

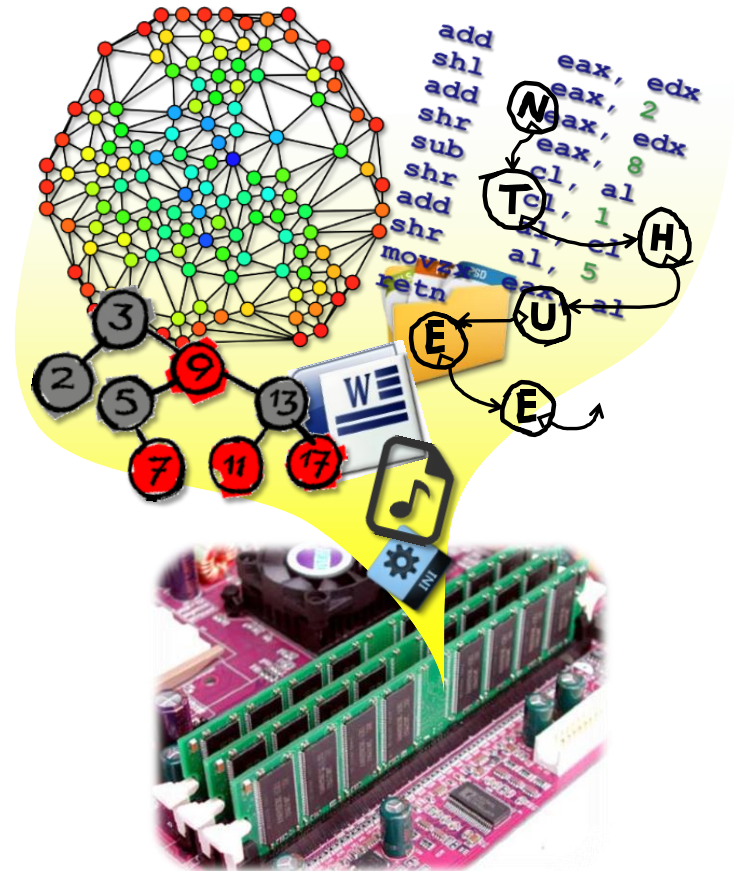
# Data Structures

## CH1 Basic Concepts

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NTHU EE

Spring 2017





# Outline

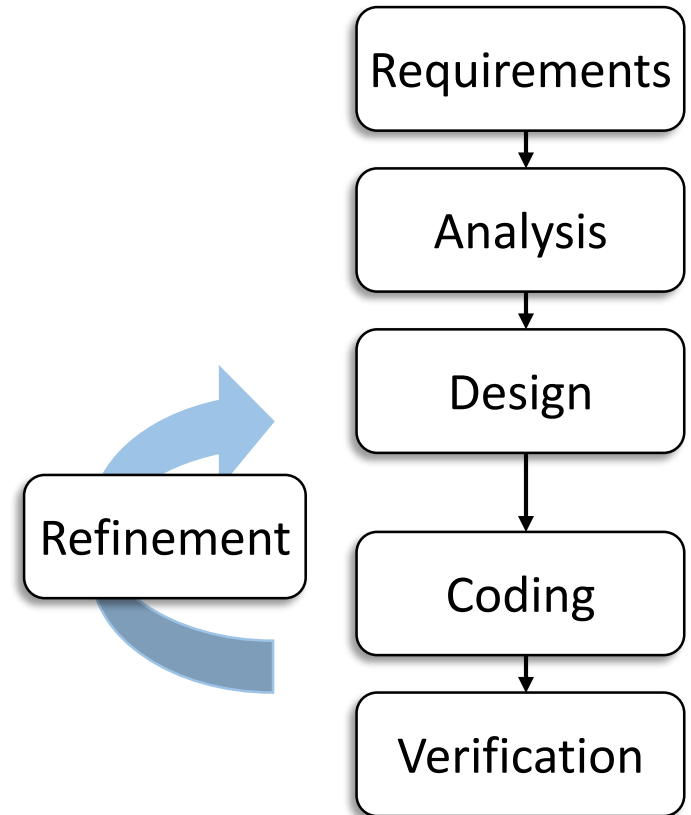
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- **1.1 Overview: System Life Cycle**
- 1.2 Object-Oriented Design
- 1.3 Data Abstraction and Encapsulation
- (1.4 Basics of C++)
- 1.5 Algorithm Specification
- (1.6 Standard Template Library)
- 1.7 Performance Analysis and Measurement



# System Life Cycle

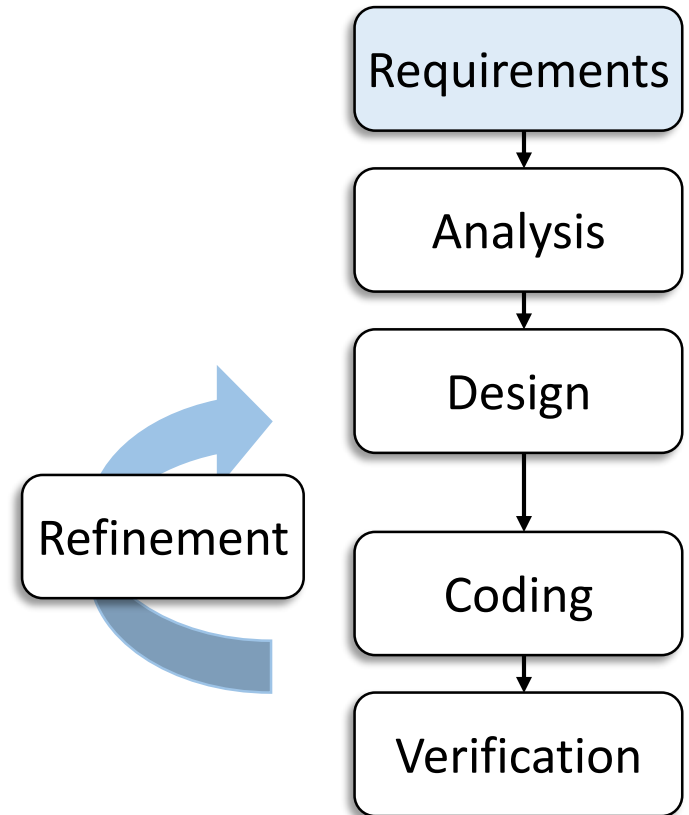
- Five phases
  1. Requirements
  2. Analysis
  3. Design
  4. Refinement and coding
  5. Verification





# Requirements

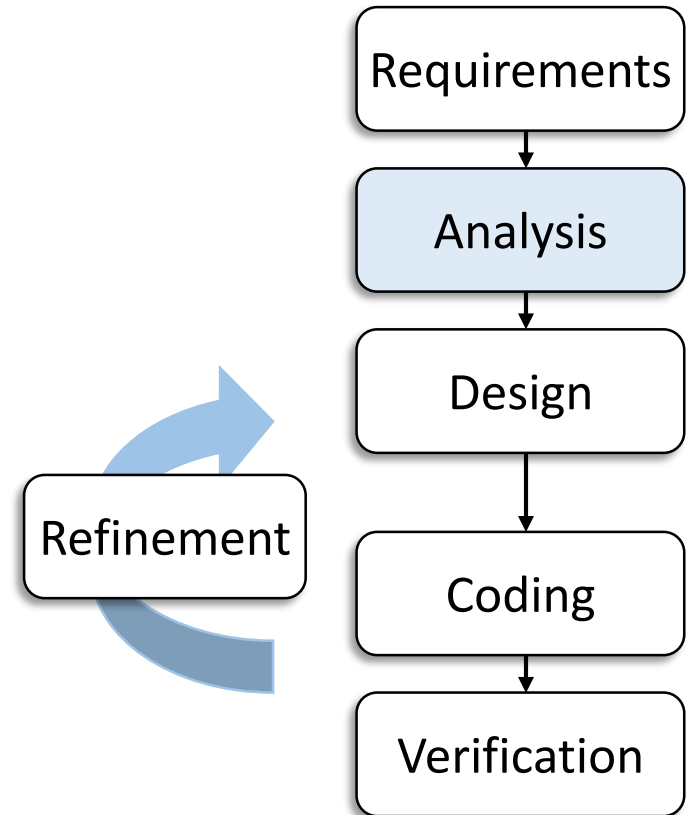
- Clarify **problem specifications**
  - **Input**
    - What are given
  - **Output**
    - What must be produced
- Initially vague → more precise





# Analysis

- Break the problem down
  - Into manageable pieces
  - Also known as **divide and conquer**
- Two approaches
  1. **Bottom-up** (not good)
  2. **Top-down** (better)





# Bottom-up Analysis

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- Issues

- Too early emphasis on implementing fine points
- Lack of **prior planning** and a **big picture**

- Risks and difficulties

- Resulting system can have many loosely connected and error-ridden segments ☹️
- Unpractical for tackling large-scale, complex problem



# Top-down Analysis

---

- Strategies
  - Start from a **high-level plan**
    - Breaking a problem down into manageable pieces
  - Subsequently refining the plan
    - Gradually taking into account low-level details
- Advantages
  - Necessary for tackling large-scale, complex problem

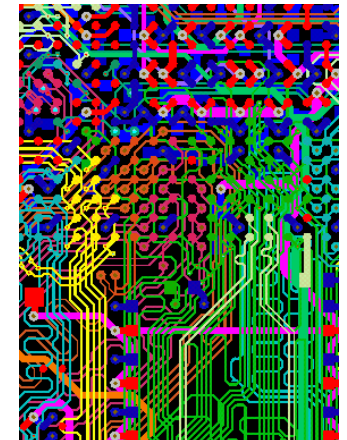
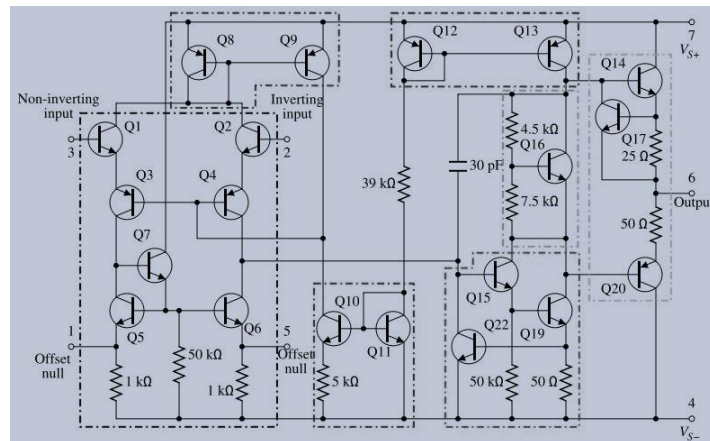
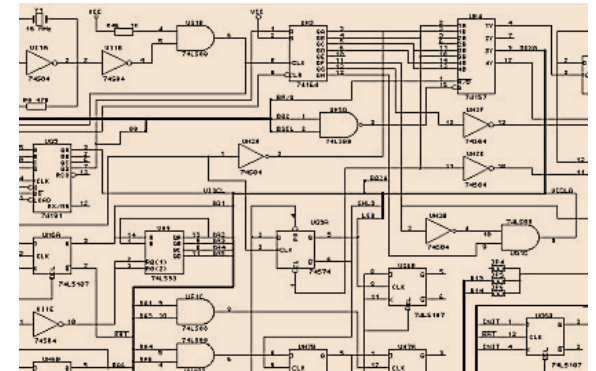
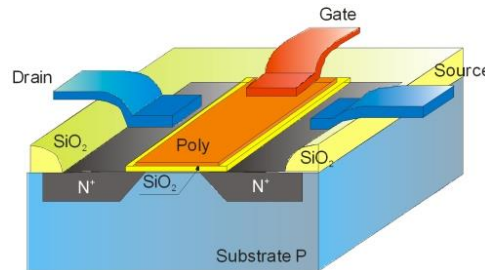
# Risks of Bottom-Up





