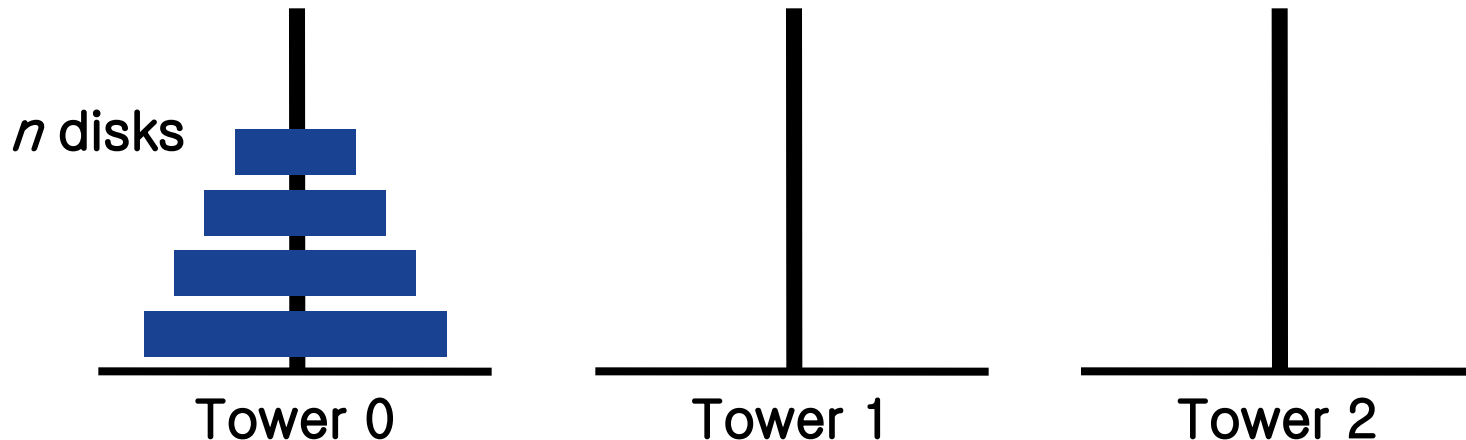
- initial call  
`perm(list, 0, n-1);`

**practice**  
**understand how it works**

# Example: Tower of Hanoi

---

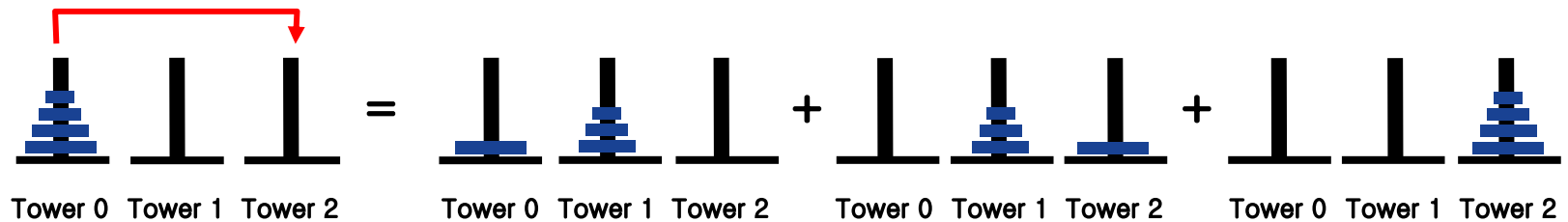
- Input
  - Three towers and  $n$  disks of different diameters
  - placed on the first tower in order of decreasing diameter
- Problem (mission)
  - Move all disks from the first tower to the third tower
- Restrictions
  - Only one disk can be moved at any time
  - No disk can be placed on top of a disk with a smaller diameter



# Towers of Hanoi

---

- Non-recursive solution: difficult and complex
- Solution using recursion
  - Divide the problem into smaller problems
    - Moving  $n$  disks from **tower a** to **tower b** =  
Move  $n-1$  disks from **tower a** to **tower c**  
+ Move **1** disk from **tower a** to **tower b**  
+ Move  $n-1$  disks from **tower c** to **tower b**



- Termination condition
  - If  $n == 1$ , just move a single disk from tower a to tower b

# Towers of Hanoi

---

```
void Hanoi(int n, int begin, int aux, int end)
{
    if (n > 1)
    {
        Hanoi(n-1, begin, end, aux); // from begin to aux
        Hanoi(1, begin, aux, end);    // from begin to end
        Hanoi(n-1, aux, begin, end);  // from aux to end
    }
    else
        printf("move the disk in %d to %d\n", begin, end);
}
```