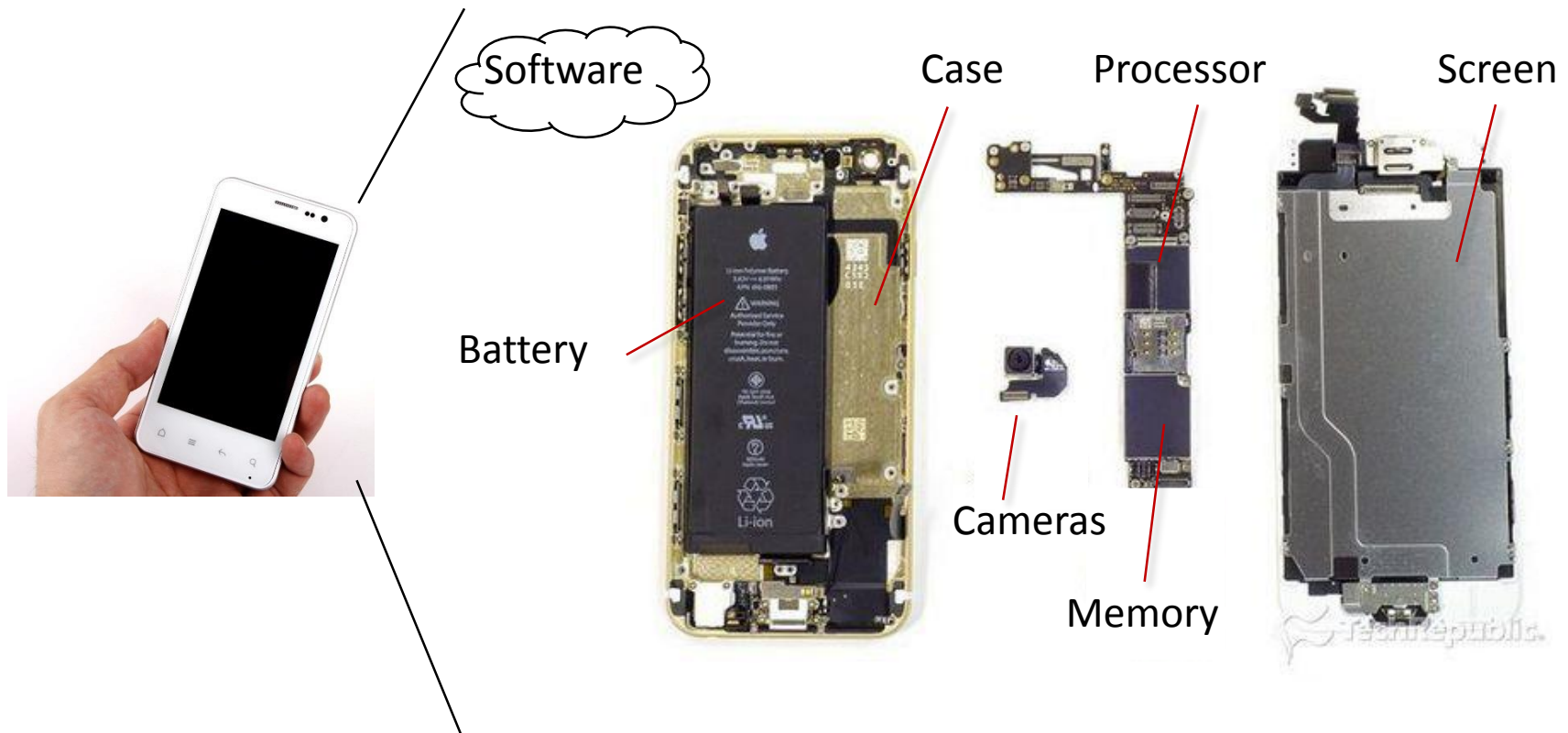
# Difficulties of Bottom-Up

- Please imagine analyzing a smartphone **bottom-up**
  - Things become complicated



# Benefits of Top-Down

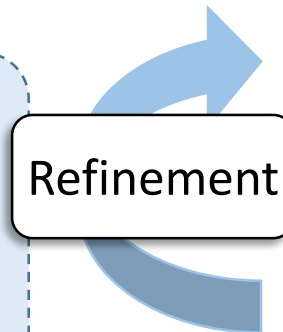
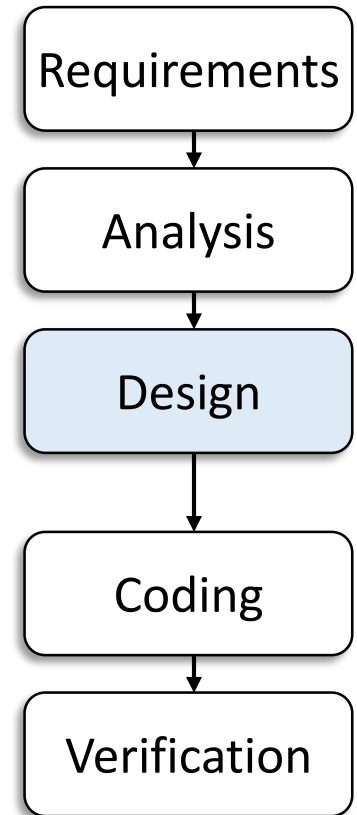
- Now let's alternatively analyze a smartphone **top-down**





# Design

- Identify
  - Data **objects**
  - **Operations** performed on the data types
  - ~~Implementation~~ (Not decided in this phase)
- Produce **implementation-independent** results
  - Abstract data types
  - Algorithm specifications



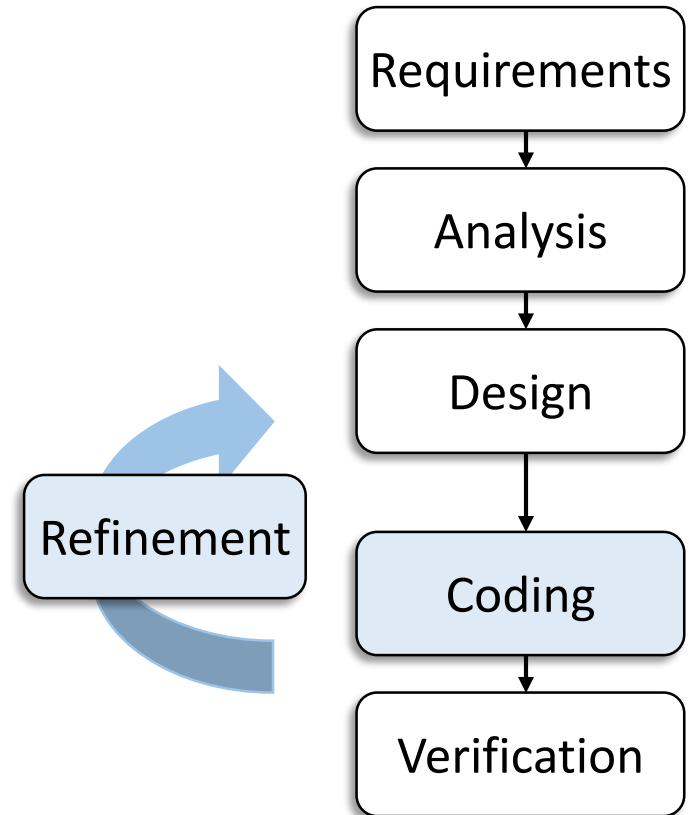
## Scheduling system for NTHU

- **Data objects**
  - Students
    - Name, ID, major, and phone #
  - Courses
  - Professors
- **Operations**
  - Inserting, removing, and searching



# Coding and Refinement

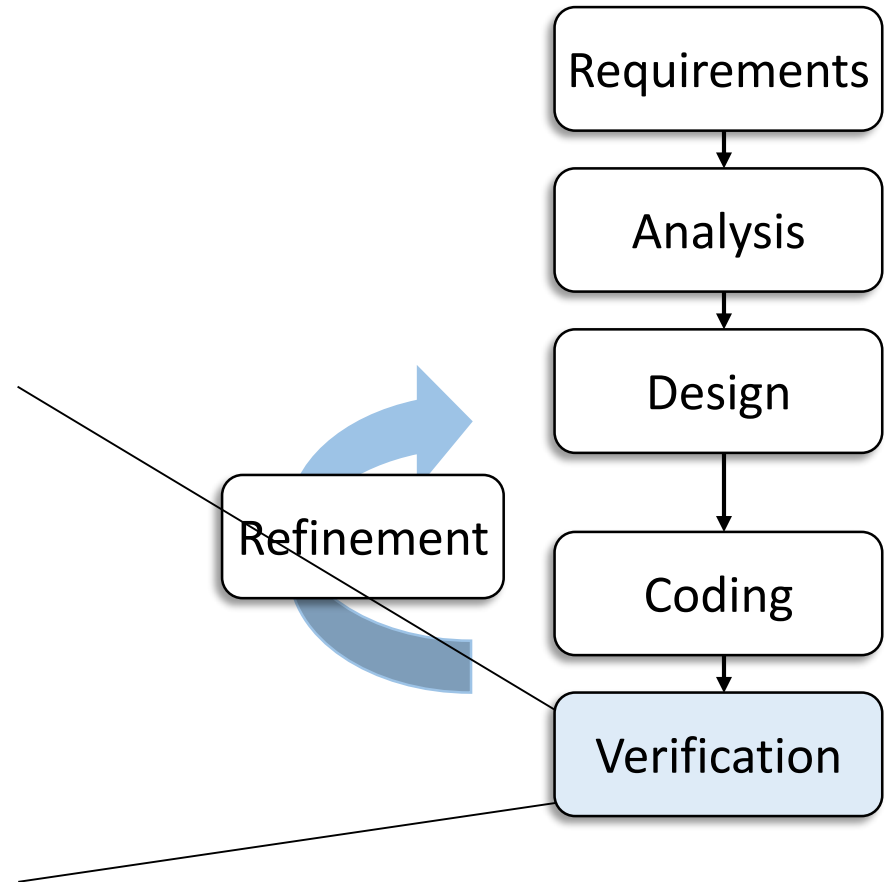
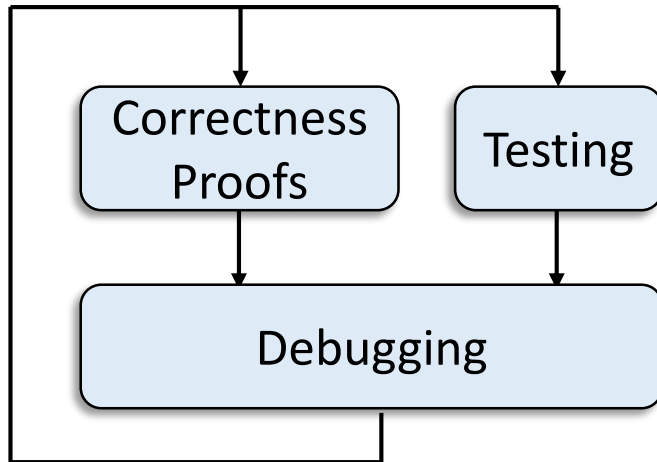
- Decide implementation
  - Representations for objects
  - Algorithms for operations
- Algorithm and object representations affect the **efficiency** of each other
  - Design the algorithms that are independent of data objects first
- Good design can absorb changes found in this stage easily





# Verification

- Three techniques
  1. Correctness proofs
  2. Testing
  3. Debugging

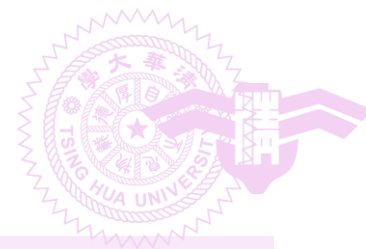




# Verification (Cont'd)

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- Correctness proofs
  - Formal method
  - Typically required for individual algorithm
  - Not easily achievable for the whole program



# Verification (Cont'd)

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- **Testing**

- Run a program against possible **inputs**
  - Check correctness
  - Check performance (e.g., execution time)
- **Coverage** – a metric for assessing the completeness of testing
  - Testing inputs should be developed to **cover as many percentages of codes** as possible
    - E.g., all the cases within a switch statement should at least be touched

- **Debugging**

- Removal of errors found
- **Well-documented** and **well-structured** program eases debugging



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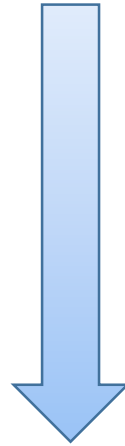




# Programming Paradigms

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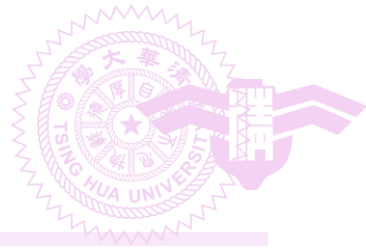
- Non-structured
- Structured
- Object-oriented



More disciplines are imposed on programmers

# Non-Structured Programming

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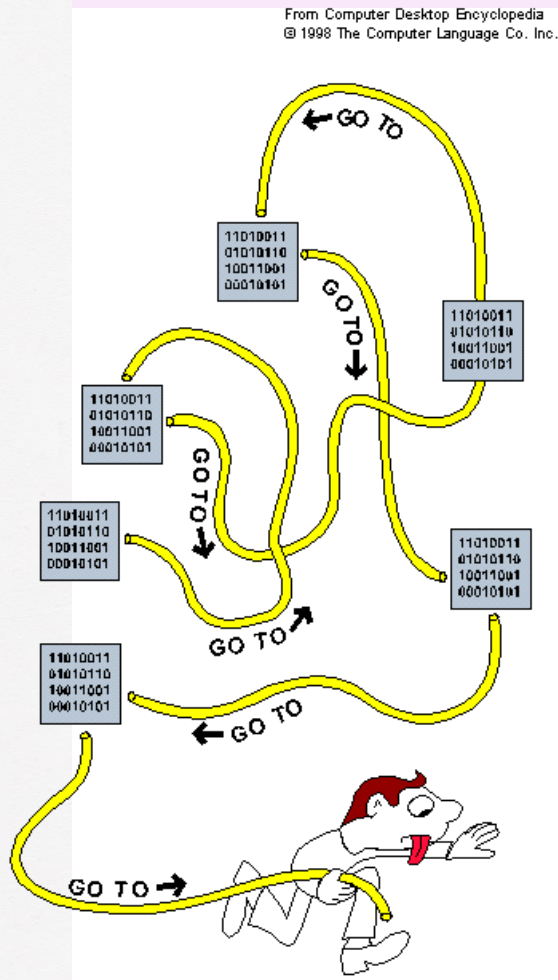


- Characteristics
  - Sequentially ordered commands
  - Lines are numbered or labeled
  - **Unrestricted** jump/branch to any line
- **Pros**
  - Extremely skillful programmers can find **tricky methods** to produce high performance or compact code
- **Cons**
  - Encourage **spaghetti** codes
  - Poor **maintainability**
  - Difficult in building large programs (poor **scalability**)



# Spaghetti Code

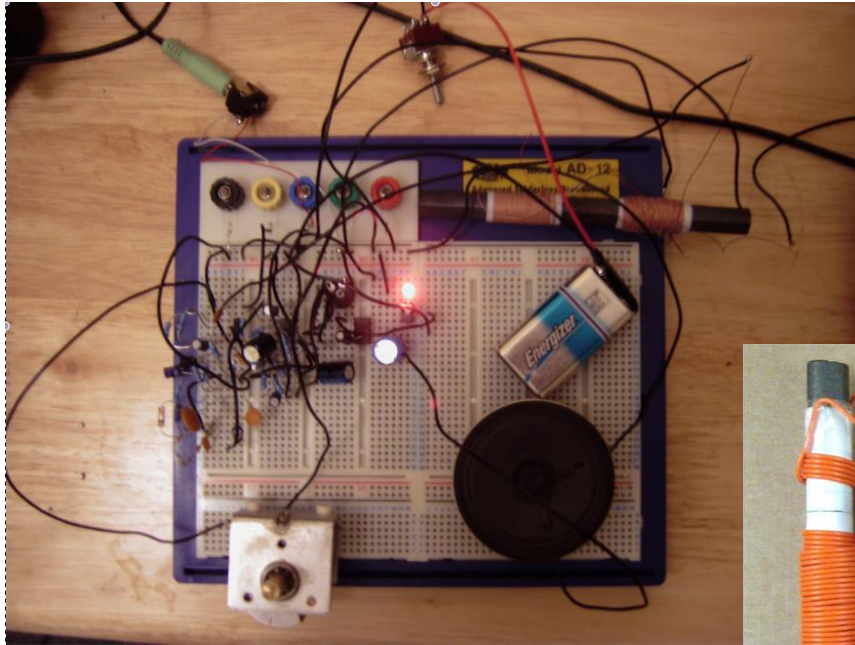
```
PROGRAM PI
DIMENSION TERM(100)
N=1
3  TERM(N)=((-1)**(N+1))*(4./(2.*N-1.))
  N=N+1
  IF (N-101) 3,6,6
6  N=1
7  SUM98 = SUM98+TERM(N)
  WRITE(*,28) N, TERM(N)
  N=N+1
  IF (N-99) 7, 11, 11
11 SUM99=SUM98+TERM(N)
  SUM100=SUM99+TERM(N+1)
  IF (SUM98-3.141592) 14,23,23
14 IF (SUM99-3.141592) 23,23,15
15 IF (SUM100-3.141592) 16,23,23
16 AV89=(SUM98+SUM99)/2.
  AV90=(SUM99+SUM100)/2.
  COMANS=(AV89+AV90)/2.
  IF (COMANS-3.1415920) 21,19,19
19 IF (COMANS-3.1415930) 20,21,21
20 WRITE(*,26)
  GO TO 22
21 WRITE(*,27) COMANS
22 STOP
23 WRITE(*,25)
  GO TO 22
25 FORMAT('ERROR IN MAGNITUDE OF SUM')
26 FORMAT('PROBLEM SOLVED')
27 FORMAT('PROBLEM UNSOLVED', F14.6)
28 FORMAT(I3, F14.6)
END
```



**FORTRAN's three-way arithmetic IF**  
Jump to one of three locations in the program depending on the whether expression was negative, zero, or positive.

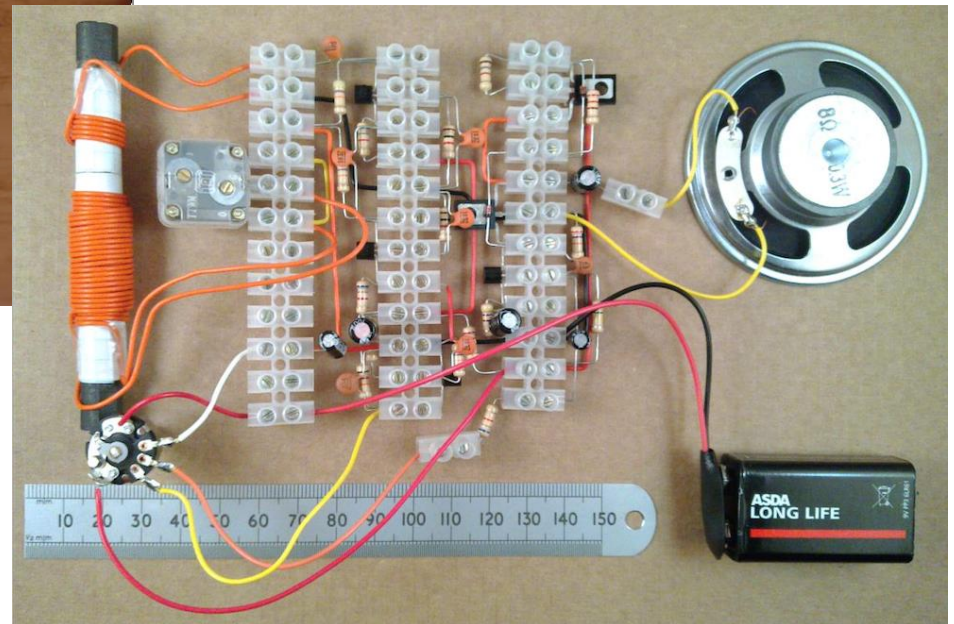


# Spaghetti Circuit



← Spaghetti circuit

↓ Clean circuit

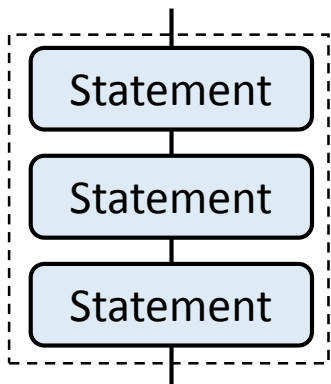


What do you think the possible function of these circuits is?

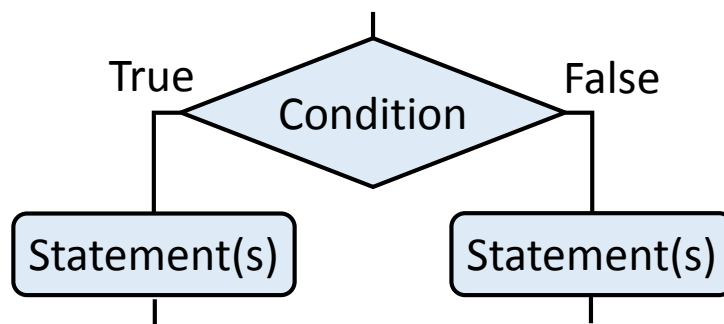


# Structured Programming

- Basic structures

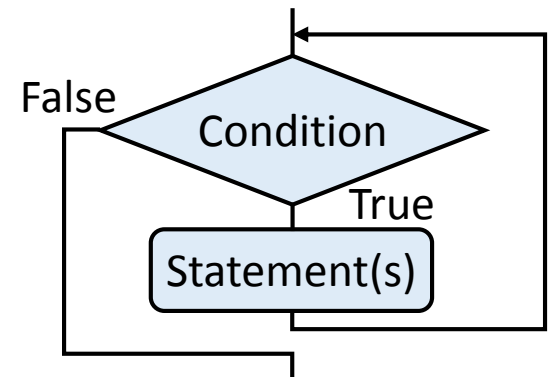


**Sequence**



**Selection (or choice)**

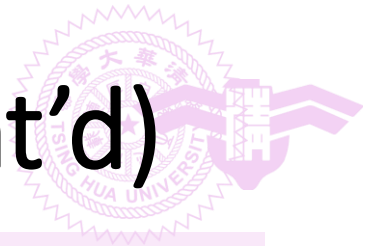
If(condition) {...} else {...}



**Repetition (or looping)**

While(condition) {...}

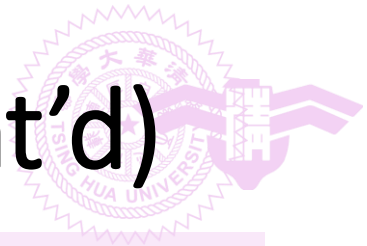
- All programs can be equivalently transformed to that use only the above three structures without using *goto*



# Structured Programming (Cont'd)

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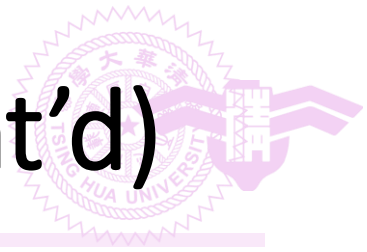
- Pros
  - Easy to understand
  - Easy to maintain
  - Easy to analyze
- Pure structured languages strictly dis-allow
  - goto
  - break
  - continue



# Structured Programming (Cont'd)

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- Compared with non-structured programming
  - Structured programming restricts programmers' freedom
  - Structured programming prevent spaghetti codes
  - Structured programming does not change programmability
    - What problem non-structured programming can solve can also be done using structured programming (and vice versa)



# Structured Programming (Cont'd)

- C and C++ are structured languages but NOT pure ones
  - *goto*, *break*, *continue* statements are allowed
- *goto* statement is notorious but not always bad
  - See the example on the right

```
for(x=0; x<1000; x++){  
  for(y=0; y<1000; y++){  
    for(z=0; z<1000; z++){  
      if( g(x, y, z) > 0 ){  
        cout << x << ", "  
              << y << ", " << z;  
        goto END;  
      }  
    }  
  }  
}  
END:
```

Code snippet for searching an integer solution of  $g(x, y, z) > 0$  in a brute force way. In this example, it is convenient to use *goto* to leave the nested loops.



# Object-Oriented Programming

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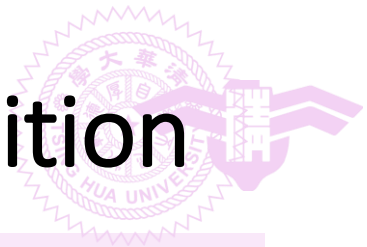
- Philosophy of **divide-and-conquer** is the same as structured programming
- How a project should be **decomposed** is changed
- **Decomposition** methods
  1. **Algorithmic (functional)** decomposition is used for the structured programming method
  2. **Object-oriented** decomposition is used for the object-oriented programming method



# Algorithmic/Functional Decomposition

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- Used by structured programming
- View software as a **process**
- Decompose software into **modules** that represent **steps of the process**
  - In C, the modules are implemented by **functions**
- Compute-centric perspective
- Data structures are a secondary concern



# Object-Oriented (OO) Decomposition

---

- Used by object-oriented programming
- View software as **a set of well-defined objects**
  - Objects model **entities** in the **application domain**
    - e.g., students, courses, and teachers in a course scheduling system
  - Objects interact with one another
- Algorithmic or functional decomposition is addressed after the system has been decomposed into objects



# OO Decomposition (cont'd)

---

- Pros
  - Encourage the **reuse** of software
  - Software becomes more **flexible** that can evolve as requirements change
  - More intuitive because objects naturally **model** entities in the **application domain**



# Definitions

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- Object
  - Entity that has a local state and performs computations
    - i.e., a combination of **data** and **operations**
- Object-oriented programming
  - Method of implementation in which ...
    - **Objects** are the fundamental building blocks
    - Each object is an instance of some **type** (or **class**)
    - Classes are related to each other by **inheritance** relationships



# Definitions

---

- A language is said to be an **object-oriented** language if
  - It supports **objects**
  - It requires objects to belong to a **class**
  - It support **inheritance**
- A language is said to be merely an **object-based** language if it supports the first two features but does not support **inheritance**



# Evolution of Programming

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- Four generations of higher level languages
  - FORTRAN, etc.
    - Salient feature of evaluating **mathematical expression**
  - C, Pascal, etc.
    - Emphasis on effectively **expressing algorithm**
  - Modula, etc.
    - Introduce of the concept of **abstract data types (ADT)**
  - Smalltalk, Objective C, C++, etc.
    - Emphasis on **inheritance** between ADTs



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# Definition

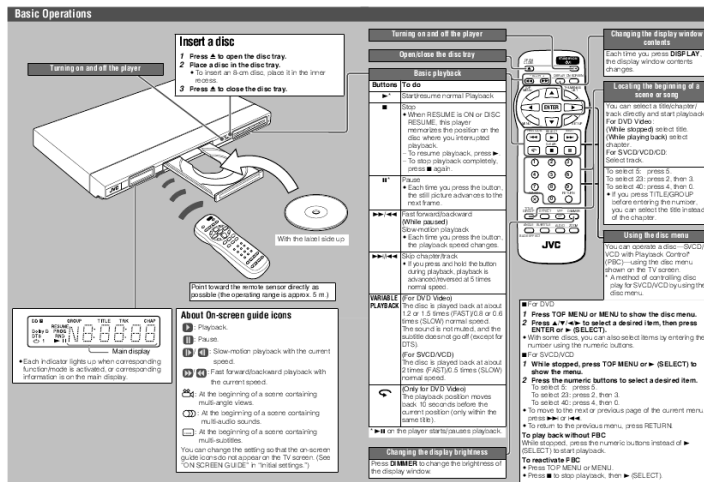
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- Data Encapsulation (or Information Hiding) (封裝)
  - Conceal the implementation details of a data object from the outside world
- Data Abstraction (抽象化)
  - Separation between the **specification** of a data object and its **implementation**