**practice  
understand how it works**

# Agenda

---

- System life cycle
- Pointers and dynamic memory allocation
- Algorithm specification
- Recursion
- **Data abstraction**
- Performance analysis

# Data Types

---

- Built-in data type of C/C++
  - Element type: char, int, float, double
  - Collection type: array, structure, ...
    - Ex) struct student {  
    char last\_name;  
    int student\_id;  
    char grade;  
};
  - Pointer type: char\*, int\*, void\*, ...
- User defined type



# Data Abstraction

---

- Data type: a collection of **objects** and a set of **operations** that act on those objects

Ex) int type

- Object: numbers in  $\{\text{INT\_MIN}, \dots, -1, 0, 1, \dots, \text{INT\_MAX}\}$ 
  - 16bit integer
    - $\text{INT\_MIN} = -32768, \text{INT\_MAX} = 32767$
  - 32bit integer
    - $\text{INT\_MIN} = -2147483648, \text{INT\_MAX} = 2147483647$
- Operation:  $+, -, *, /, \%$

# Data Abstraction

---

- ADT (Abstract Data Type): data type organized by **specification of objects** and **specification of operation**, NOT including
  - Representation of objects
  - Implementation of operation
  - Focuses on what, not how
  - Necessary for managing large, complex software projects
- Specification of operation
  - Description of what the function does.
    - names, arguments, result of each functions
  - The function call depends on the function's specification (description), not its implementation (algorithm)
    - e.g., double = sqrt(double), or double = pow(double, double)

# Example of ADT

---

- Natural number (Nat\_No)
  - Objects: integers from zero to INT\_MAX
  - Functions (or operations)
    - Nat\_No Zero() ::= return 0
    - Boolean Is\_Zero(x) ::= if (x) return FALSE, else return TRUE
    - Nat\_No Add(x, y) ::= if(x+y <= INT\_MAX) return x+y  
else return INT\_MAX
    - Nat\_No Subtract(x, y) ::= if(x < y) return 0;  
else return x - y;
    - Nat\_No Power(x, y) ::= if ( $x^y < \text{INT\_MAX}$ ) return  $x^y$   
else return INT\_MAX
    - etc

# ADT

---

- Why ADT?
  - Implementation-independent
- ADT frequently includes
  - Creator/constructor: create new instance
  - Transformers: create new instance from one or more other instances
  - Observers/reporters: provides information about instance
  - Destructor: discard instance of ADT (optional)
- we will discuss the specification first, then implementation

# Agenda

---

- System life cycle
- Pointers and dynamic memory allocation
- Algorithm specification
- Recursion
- Data abstraction
- **Performance analysis**

# Performance Analysis

---

- Criteria to evaluate a program
  - Does it meet specification of the task?
  - Does it work correctly?
  - Does it contain documentation about how to use and how it works?
  - Does it effectively use functions to create local units?
  - Is the code readable?

- **Does it efficiently use primary/secondary storage?**

- **Is the running time acceptable?**

**Performance issues**

# Performance Evaluation

---

- Performance analysis
  - Mathematical analysis of algorithm
    - Complexity theory
    - Machine independent
- Performance measurement
  - Execution time on a specific computer
    - Machine-dependent

# Complexities

---

- Space complexity of a program:
  - the amount of memory that it needs to run to completion
- Time complexity of a program:
  - the amount of computer time that it needs to run to completion



# Space Complexity

---

- Fixed space requirements
  - Space requirement independent from number and size of input/output
    - Composed of instruction space, simple var., structures, constants
- Variable space requirements
  - Space requirement dependent on a particular instance  $I$ 
    - $S_p(I)$  - Variable space requirement of *a program  $P$*  working on *an instance  $I$* 
      - $S_p(I)$  is usually a function of **characteristics** of  $I$ 
        - the number, size, values of input and output
    - Ex) input is an array whose size is  $n$ : *instance characteristic*
    - $S_p(I)$  can be replaced by  $S_p(n)$  when  $S_p(I)$  depends on only  $n$
- Total space requirement  $S(P) = c + S_p(I)$

# Examples

---

- Fixed space requirement

```
float abc(float a, float b, float c)
{
    return a+b+b*c + (a+b-c)/(a+b) + 4.00;
}
```

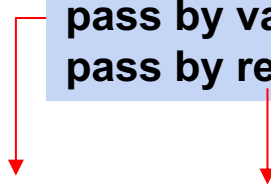
$S_{abc}(I) = 0$

- Variable space requirement

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

$S_{sum}(I) = S_{sum}(n) = (n \text{ in some languages like Pascal}) (0 \text{ in C})$