# DVD Player Analogy



- **Encapsulation** — the buttons and remote control
  - The only **interfaces** exposed to users
  - **Hide and protect** internal (vulnerable, dangerous, and proprietary) design from users



- **Abstraction** — the user manual
  - Only specify what the function of each button is
  - How the player achieve the function **is not mentioned nor restricted**



# Definition

- Data Type
  - objects
  - +  
operations on the objects

- Abstract Data Type (ADT)

- Object

Specification

Representation

- Operation

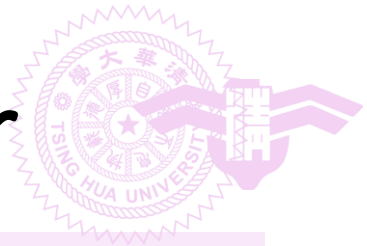
Specification

Implementation



# Data Types in C++

- Predefined (built-in) types
  - Fundamental types
    - *char*
    - *int*
    - *float*
    - *double*
  - Modifiers
    - *short*
    - *long*
    - *signed*
    - *unsigned*
- Derived types
  - Pointer (*\**)
  - Reference (*&*)
- Aggregate types
  - Arrays
  - *struct*
  - *class*
- User-defined types
  - *struct*
  - *class*



# ADT Example: *NaturalNumber*

ADT *NaturalNumber* is

## objects:

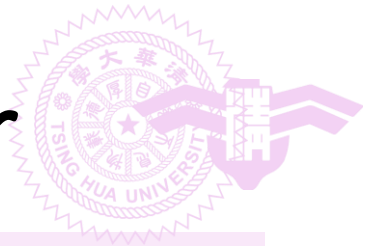
An ordered subrange of the integers starting at zero and ending at MAXINT on the computer.

## functions:

for all  $x, y \in \textit{NaturalNumber}$ ; **true**, **false**  $\in \textit{Boolean}$   
and where +, -, <, ==, = are the usual integer operations

```
Zero (): NaturalNumber ::= 0
IsZero (x): Boolean ::= if (x == 0) IsZero = true
                           else IsZero = false
Add (x, y): NaturalNumber ::= if (x+y <= MAXINT) Add = x + y
                           else Add = MAXINT
Equal (x, y): Boolean ::= if (x == y) Equal = true
                           else Equal = false
Successor (x): NaturalNumber ::= if (x == MAXINT) Successor = x
                           else Successor = x + 1
Subtract (x, y): NaturalNumber ::= if (x < y) Subtract = 0
                           else Subtract = x - y
```

**end** *NaturalNumber*



# ADT Example: *NaturalNumber*

## objects:

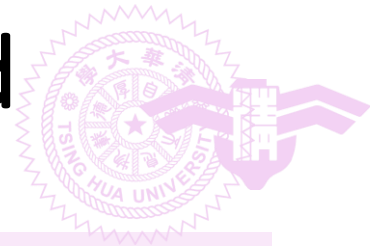
An ordered subrange of the integers starting at zero and ending at MAXINT on the computer.

## functions specification:

| Format                 | Return Type          | Behavior                                                                       |
|------------------------|----------------------|--------------------------------------------------------------------------------|
| <i>Zero</i> ()         | <i>NaturalNumber</i> | 0                                                                              |
| <i>IsZero</i> (x)      | <i>Boolean</i>       | <b>if</b> (x == 0)<br><b>return true</b><br><b>else</b><br><b>return false</b> |
| <i>Add</i> (x, y)      | <i>NaturalNumber</i> | <b>if</b> (x+y <= MAXINT) <b>return</b> x + y<br><b>else return</b> MAXINT     |
| <i>Equal</i> (x, y)    | <i>Boolean</i>       | <b>if</b> (x == y) <b>return true</b><br><b>else return false</b>              |
| <i>Successor</i> (x)   | <i>NaturalNumber</i> | <b>if</b> (x == MAXINT) <b>return</b> x<br><b>else return</b> (x+1)            |
| <i>Subtract</i> (x, y) | <i>NaturalNumber</i> | <b>if</b> (x < y) <b>return</b> 0<br><b>else return</b> (x-y)                  |

# Advantages of Encapsulation and Abstraction

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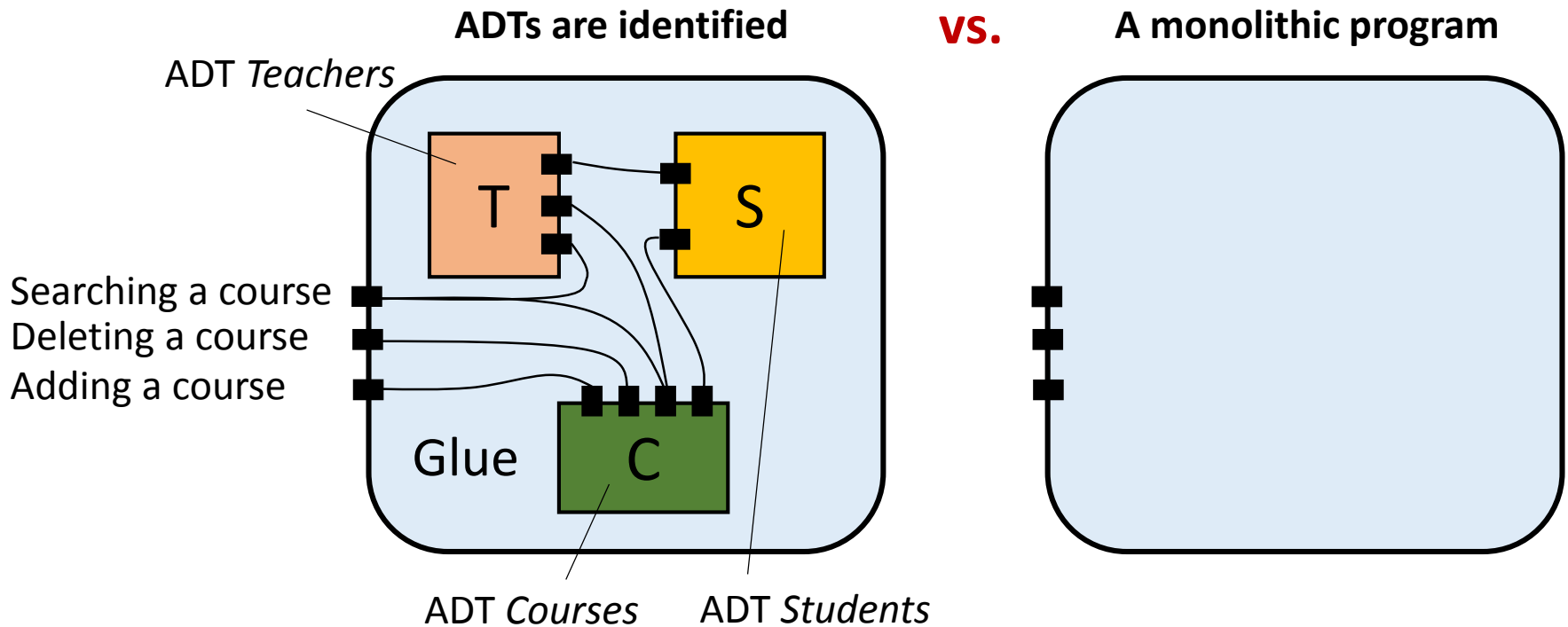


1. Simplify software development
2. Ease testing and debugging
3. Enable reusability
4. Support modifications to the representation of a data type

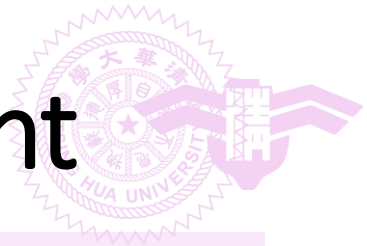


# Comparing Two Scenarios

- Consider developing a course scheduling program for NTHU
  - One can either adopt ADTs or directly dive into coding

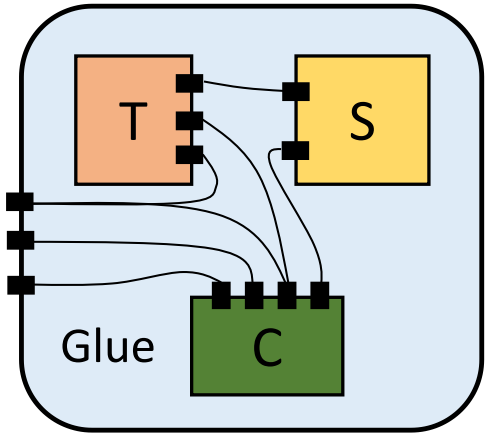






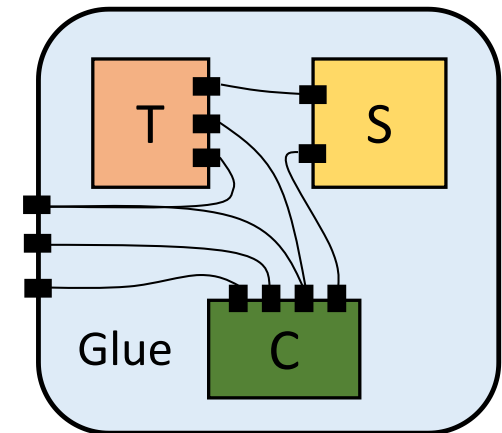
# Simplify Software Development

- With encapsulation and abstraction
  - If we have four programmers
    - They can **parallelly work** on T, S, C, and Glue
    - No one need to know how another one implement their portion of code
    - More concentration and less interference (especially when the project is large)
  - If we have only one programmer
    - Focus on **T, S, C,** and **Glue** one at a time
    - Less things need to be kept in mind



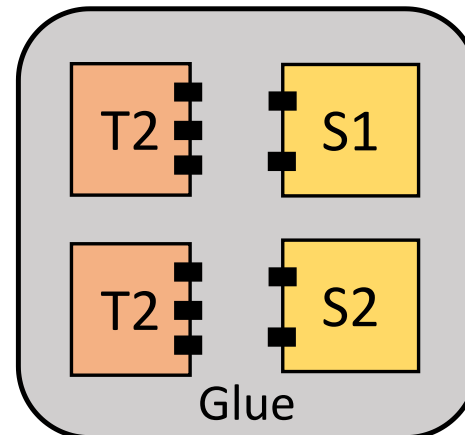
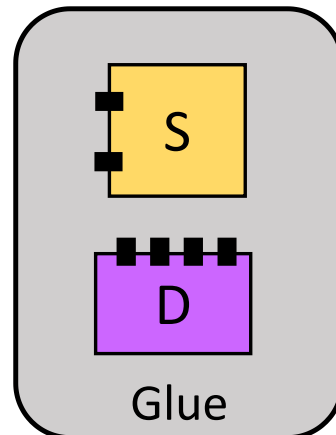
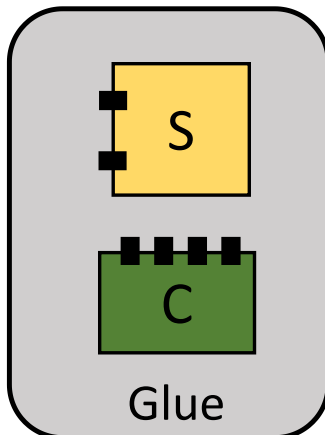
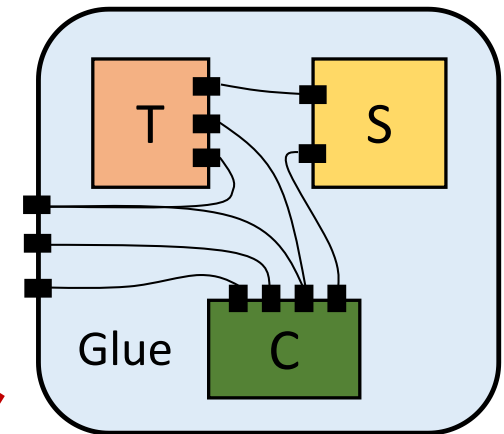
# Testing and Debugging

- With encapsulation and abstraction
  - **T**, **S**, **C**, and **Glue** can be individually tested and debugged
    - Testing efforts are  $T(T) + T(S) + T(C) + T(\text{Glue}) \leq T(T+S+C+\text{Glue})$
  - Assume we are confident that some portions, e.g., **T**, **S**, and **C**, are good, but a bug exists...
    - → The remainder, i.e., **Glue**, has the bug
  - Assume we notice the bug is related to a specific operation on a data type, say **mistakenly deleting a course**...
    - → The bug resides in the corresponding objects and operations