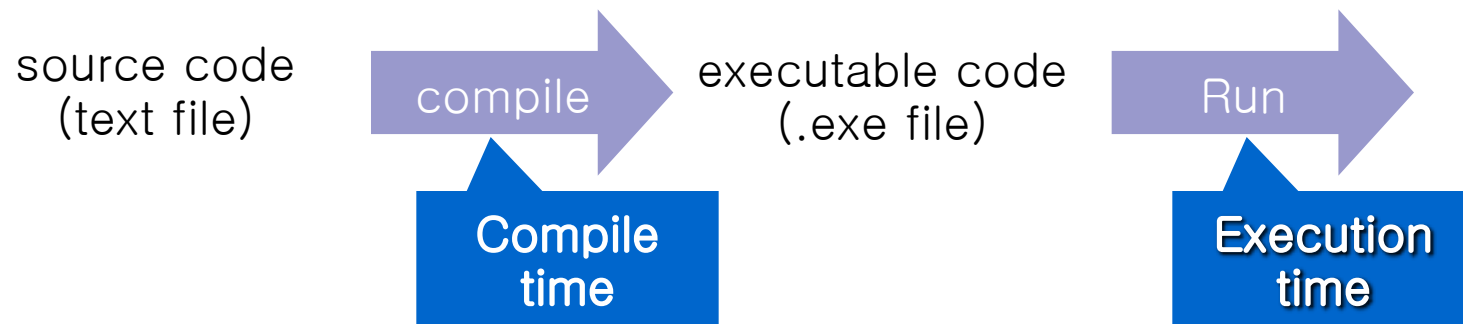
pass by value  
pass by reference



# Time Complexity

---

- Total time taken by a program P
  - $T(P) = \text{compile time} + \text{run time (execution time)}$ 
    - Compile time is not very important
      - Independent from instance characteristics
      - After development is finished, no need for recompile.
    - **Run time ( $T_p$ )**
      - **Major target of complexity analysis**



# Time Complexity

---

- However,  $T_p$  does not depend only on  $P$   
Ex)  $T_p(n) = c_a \text{ADD}(n) + c_s \text{SUB}(n) + c_l \text{LDA}(n) + c_{st} \text{STA}(n)$ 
  - $c_a, c_s, c_l, c_{st}$ : constants
  - Requires knowledge about compiler and H/W For more details, computer architecture class
  - Rarely worthy
- **Program step**: alternative unit to measure execution time
  - Syntactically or semantically meaningful program segment,
  - whose **execute time is independent of the instance characteristics**

Ex) Assignment , comparison, addition, ...

$a = 2$

$a = 2*b+3*c/d-e+f/g/a/b/c$

Both assignments are just one step

# Measuring # of Steps Using Step Count Variable

- Iterative summing of a List of Numbers

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

- Iterative version with count statements

```
float sum(float list[], int n)
{
    float tempsum = 0;
    count++;          /* for assignment */
    int i;
    for (i = 0; i < n; i++) {
        count++;      /* for the for loop */
        tempsum += list[i];
        count++;      /* for assignment */
    }
    count++;          /* last execution of for */
    count++;          /* for return */
    return tempsum;
}
```

- Number of executed steps:  $2n + 3$



# Measuring # of Steps Using Step Count Variable

---

- Recursive version with count statements

```
float rsum(float list[], int n)
{
    count++; /* for if conditional */
    if (n > 1) {
        count++; /* for return and rsum invocation */
        return rsum(list, n-1) + list[n - 1];
    }
    count++;
    return list[0];
}
```

- Number of executed steps:  $2n + 2$
- However, recursive version is usually slower than iterative version because of function-call overhead.

# Measuring # of Steps Using Step Count Variable

---

- Matix Addition

```
void add_mat (int a[ ][MAX_SIZE], int b[ ][MAX_SIZE],
              int c[ ][MAX_SIZE], int rows, int cols)
{
    int i, j;
    for (i=0; i < rows; i++){           // rows + 1
        // count++;
        for(j=0; j < cols; j++){         // rows × (cols + 1)
            // count ++;
            c[i][j] = a[i][j] + b[i][j]; // rows * cols
            // count ++;
        }
        // count ++;
    }
    // count ++;
}
// Total = 2 × rows × cols + 2 × rows + 1
```

# Measuring # of Steps Using Tabular Method

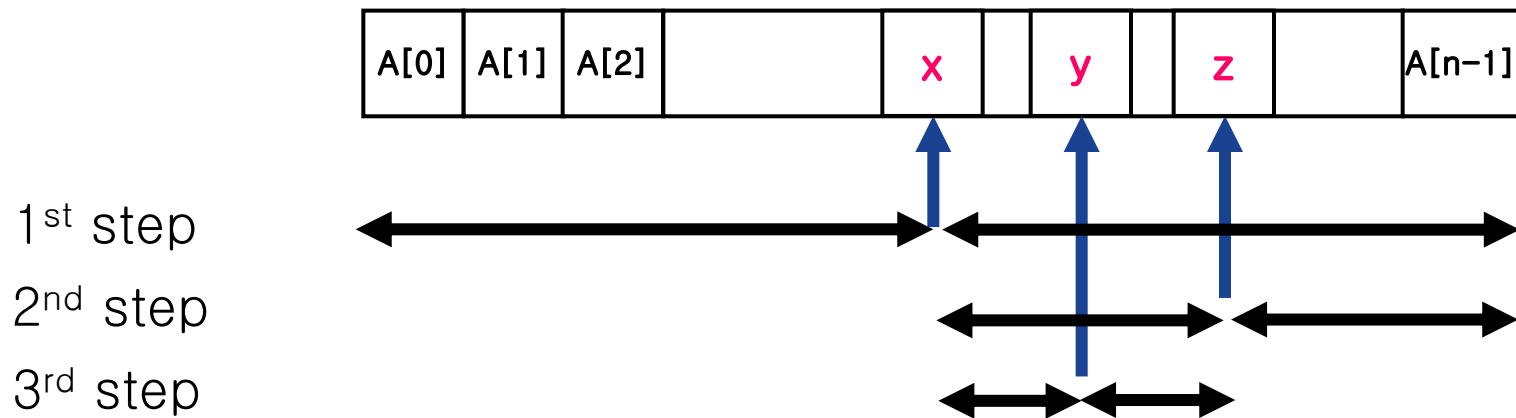
1. Determine **step count** for each statement (steps/execution or s/e)
2. Figure out number of times each statement is executed (**frequency**)
3. Multiply s/e by frequency to get total steps of each statement
4. Sum total steps of all statements

<pre>void add_mat (int a[ ][MAX_SIZE],                int b[ ][MAX_SIZE], int c[ ][MAX_SIZE],                int rows, int cols)</pre>	s/e	Frequency	Total Step
<pre>{</pre>			
<pre>    int i,j;</pre>	0	0	0
<pre>    for (i=0; i &lt; rows; i++)</pre>	1	rows+1	rows+1
<pre>        for(j=0; j &lt; cols; j++)</pre>	1	rows*(cols+1)	rows*(cols+1)
<pre>            c[i][j] = a[i][j] + b[i][j];</pre>	1	rows*cols	rows*cols
<pre>}</pre>	0	0	0



# Three Kinds of Steps Counts

Ex) Step count of binarysearch depends on values of search target and contents of sorted list



- Step count should be estimated for three cases
  - **Best case:** minimum number of steps for execution
  - **Worst case:** maximum number of steps for execution
  - **Average case:** average number of steps for execution

# Limits of Step Count

---

- Problems of step count for performance analysis
  - Increasingly difficult
  - Exact step count is not very necessary.
  - Step count itself is not exact
- Comparing programs
  - We can say  $3n+3$  is faster than  $100n+10$ .
  - But, it is difficult to compare  $80n + 10$  to  $85n$  or  $75n + 20$ .
- So, we need asymptotic notations