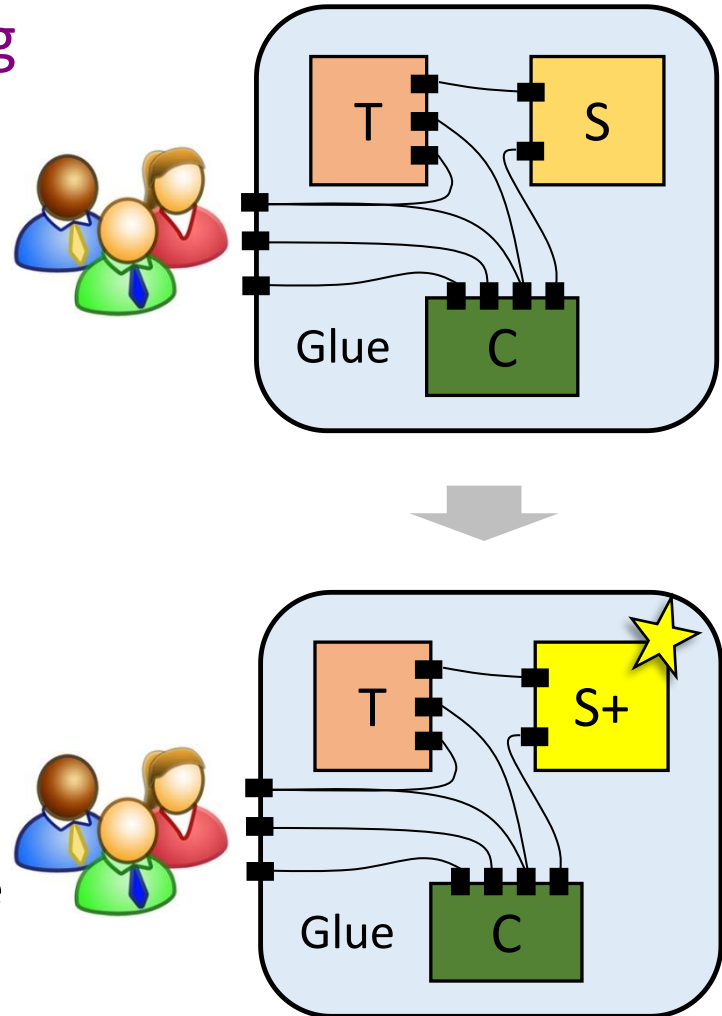
# Reusability

- When we (or other people) develop
  - Textbook ordering program
  - Dorm allocation program
  - NTHU-NCTU tournament program
  - ...



# Modifications

- ADTs lead to **information hiding**
  - Implementation of a data type is invisible to **users** and **the rest of the program**
  - Ease **changing** (e.g., upgrade) a data type without rewriting the entire program or affecting any users
  - Allow us to start from a quick implementation then **progressively refine** the program
  - Even if we need to modify the interface of a data type
    - We can **systematically identify the required modifications** to the other parts





# Overhead of Adopting ADT

---

- **Execution time** overhead
  - Accessing data through interfacing operations is potentially slower than directly accessing them
- **Memory space** overhead
  - Every object maintains a table specifying its operations
- Coding is more tedious
- Therefore, C (not C++) is still widely used for programming the following things
  - Operating systems
  - Performance sensitive systems
  - Resource constrained systems



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# Algorithm

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- Criteria of an algorithm
- Exemplifying algorithms
  - Selection **sort**
  - Binary **search**
- **Recursion**
  - Selection **sort**
  - Binary **search**
  - **Permutation**



# Algorithm (Definition)

---

- A finite set of instructions with the following properties
  - Input
    - Read **zero or more** quantities
  - Output
    - Produce **one or more** quantities
  - Correctness
    - Accomplishes a particular task **for all possible inputs**
  - Definiteness
    - Each instruction is **unambiguous**
  - Effectiveness
    - Each instruction is **basic** enough
  - Finiteness
    - **Terminates** after a finite number steps **for all possible inputs**



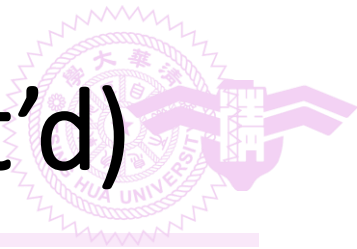


# Algorithms vs. Programs

---

- (From computational theorists' perspective)
- Unlike an algorithm, a program needs not always satisfy “finiteness”
  - Kernel of an **operating system** is an infinite loop
    - Continuously wait until more tasks are entered
    - Continuously dispatch available tasks

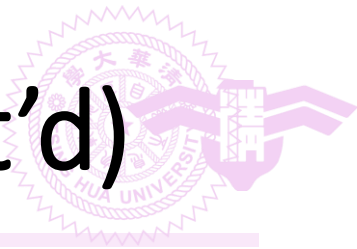
# Algorithms vs. Programs (Cont'd)



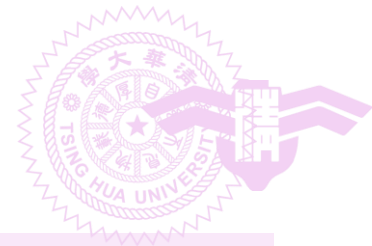
**Which program(s) can always terminate in a finite number of steps?**

1. Testing whether any given number is a **prime**
2. Calculating **10000!** (i.e, factorial(10000))
3. Displaying all **prime numbers**
4. Deciphering an **RSA-encoded** message without knowing the private key
5. Testing whether **an arbitrary program** terminates in a finite number of steps

# Algorithms vs. Programs (Cont'd)



- **Primality** test
  - Even with the brutal force method, it can terminate in a finite number step
- Calculating **factorial(10000)**
  - Factorial(10000) is an astronomical figure (天文數字) though, it involves **a finite number** of digits. So the program can terminate in a finite number step
- Displaying **all prime numbers**
  - Since there are infinitely many primes, this program never terminates



# 10000 Factorial

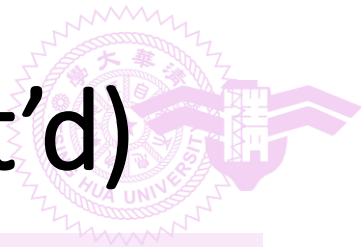
- 10000 factorial is 35,659 digits long. Here it is:

2846259680917054518906413212119868890148051401702799230794179994274411  
3400037644437729907867577847758158840621423175288300423399401535187390  
5242116138271617481982419982759241828925978789812425312059465996259867  
0656016157203603239792632873671705574197596209947972034615369811989709  
2611277500484198845410475544642442136573303076703628825803548967461117  
0973695786036701910715127305872810411586405612811653853259684258259955  
8468814643042558983664931705925171720427659740744613340005419405246230  
3436869154059404066227828248371512038322178644627183822923899638992827  
2218797024593876938030946273322925705554596900278752822425443480211275  
5901916942542902891690721909708369053987374745248337289952180236328274  
1217040268086769210451555840567172555372015852132829034279989818449313  
6106403814893044996215999993596708929801903369984844046654192362584249  
4716317896119204123310826865107135451684554093603300960721034694437798  
2349430780626069422302681885227592057029230843126188497606560742586279  
4488271559568315334405344254466484168945804257094616736131876052349822  
8632645292152942347987060334429073715868849917893258069148316885425195  
6006172372636323974420786924642956012306288720122652952964091508301336  
6309827338063539729015065818225742954758943997651138655412081257886837  
0423920876448476156900126488927159070630640966162803878404448519164379  
0807186112370622133415415065991843875961023926713276546986163657706626

...

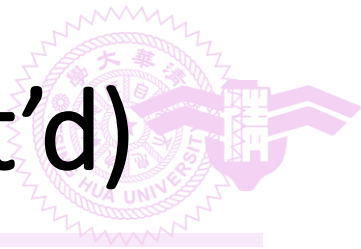
[http://gimbo.org.uk/texts/ten\\_thousand\\_factorial.txt](http://gimbo.org.uk/texts/ten_thousand_factorial.txt)

# Algorithms vs. Programs (Cont'd)



- Breaking RSA
  - This problem corresponds to **factorization** (質因數分解)
    - Factorization is feasible in a finite number of steps
  - RSA is based on the **belief (not proof)** that factoring large integers (particularly that with exactly two huge prime factors) is difficult (i.e., takes unreasonably long time)
    - E.g., thousands of years with a GHz computer
  - Conspiracy theory (陰謀論)
    - Since the proof is unavailable nowadays, some people oppositely believe that some countries have efficient ways to do factorization!!
- Interested students may want to take a **Cryptography** class

# Algorithms vs. Programs (Cont'd)



- Testing whether **an arbitrary program** (with an input) terminates in a finite number of steps
  - Very useful tool to check whether our program contain bugs that lead to infinite looping
  - Discussions on this problem is out of the scope of this course
- Interested students may want to
  - Google “**halting problem**”
  - Take a **Computational Theory** class

# Halting Problem Explanations

---



- Barber paradox (理髮師悖論)
  - A barber shaves all, and only, people who do not shave themselves
  - Who shaves the barber?
- Halting problem paradox
  - Program has difficulty in testing whether another program derived from itself terminates or not



# Halting Problem Explanations

- Suppose someone claims
  - $\text{Terminate}(\text{program}, \text{input}) = \text{true}$  if the  $\text{program}(\text{input})$  terminates
  - $\text{Terminate}(\text{program}, \text{input}) = \text{false}$  if the  $\text{program}(\text{input})$  does not terminate
- You can develop a counterexample program

```
void f(program)
{
    if(Terminate(program, program) == true)
        for(;;); // do not terminate
    else
        return;
}
```

- What is the answer of  $\text{Terminate}(f, f)$ 
  - $\text{Terminate}(f, f) = \text{true} \rightarrow f(f)$  should terminate, but actually it doesn't
  - $\text{Terminate}(f, f) = \text{false} \rightarrow f(f)$  should not terminate, but it actually does





# Describing Algorithms

---

- Many allowable ways
  - Programming languages (e.g., C++)
  - Natural languages
    - Must assure definiteness and effectiveness
  - Pseudocode (e.g., combining C, C++, and English)
    - Less language-dependent
    - More flexibility
  - Graphic representations (i.e., flowcharts)
    - Typically for small and simple algorithms only



# Algorithm Specification

---

- Examples
  - Selection sort
  - Binary search
  - Permutation generator
- Focuses
  - **Inputs** and **outputs**
  - Clear and basic-enough **instructions**
  - Finiteness and correctness **proofs**



# Selection Sort (Algorithm)

- Input
  - A collection of  $n$  integers,  $n \geq 1$
- Output
  - A collection of  $n$  integers
- Instructions (in pseudocode)

```
void SelectionSort(int *a, const int n)
{ //Sort the n integers a[0] to a[n-1] into non-decreasing order.
  for(int i=0; i<n; i++) {
    exam a[i] to a[n-1] and suppose the smallest one is at a[j];
    interchange a[i] and a[j];
  }
}
```