# Asymptotic Notation

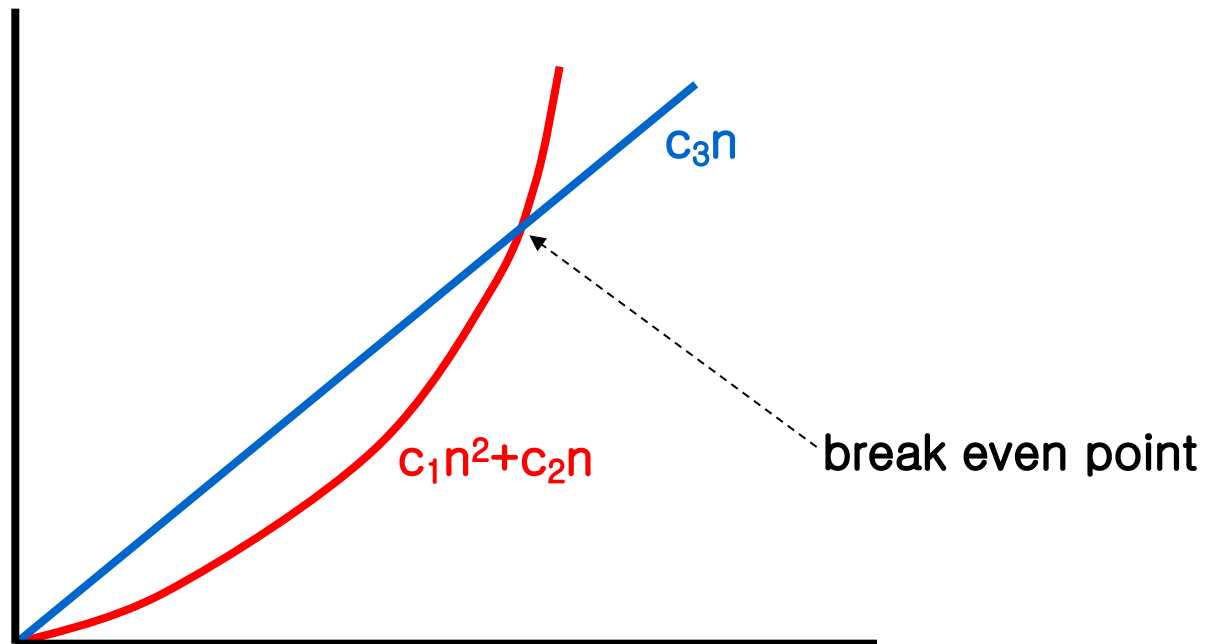
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- More adequate method than step count
    - $c_1n^2 \leq T_p(n) \leq c_2n^2$  or  $T_q(n,m) = c_1n + c_2m$ 
      - $c_1, c_2$  are non-negative constants
  - We can compare  $c_1n^2 + c_2n$  with  $c_3n$  for sufficiently large  $n$ .
    - No matter how much  $c_3$  is bigger than  $c_1$  and  $c_2$ ,  $c_1n^2 + c_2n > c_3n$  if  $n$  is very large
- Ex)  $c_1 = 1, c_2 = 2, c_3 = 100$
- for  $n \leq 98$ ,  $c_1n^2 + c_2n < c_3n$
  - for  $n > 98$ ,  $c_1n^2 + c_2n > c_3n$

# Asymptotic Notation

---

- Break even point: a value of  $n$ , beyond which  $c_3n$  is always faster than  $c_1n^2+c_2n$ , regardless of  $c_1$ ,  $c_2$ , and  $c_3$ 
  - Exact estimation of break even point is difficult and little advantage
  - Knowing whether a break even point exists is sufficient



# Order of Complexities

---

- If the size of data is large, the order of complexity dominates the constant coefficient.
- Order of complexities
  - 3, 100, 35000, ...  $\rightarrow O(1)$
  - $2\log_4(n)$ ,  $6\log_2(n)$ ,  $10\log_8(5n)$   $\rightarrow O(\log n)$
  - $30n$ ,  $65n+30$ ,  $10000n + 329858$ , ...  $\rightarrow O(n)$
  - $9n^2$ ,  $2n^2+100$ ,  $582n^2+28$ , ...  $\rightarrow O(n^2)$
  - $n^3+130$ ,  $9n^3+20$ ,  $128n^3+32$ , ...  $\rightarrow O(n^3)$
  - $2^n$ ,  $5 \cdot 2^n$ ,  $100 \cdot 2^n + n^3 + 100$ , ...  $\rightarrow O(2^n)$

# Polynomial Complexities

---

- $O(n)$ : algorithms with single loops  
    for( $i = 0; i < n; i++$ )  
        sum += list[i]; // executed  $n$  times
- $O(n^2)$ : algorithms with double loops  
    for( $i = 0; i < n; i++$ )  
        for( $j = 0; j < n; j++$ )  
            sum += list\_2D[i][j]; // executed  $n^2$  times
- $O(n^3)$ : algorithms with triple loops  
    for( $i = 0; i < n; i++$ )  
        for( $j = 0; j < n; j++$ )  
            for( $k = 0; k < n; k++$ )  
                sum += list\_3D[i][j][k]; // executed  $n^3$  times
- $O(n^m)$ : algorithms with  $m^{\text{th}}$  order loop

# Log Complexities

---

Ex) Worst case complexity of binary search

n	1	2	3	4	5	6	7	8	...	16	...	32	...
# of comp. (T)	1	2	2	3	3	3	3	4	4	5	5	6	...