# Selection Sort — C++

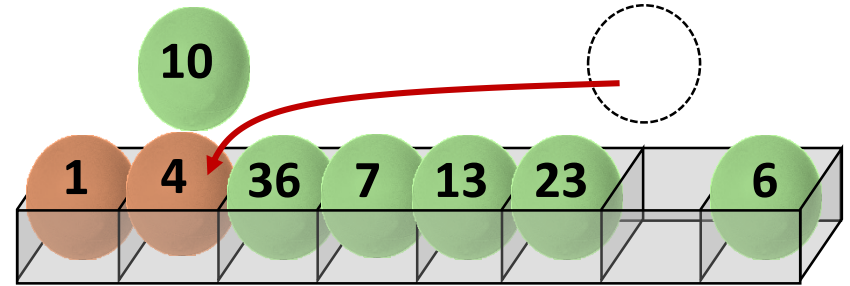
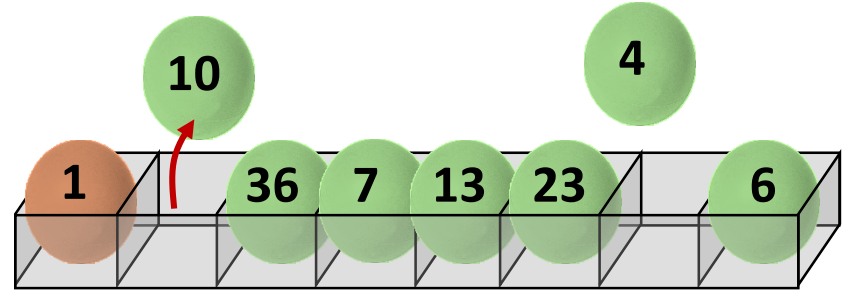
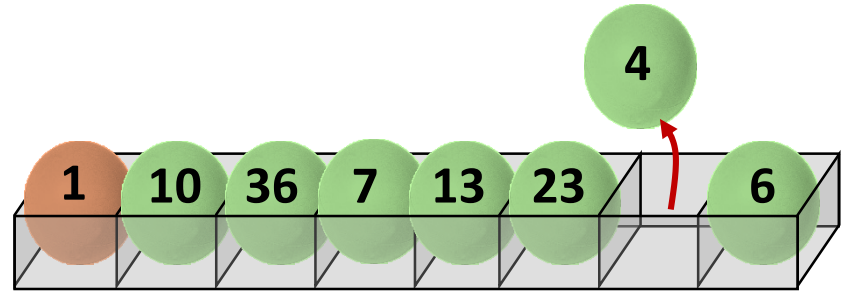
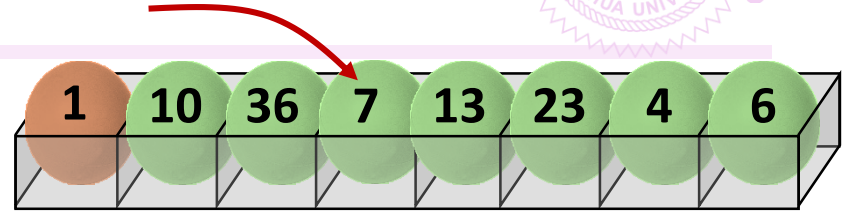
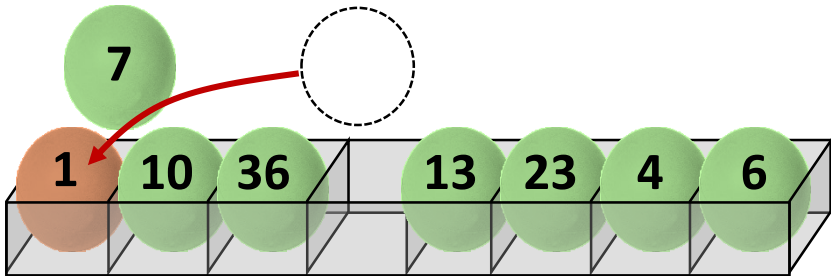
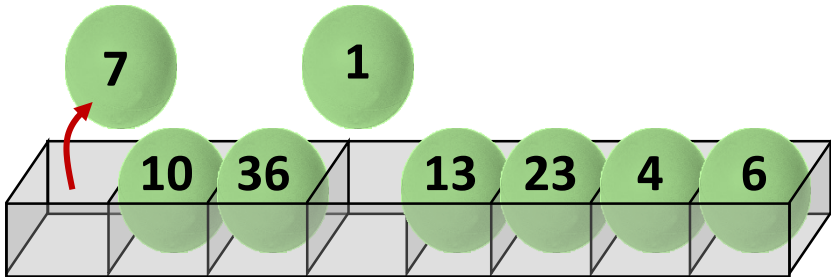
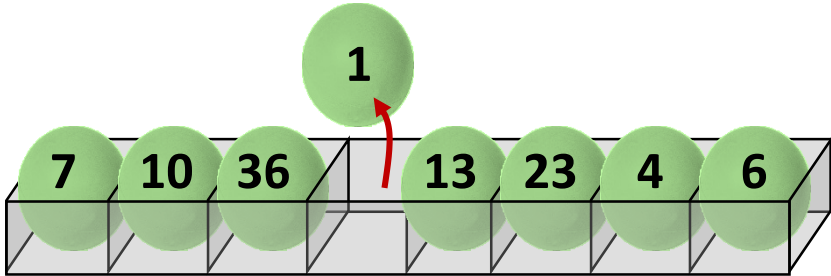
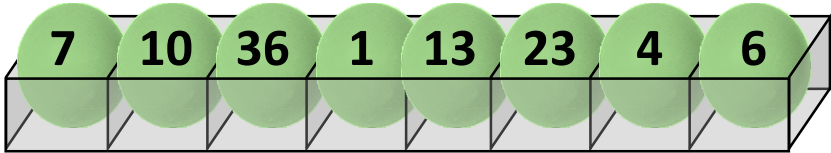
```
void SelectionSort(int *a, const int n)
{ // Sort the n integers a[0] to a[n-1] into
  // non-decreasing order.
  for(int i=0; i<n; i++)
  {
    int j=i;
    //find the smallest integer in a[i] to a[n-1]
    for(int k = i+1; k<n; k++)
      if(a[k] < a[j]) j = k;
    swap(a[i], a[j]);
  }
}
```

```
void swap(int & i, int & j)
{
  int temp = i;
  i = j;
  j = temp;
}
```

Passed by reference



# Illustration





# Selection Sort — Proof

- At the end of loop  $q$  ( $i=q$ )  
 $a[q] \leq a[r]$ ,  $q+1 \leq r \leq n-1$ .
- When  $i$  becomes greater than  $q$ ,  
 $a[0] \dots a[q]$  is unchanged.
- Hence, after the lines are executed  
for  $n-1$  times (i.e.,  $0 \leq i \leq n-2$ ), the  
following  $n-1$  inequalities hold
  - $a[0] \leq a[r]$ ,  $1 \leq r \leq n-1$
  - ...
  - $a[n-3] \leq a[r]$ ,  $n-2 \leq r \leq n-1$
  - $a[n-2] \leq a[r]$ ,  $n-1 \leq r \leq n-1$
- $a[0] \dots a[n-1]$  is unchanged for the  
last iteration (i.e.,  $i = n-1$ )
- Combining these inequalities leads  
to  $a[0] \leq a[1] \leq \dots \leq a[n-2] \leq a[n-1]$

```
void SelectionSort(int a[], const int n)
{ // Sort the n integers into
  // non-decreasing order.
  for(int i=0; i<n; i++)
  {
    int j=i;
    //find the smallest integer in
    for(int k = i+1; k<n; k++)
      if(a[k] < a[j]) j = k;
    swap(a[i], a[j]);
  }
}
```



# Binary Search

---

- Input
  - $n \geq 1$  distinct integers that are already sorted and stored in the array  $a[0] \dots a[n-1]$
  - Integer  $x$
- Output
  - If  $x$  is present in the array, produce  $j$  such that  $x == a[j]$
  - Otherwise, produce  $-1$

# Binary Search — Pseudocode



```
void BinarySearch(int *a, const int x, const int n)
{ // Search the sorted array a[0], ... , a[n-1] for x
  // left and right are set to the two ends of a[]
  while(there're elements between the two ends)
  {
    Let middle be the middle element;
    if(x < a[middle])      set right to middle-1;
    else if(x > a[middle]) set left to middle+1;
    else                    return middle;
  }
  Not found;
}
```





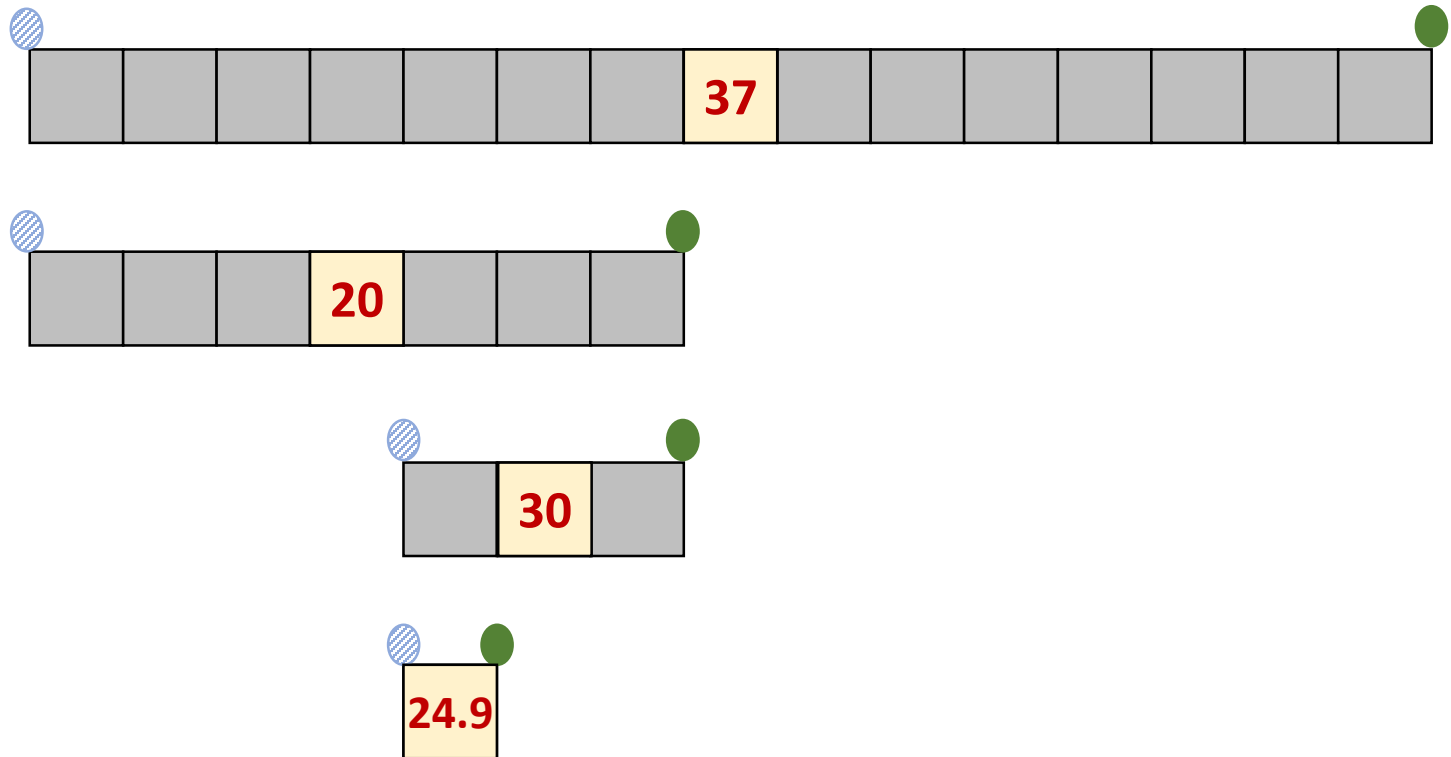
# Binary Search — C++

```
int BinarySearch(int *a, const int x, const int n)
{ //Search the sorted array a[0]...a[n-1] for x
  int left = 0, right = n-1;
  while(left <= right)
  { //there are more elements
    int middle =(left+right)/2;
    if(x < a[middle])      right=middle-1;
    else if(x > a[middle]) left = middle+1;
    else                  return middle;
  } //end of while
  return -1;
}
```



# Binary Search — Illustration

Search a number, 25, in a **sorted** array of boxes



● Left      ● Right



# Recursion

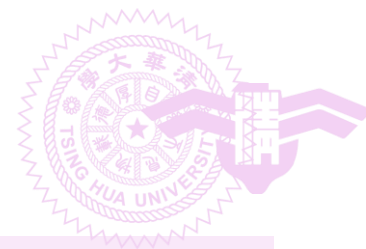
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- **Definition**

- Functions that invoke (呼叫、使用) themselves
  - **Directly** or **Indirectly** through other functions

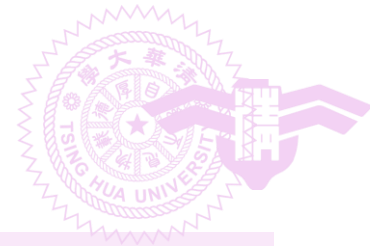
- **Recursion is powerful**

- **Divide and conquer**
- Method of **induction** (歸納法)
- Can simplify the expression of an otherwise complex process



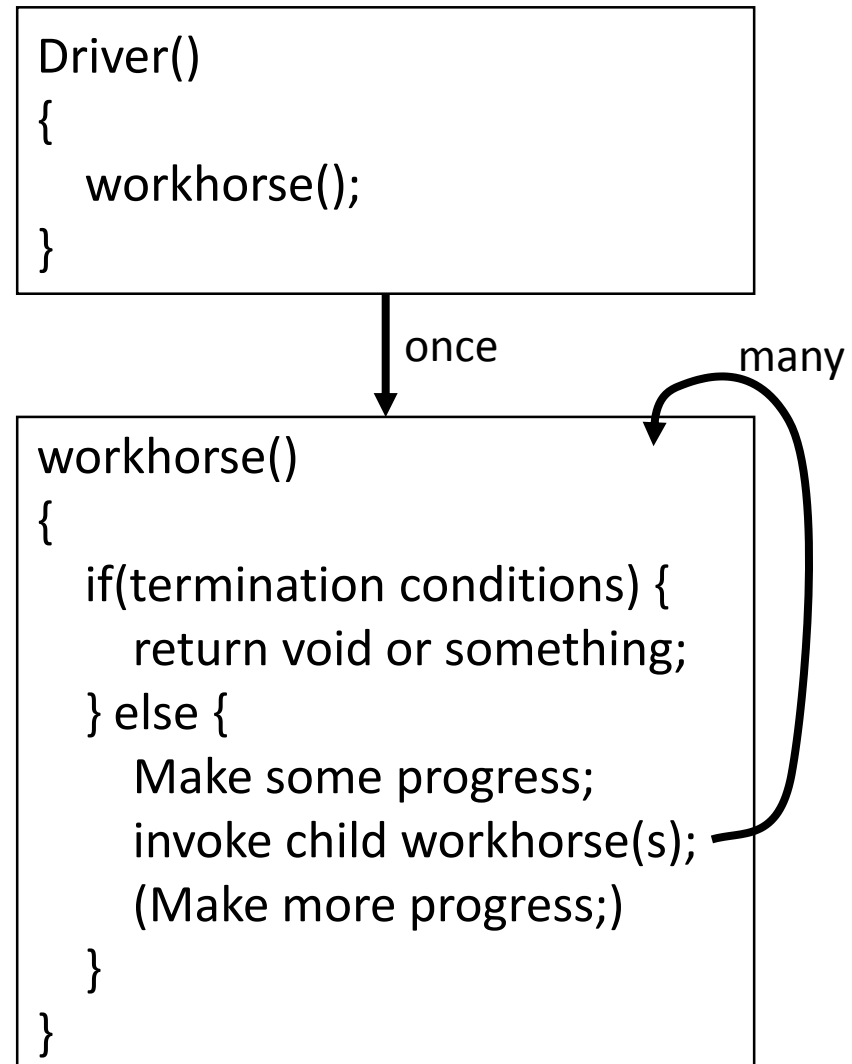
# Recursion (Cont'd)