- $n \approx 2^{(T-1)}$
- $T \approx \log_2 n$

# Exponential Complexities

---

Ex) # of possible patterns using n bits

n	0	1	2	3	4	5	6	7	8	9	10	11	12
# of possible patterns (T)	1	2	4	8	16	32	64	128	256	512	1024	2048	4096

- $T(n) \approx 2^n$



# Asymptotic Notation

---

- Upper bound complexity

Def) Big “Oh”:  $f(n) = O(g(n))$  iff there exist positive constants  $c$  and  $n_0$  s.t.  $f(n) \leq cg(n)$  for all  $n, n \geq n_0$

- Examples

- $3n+2 = O(n)$ 
  - $3n+2 \leq 4n$  for all  $n \geq 2$
- $100n+6 = O(n)$ 
  - $100n+6 \leq 101n$  for  $n \geq 10$
- $10n^2+4n+2 = O(n^2)$ 
  - $10n^2+4n+2 \leq 11n^2$  for  $n \geq 5$
- $6 \cdot 2^n + n^2 = O(2^n)$ 
  - $6 \cdot 2^n + n^2 \leq 7 \cdot 2^n$  for  $n \geq 4$
- $3n+3 = O(n^2)$ 
  - $3n+3 \leq 3n^2$  for  $n \geq 2$
- $10n^2+4n+2 \neq O(n)$

Note!

$3n+3=O(n)$

$3n+3=O(n^2)$

But,

$O(n) \neq O(n^2)$

# Asymptotic Notation

---

- Big Oh notations
  - $O(1)$ : constant
  - $O(n)$ : linear
  - $O(n^2)$ : quadratic
  - $O(n^3)$ : cubic
  - $O(2^n)$ : exponential

- Comparison

$$O(1) < O(\log n) < O(n) < O(n \log n) < O(n^2) < O(n^3) < O(2^n) < O(n!)$$

Note!  $f(n) = O(g(n))$  doesn't mean  $g(n) = O(f(n))$

# Asymptotic Notation

---

Theorem) If  $f(n) = a_m n^m + \dots + a_1 n + a_0$ , then  $f(n) = O(n^m)$

- Proof

$$f(n) \leq \sum_{i=0}^m |a_i| n^i$$

$$= n^m \sum_{i=0}^m |a_i| n^{i-m}$$

$$\leq n^m \sum_{i=0}^m |a_i|, \text{ for } n \geq 1$$

$$= c n^m, \text{ for } n \geq 1, \text{ where } c = \sum_{i=0}^m |a_i|.$$

So,  $f(n) \leq c n^m \rightarrow f(n) = O(n^m)$

For polynomial functions, simply find the largest degree

# Asymptotic Notation

---

- Lower bound complexity

Def) Omega:  $f(n) = \Omega(g(n))$  iff there exist positive constant  $c$  and  $n_0$  s.t.  $f(n) \geq cg(n)$  for all  $n, n \geq n_0$

Examples

- $3n+2 = \Omega(n)$ 
    - $3n+2 \geq 3n$  for  $n \geq 1$
  - $10n^2+4n+2 = \Omega(n^2)$ 
    - $10n^2+4n+2 \geq n^2$  for  $n \geq 1$
  - $6 \cdot 2^n + n^2 = \Omega(2^n)$
  - $6 \cdot 2^n + n^2 = \Omega(n)$
  - $6 \cdot 2^n + n^2 = \Omega(1)$
- 
- Theorem) If  $f(n) = a_m n^m + \dots + a_1 n + a_0$  and  $a_m > 0$ , then  $f(n) = \Omega(n^m)$

# Asymptotic Notation

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- Lower and upper bound complexity

Def) Theta:  $f(n) = \Theta(g(n))$  iff there exist positive constant  $c_1$ ,  $c_2$  and  $n_0$  s.t.  $c_1g(n) \leq f(n) \leq c_2g(n)$  for all  $n$ ,  $n \geq n_0$

Examples

- $3n+2 = \Theta(n)$ 
    - $3n+2 \geq 3n$  for  $n \geq 2$  and  $3n+2 \leq 4n$  for all  $n \geq 2$
  - $10n^2+4n+2 = \Theta(n^2)$
  - $6 \cdot 2^n + n^2 = \Theta(2^n)$
  - $6 \cdot 2^n + n^2 \neq \Theta(n)$
  - $6 \cdot 2^n + n^2 \neq \Theta(1)$
- 
- Theorem) If  $f(n) = a_m n^m + \dots + a_1 n + a_0$  and  $a_m > 0$ , then  $f(n) = \Theta(n^m)$

# Asymptotic Notation

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- Example: Complexity of matrix addition

void add(int a[][MAX_SIZE]..)	0
{	0
int i, j;	0
for(i=0; i<rows; i++)	$\Theta(\text{rows})$
for(j=0; j<cols; j++)	$\Theta(\text{rows} * \text{cols})$
c[i][j] = a[i][j] + b[i][j]	$\Theta(\text{rows} * \text{cols})$
}	0
Total	$\Theta(\text{rows} * \text{cols})$

- Example: How about binsearch?  $\log(n)$

# Practical Complexities

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- Time complexity of a program:
  - A function of instance characteristics, e.g.,  $f(n)$
- Time complexity is useful in ...
  - Determining how the time requirements vary as the instance characteristics change
  - Comparing two programs P and Q that perform the same task
    - If P has complexity  $\Theta(n)$  and Q is of complexity  $\Theta(n^2)$ , we can assert that P is faster than Q for “sufficiently large”  $n$

# Practical Complexities

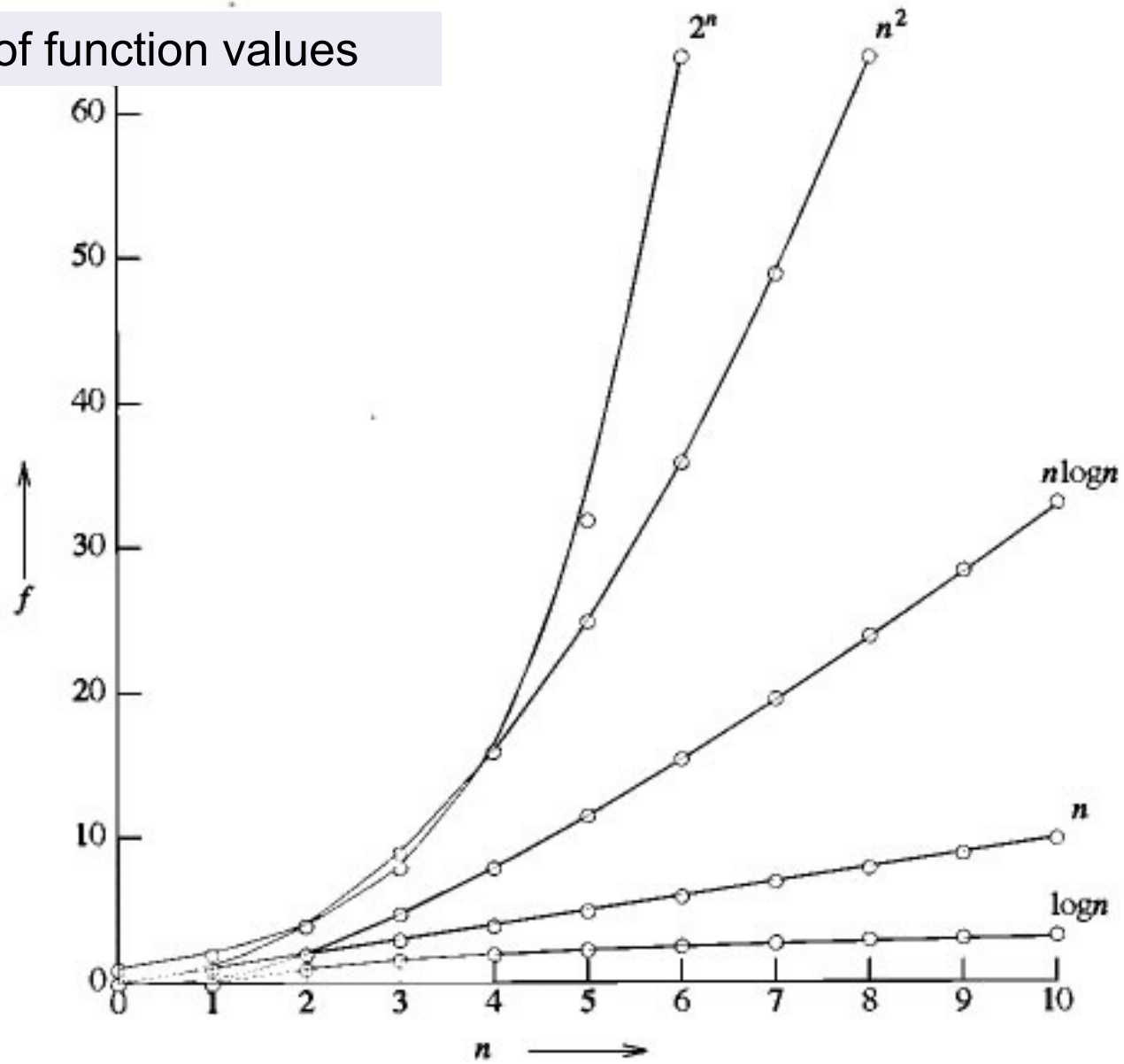
- Growth of function values

		instance characteristic n					
time	name	1	2	4	8	16	32
1	constant	1	1	1	1	1	1
log n	logarithmic	0	1	2	3	4	5
n	linear	1	2	4	8	16	32
n log n	log linear	0	2	8	24	64	160
n <sup>2</sup>	quadratic	1	4	16	64	256	1024
n <sup>3</sup>	cubic	1	8	64	512	4096	32768
2 <sup>n</sup>	exponential	2	4	16	256	65536	4294967296
n!	factorial	1	2	24	40320	20922789888000	26313*10 <sup>33</sup>



# Practical Complexities

Growth of function values



# Practical Complexities

- Time needed by a 1 billion instructions per second (GIPS) computer to execute a program of complexity  $f(n)$

Time for $f(n)$ instructions on a $10^9$ instr/sec computer							
$n$	$f(n)=n$	$f(n)=n \log n$	$f(n)=n^2$	$f(n)=n^3$	$f(n)=n^4$	$f(n)=n^{10}$	$f(n)=2^n$
10	.01 $\mu$ s	.03 $\mu$ s	.1 $\mu$ s	1 $\mu$ s	10 $\mu$ s	10sec	1 $\mu$ s
20	.02 $\mu$ s	.09 $\mu$ s	.4 $\mu$ s	8 $\mu$ s	160 $\mu$ s	2.84hr	1ms
30	.03 $\mu$ s	.15 $\mu$ s	.9 $\mu$ s	27 $\mu$ s	810 $\mu$ s	6.83d	1sec
40	.04 $\mu$ s	.21 $\mu$ s	1.6 $\mu$ s	64 $\mu$ s	2.56ms	121.36d	18.3min
50	.05 $\mu$ s	.28 $\mu$ s	2.5 $\mu$ s	125 $\mu$ s	6.25ms	3.1yr	13d
100	.10 $\mu$ s	.66 $\mu$ s	10 $\mu$ s	1ms	100ms	3171yr	4*10 <sup>13</sup> yr