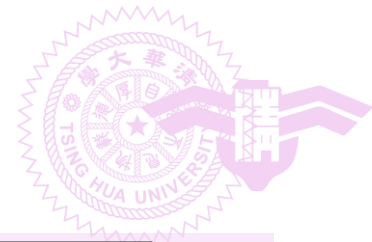
- Recursion is particular useful for
  - Factorial (階乘)
  - Binomial coefficients
  - Binary search
  - Problems that are recursively defined
- Recursion is not limited to the above tasks
  - Recursion can simulate looping  
(Looping can simulate recursion, too)
- Recursion **tends to be** (有這個傾向，但不是絕對) **slower** than looping
  - Because **function invocation** typically incurs longer latency than **loop branches**



# Develop Recursion

- Key components
  - **Driver**
    - Invoke the first workhorse
  - **Workhorse(s)**
    - Self-similar piece of the algorithm
  - **Termination condition(s)**
    - Determine whether no more progress needs be made
    - If a workhorse fails to check termination conditions, the program can never end
  - Make some **progress**
    - If nothing changes before the workhorse is again invoked, the program can never end





# Recursive Selection Sort

```
void SelectionSort(int a[], const int n)
{
    // 1-entry array does not need sorting
    if(n==1) return;

    int j=0;
    /* find the smallest in the received
       array and place it at the first */
    for(int k = 0; k<n; k++)
        if(a[k] < a[j]) j = k;

    swap(a[0], a[j]);

    SelectionSort(a+1, n-1); //recursion
}
```

Termination condition

- This is an exemplifying recursive algorithm derived from a non-recursive one. In this example, recursion is easier to understand but likely performs slower than the non-recursive one.

Create a new workhorse to sort the remaining n-1 elements



# Recursive Binary Search

```
int BinarySearch(int a[], const int x, const int left, const int right)
{
    // no entries to search
    if(left>right) return -1;

    int middle = (left+right)/2;

    if(x<a[middle])        return BinarySearch(a, x, left, middle-1);
    else if(x>a[middle])   return BinarySearch(a, x, middle+1, right);
    else                   return middle;
}
```

Termination condition

Create a new workhorse to search the half that possibly contain the target



# Permutation Generator

---

- Input
  - A set of  $n \geq 1$  elements
- Output
  - Print all  $n!$  possible permutations of this set
- Example
  - Permutations of (a, b, c)
    - (a, b, c), (a, c, b),  
(b, a, c), (b, c, a),  
(c, a, b), (c, b, a)





# Permutation Generator

## — Observation

---

- Permutations of (a, b, c, d) can be constructed by
  - 'a' followed by all permutations of (b, c, d)
  - 'b' followed by all permutations of (a, c, d)
  - 'c' followed by all permutations of (a, b, d)
  - 'd' followed by all permutations of (a, b, c)
- Clue to adopt recursion
  - Solve an **n-element problem** based on the results of an **(n-1)-element problem**

# Recursive Permutation Generator



```
void Permutations(int a[], const int k, const int m)
{
    if(k == m) {
        for(int i=0; i<=m; i++) cout << a[i] << " ";
        cout << endl;
        return;
    }

    for(int i=k; i<=m; i++) {
        swap(a[k], a[i]); //enumerate all possible elements at a[k]
        Permutations(a, k+1, m); // a workhorse to handle the rest
        swap(a[k], a[i]); //restore the element
    }
}
```

Termination condition

Note that this loop can invoke Permutatinos() multiple times

It is a bit hard (but still feasible) to transform this algorithm into an non-recursion version



# Outline

---

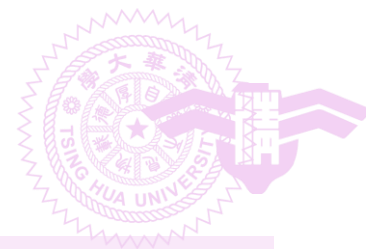
- 1.1 Overview: System Life Cycle
- 1.2 Object-Oriented Design
- 1.3 Data Abstraction and Encapsulation
- (1.4 Basics of C++)
- 1.5 Algorithm Specification
- (1.6 Standard Template Library)
- **1.7 Performance Analysis and Measurement**



# Complexity

---

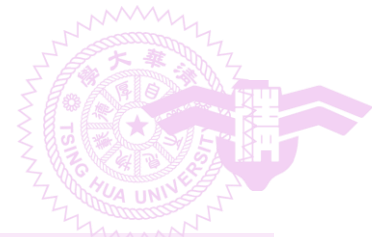
- **Time** complexity
  - Amount of **execution time** a program needs to solve a problem
- **Space** complexity
  - Amount of **memory space** a program needs to solve a problem
- We want to find complexity as a function of **problem size**
  - Problem size  $\equiv$  the total amount of input information



# Space Complexity

---

- Memory space breakdown
  - **Problem size-dependent** part
    - Variables whose size/number depends on problem size
  - Fixed part
    - Space for storing the program
    - Fixed amount of variables during computation
    - Read-only space for Inputs
    - Write-only space for outputs
- We shall concentrate on the **Problem size-dependent** part



# Space Complexity (Cont'd)

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

- **a**, **b**, and **c** are read-only inputs
  - A fixed amount of space is required to do the computation
- Problem size-dependent part is 0

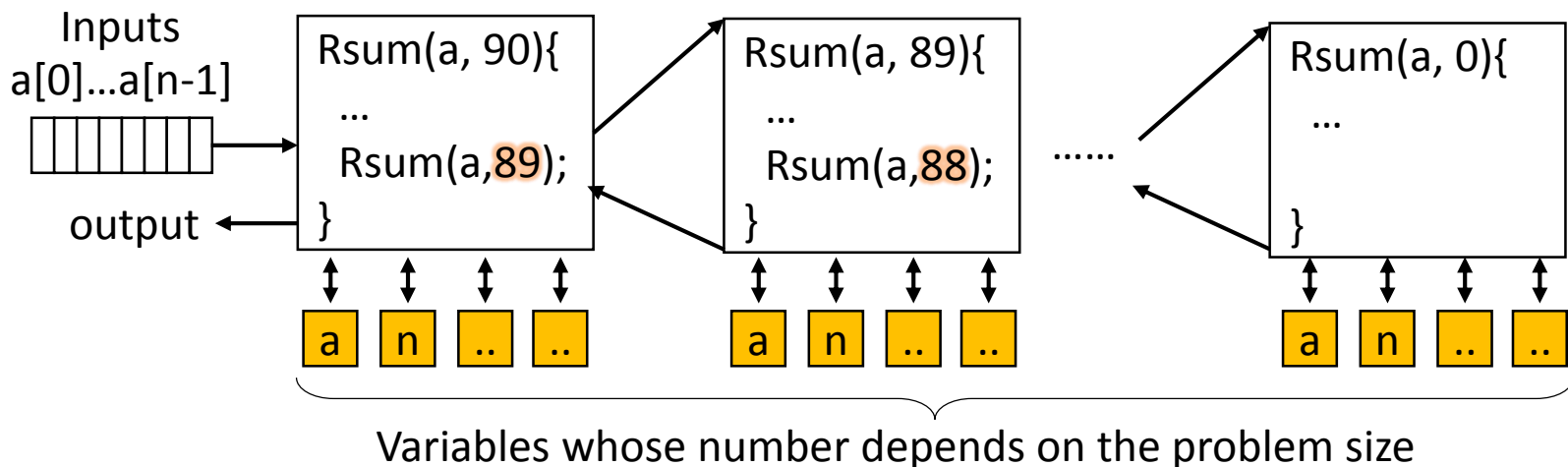
```
float sum (float *a, const int n)
{
    float s = 0;
    for (int i = 0; i < n; i++)
        s += a[i];
    return s;
}
```

- **a[0]...a[n-1]** are read-only inputs
  - Float \* **a**, const int **n**, float **s**, int **i**, etc. consume a fixed amount of space
- Problem size-dependent part is 0

# Space Complexity

```
float Rsum (float *a, const int n)
{
  if (n <= 0)
    return 0;
  else
    return (Rsum(a, n-1) + a[n-1]);
}
```

- $a[0] \dots a[n-1]$  are read-only inputs
- $\text{float } *a$  and  $\text{int } n$  (and other variables local to  $\text{Rsum}()$ ) consume a fixed amount of space for each execution of  $\text{Rsum}$  though,  $\text{Rsum}$  is called  $n+1$  times.
- Variable part is  $c(n+1)$ , where  $c$  is a constant, say  $c=4$





# Time Complexity

---

- Time consumption breakdown
  - Execution time
  - Compile time
- Execution time is important
  - Problem size,  $n$ ,  $\uparrow \Rightarrow$  execution time,  $t_p(n)$ ,  $\uparrow$
- Compile time is less important
  - Independent of problem size,  $n$
  - Only present for the first execution





# Methods to Derive Execution Time

---

- Derive the exact formula
  - $t_p(n) = c_a ADD(n) + c_s SUB(n) + c_m MUL(n) + \dots$ 
    - Almost impossible to obtain such a formula
- Step counts
- Asymptotic notation (漸近表示法) of step counts
- Real system measurement



# Step Count

- Definition of a step
  - A segment of program whose **execution time is independent of problem size**
- Example of a step
  - One addition → a step
  - One multiplication → a step
  - 1000 additions → a step
  - 1000 multiplications → a step
  - $r = a+b+b*c+(a+b-c)/(a+b)+4.0$  → a step
- The following one is NOT a step
  - $n$  additions, where  $n$  is the size of the input array

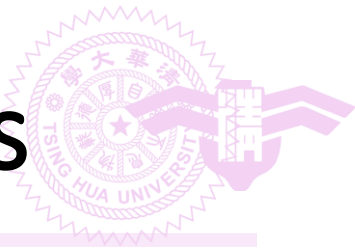


# Zero-Step Program Segments

---

- Comments
  - `// this is binary search`
  - `/* this is`
    - `* selection sort`
    - `*/`
- Declarative statements of variables and functions
  - `int a;`
  - `float b, c, d;`
  - `int max(a, b);`
- Brackets
  - `{`
  - `}`

# Single-Step Program Segments



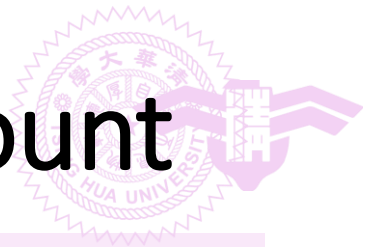
- Assignments and expressions
  - `int a = 10;`
  - `b = 0.1;`
  - `c = a + b * d;`
- Control statements of loops
  - `for(int i=0; i<n; i++)`
  - `while(j<n)`
  - `do ... while(1)`
- Function independent of problem size
  - `a = max(b, c)`
- Conditional statements
  - `if(a > 10)`
- Unconditional branches
  - `goto, break, continue, return`



# Those May Depend on Problem Size

---

- Object/variable construction
  - `int *a = new int[size(input)];`
- Function execution
  - `MatrixAdd(a, b, c);` // adding two matrixes
- Parameter passing
  - Passing an object whose size depends on problem size
- Statements that involve the above events
  - `int a = sum(a, n);`
  - `if(search(a, x, n) == true)`



# Methods of Obtaining Step Count

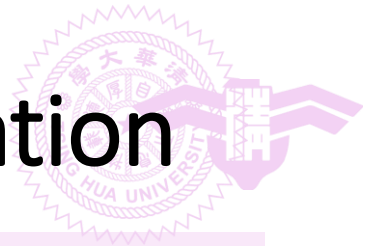
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- Instrumentation
  - Introduce a new global variable *count*
  - Initialize *count* to zero
  - Add statements to increment *count* for each step
  - Report *count*
- Table analysis
  - List the step count of each program segment
  - List the frequency of each program segment
  - Summarize the total step count



# Step Counting — Example 1

```
float sum (float *a, const int n)
{
    float s = 0;
    for (int i = 0; i < n; i++)
        s += a[i];
    return s;
}
```



# Step Counting Using Instrumentation

```
float sum (float *a, const int n)
{
    float s = 0;
    count++; // count is global
    for (int i = 0; i < n; i++) {
        count++; // for loop
        s += a[i];
        count++; // assignment
    }
    count++; // last time of for
    count++; // return
    return s;
}
```

Simplified version

```
void sum (float *a, const int n)
{
    for (int i = 0; i < n; i++) {
        count+=2;
    }
    count+=3;
    return;
}
```





# Step Counting Using a Table

| <code>float sum (float *a, const int n)</code> | <u>s/e</u> | <u>freq.</u>  | <u>subtotal</u> |
|------------------------------------------------|------------|---------------|-----------------|
| <code>{</code>                                 | 0          |               |                 |
| <code>  float s = 0;</code>                    | 1          | 1             | 1               |
| <code>  for (int i = 0; i &lt; n; i++)</code>  | 1          | n+1           | n+1             |
| <code>    s += a[i];</code>                    | 1          | n             | n               |
| <code>  return s;</code>                       | 1          | 1             | 1               |
| <code>}</code>                                 | 0          |               |                 |
|                                                |            | <b>total:</b> | <b>2n+3</b>     |

**s/e**: steps per execution

The frequency of executing the control statement is one time more than that of the loop body.