1,000	1.00 $\mu$ s	9.96 $\mu$ s	1ms	1sec	16.67min	3.17*10 <sup>13</sup> yr	32*10 <sup>283</sup> yr
10,000	10.00 $\mu$ s	130.03 $\mu$ s	100ms	16.67min	115.7d	3.17*10 <sup>23</sup> yr	
100,000	100.00 $\mu$ s	1.66ms	10sec	11.57d	3171yr	3.17*10 <sup>33</sup> yr	
1,000,000	1.00ms	19.92ms	16.67min	31.71yr	3.17*10 <sup>7</sup> yr	3.17*10 <sup>43</sup> yr	

Note that only programs of small complexity (such as  $n$ ,  $n^2$ ,  $n^3$ ) are feasible (for reasonably large  $n$ , say  $n > 100$ ).

# Performance Measurement

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- Performance measurement: measuring execution time on an actual machine
  - Measuring time using functions in C standard library (declared in time.h)

`#include <time.h>`

`clock()`: elapsed time since the program began  
`time()`: elapsed time since Jan. 1, 1970

	Method 1	Method 2
Start timing	<code>start = clock();</code>	<code>start = time(NULL);</code>
Stop timing	<code>stop = clock();</code>	<code>stop = time(NULL);</code>
Type returned	<code>clock_t</code>	<code>time_t</code>
Result in seconds	<code>duration = ((double)(stop - start))/CLOCKS_PER_SEC;</code>	<code>duration = (double)difftime(stop, start)</code>
Remark	Internal processor time	Measured in second

**\* note: exact syntax of functions and constant varies with systems**

# Performance Measurement

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- Example:

```
#include <time.h>
int main()
{
    clock_t start = 0, stop = 0;
    double duration = 0;
    start = clock();

    /// ...
    // routines to measure execution time
    /// ...

    stop = clock();
    duration = ((double) (stop - start) / CLOCKS_PER_SEC);
    return;
}
```

# Performance Measurement

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- Generating test data
  - Usually it is very difficult to generate worst-case data
  - Alternative way:
    - generate suitably large number of random test data
    - estimate the worst-case and the average-case.

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questions or comments?

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