# Step Counting — Example 2

```
float Rsum (float *a, const int n)
{
    if (n <= 0)
        return 0;
    else
        return (Rsum(a, n-1) + a[n-1]);
}
```

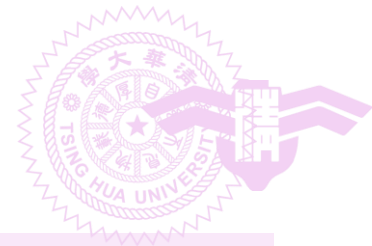
- Recursion



# Step Counting — Instrumentation

```
float Rsum (float *a, const int n)
{
    count++; // if conditional
    if (n <= 0) {
        count++; // return statement
        return 0;
    } else {
        count++; // return statement
        return (Rsum(a, n-1) + a[n-1]);
    }
}
```

count is a global variable and will be incremented throughout the entire recurrent computation.



# Step Counting — Table

| float Rsum (float *a, const int n) | s/e      | freq. |              | subtotal |                 |
|------------------------------------|----------|-------|--------------|----------|-----------------|
|                                    |          | n=0   | n>0          | n=0      | n>0             |
| {                                  | 0        |       |              |          |                 |
| if (n <= 0)                        | 1        | 1     | 1            | 1        | 1               |
| return 0;                          | 1        | 1     | 0            | 1        | 0               |
| else                               | 0        |       |              |          |                 |
| return (Rsum(a, n-1) + a[n-1]);    | 1+t(n-1) | 0     | 1            | 0        | 1+t(n-1)        |
| }                                  | 0        |       |              |          |                 |
|                                    |          |       | <b>total</b> | <b>2</b> | <b>2+t(n-1)</b> |

s/e: steps per execution

Recurrence relations:  

$$t(n) = \begin{cases} 2 + t(n - 1), n > 0 \\ 2, otherwise \end{cases}$$



# Solving Recurrence

- Technique
  - Repeatedly substituting

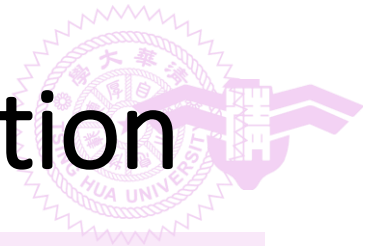
- $t(n) = 2 + t(n - 1)$ 
  - $= 2 + 2 + t(n - 2)$
  - $= 2 + 2 + \dots + 2 + t(0)$
  - $= 2n + t(0)$
  - $= 2n + 2$



# Step Counting — Example 3

```
void MatAdd (int **a, int **b, int **c, int m, int n)
{
    for (int i = 0; i < m; i++) {
        for (int j = 0; j < n; j++) {
            c[i][j] = a[i][j] + b[i][j];
        }
    }
    return;
}
```

Program containing nested loops



# Step Counting — Instrumentation

```
void MatAdd (int **a, int **b, int **c, int m, int n)
{
    for (int i = 0; i < m; i++) {
        count++; // for loop i
        for (int j = 0; j < n; j++) {
            count++; // for loop j
            c[i][j] = a[i][j] + b[i][j];
            count++; // assignment
        }
        count++; // last time of for loop j
    }
    count++; // last time of for i
    count++; // return statement
    return;
}
```

The textbook omits the return



# Step Counting — Table

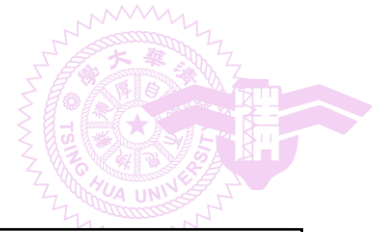
```
void MatAdd (int **a, int **b, int **c, int m, int n)
```

|                              | s/e | freq.         | subtotal        |
|------------------------------|-----|---------------|-----------------|
| {                            | 0   |               |                 |
| for (int i = 0; i < m; i++)  | 1   | m+1           | m+1             |
| for (int j = 0; j < n; j++)  | 1   | m(n+1)        | mn+m            |
| c[i][j] = a[i][j] + b[i][j]; | 1   | mn            | mn              |
| return;                      | 1   | 1             | 1               |
| }                            | 0   |               |                 |
|                              |     | <b>total:</b> | <b>2mn+2m+2</b> |

The textbook omits the return

We are allowed to use more than one variables to describe problem size





# Step Counting — Example 4

```
void fibonacci (int n) //compute the Fibonacci number F[n]
{
    if (n <= 1) // steps = 1
        cout << n << endl; // F[0] = 0 and F[1] = 1 // steps = 1
    else { // compute F[n]
        int fn; int fnm2 = 0; int fnm1 = 1; // steps = 2
        for (int i = 2; i<=n; i++) { // steps = n
            fn = fnm1 + fnm2;
            fnm2 = fnm1;
            fnm1 = fn;
        } // end of for // steps = 1
        cout << fn << endl; // steps = 1
    } // end of else
    return; // steps = 1
} // end of fibonacci
```

If  $n > 1$ ,  
 $t(n) = 1 + 2 + n + 3(n-1) + 1 + 1 + 1$   
 $= 4n + 2$   
Otherwise,  $t(n) = 1 + 1 + 1 = 3$



# Inexactness of Step Count

- We cannot know which following program exhibits the shortest execution time for the same problem size

- $t_1(n) = n+1$
- $t_2(n) = n+1000$
- $t_3(n) = 1000n$
- $t_4(n) = 1000n+1000$

Since the notion of a step is (deliberately) imprecise  
One multiplications  $\rightarrow$  1 step  
100 multiplications  $\rightarrow$  1 step

- But we know the execution time of these programs **linearly increases** with problem size

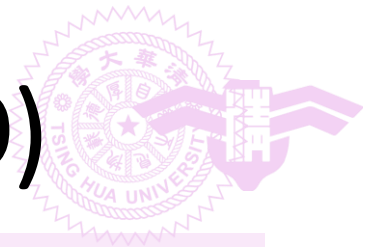


# Motivation of Asymptotic Notation

- We also know the fifth program exhibits the shortest execution time once the problem size,  $n$ , is large enough

- $t_1(n) = n+1$
  - $t_2(n) = n+1000$
  - $t_3(n) = 1000n$
  - $t_4(n) = 1000n+1000$
  - $t_5(n) = \log(n)+1$
- } Linearly increase
- } Logarithmically increase

- **Asymptotic Notations** are introduced to emphasize
  - **Trend** that step count increases with problem size
  - **Classification** of problems/algorithms based on the trend

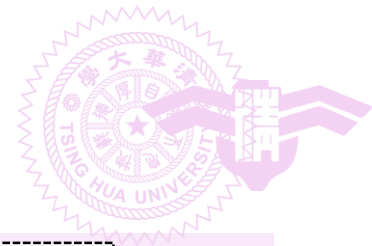


# Asymptotic Notations ( $O$ , $\Omega$ , $\Theta$ )

|          |       |                                                         |
|----------|-------|---------------------------------------------------------|
| $O$      | Big O | Upper bound                                             |
| $\Theta$ | Theta | Tight bound (i.e., both an upper bound and lower bound) |
| $\Omega$ | Omega | Lower bound                                             |

- “ $f(n) = O(n)$ ” reads as
  - “ $f$  of  $n$  is big O of  $n$ ”
- We can alternatively say “ $f(n) \in O(n)$ ”
  - “ $f$  of  $n$  belongs to big O of  $n$ ”

- “**Big**”  $O \rightarrow$  Upper
- “ **$\Theta$** ”  $\rightarrow$  A hyphen in the middle  $\rightarrow$  tight bound



# Big O (Cont'd)

•  $f(n) = O(g(n))$  iff

“iff” means “if and only if” (“ $\Leftrightarrow$ ”)

- there exist positive constants  $c$  and  $n_0$  such that  $f(n) \leq c \cdot g(n)$  for all  $n, n \geq n_0$

“ $\leq$ ” suggests that  $c \cdot g(n)$  is an upper bound of  $f(n)$

“ $\forall$ ” means “for all”

## • Example

- |                                  |                 |                                                                  |                     |
|----------------------------------|-----------------|------------------------------------------------------------------|---------------------|
| • <u><math>n+1</math></u>        | $= O(n),$       | <u><math>n+1</math></u> $\leq 2 \cdot \underline{n}$             | $\forall n \geq 1$  |
| • <u><math>n+1000</math></u>     | $= O(n),$       | <u><math>n+1000</math></u> $\leq 1001 \cdot \underline{n}$       | $\forall n \geq 1$  |
| • <u><math>1000n</math></u>      | $= O(n),$       | <u><math>1000n</math></u> $\leq 1000 \cdot \underline{n}$        | $\forall n \geq 1$  |
| • <u><math>1000n+1000</math></u> | $= O(n),$       | <u><math>1000n+1000</math></u> $\leq 2000 \cdot \underline{n}$   | $\forall n \geq 1$  |
| • <u><math>\log(n)+1</math></u>  | $= O(\log(n)),$ | <u><math>\log(n)+1</math></u> $\leq 2 \cdot \underline{\log(n)}$ | $\forall n \geq 10$ |



# Big O (Cont'd)

- More examples

- $\frac{2n^2+3n+4}{n^2} = O(\underline{n^2}),$

$$\frac{2n^2+3n+4}{n^2} \leq 9 \cdot \underline{n^2} \quad \forall n \geq 1$$

- $\frac{2n^2+3n+4}{n^2} = O(\underline{n^2}),$

$$\frac{2n^2+3n+4}{n^2} \leq 90 \cdot \underline{n^2} \quad \forall n \geq 40$$

We may have an infinite number of c and n0 satisfying the inequality.

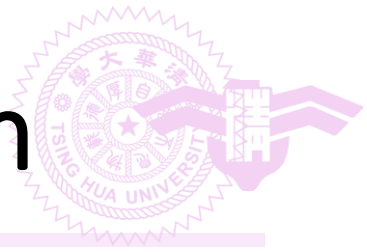
- $\frac{2n^2+3n+4}{n^2} = O(\underline{n^{2.1}}),$

- $\frac{2n^2+3n+4}{n^2} = O(\underline{n^3}),$

- $\frac{2n^2+3n+4}{n^2} = O(\underline{n^{99}}),$

Since by definition, Big O does not need to be a tight bound, we may have infinite number of g(n) satisfying the inequality.

- $\frac{2n^2+3n+4}{n^2} \neq O(\underline{n^{1.9}}),$



# Big O of a Polynomial Function

- **Theorem 1.2**

- $f(n) = a_m n^m + \dots + a_1 n + a_0$   
 $\Rightarrow f(n) = \mathbf{O}(n^m)$

- **Proof**

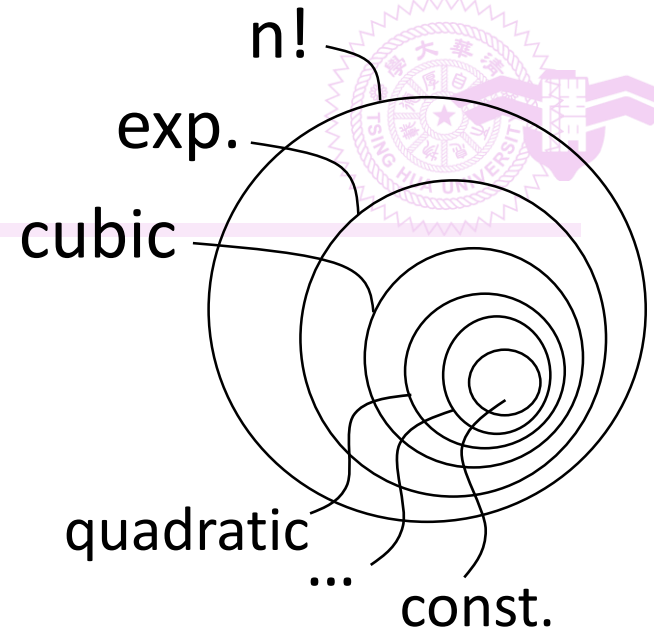
- $f(n) = \sum_{i=0}^m a_i n^i \leq \sum_{i=0}^m |a_i| n^i$

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

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

# Big O Hierarchy

- $O(n!)$  factorial
- $O(2^n)$  exponential
- $O(n^k)$
- ...
- $O(n^3)$  cubic
- $O(n^2)$  quadratic
- $O(n \log(n))$  log-linear
- $O(n)$  linear
- $O(n^{0.5})$  sub-linear
- $O(\log(n))$  logarithm
- $O(1)$  constant

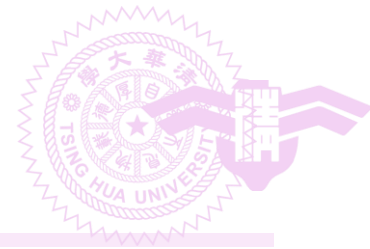


$O(n^2)$  algorithms/problems are also  $O(n^3)$  ones, and so on

Many other classes are not listed here, e.g.,  $O(n^{1.5})$ ,  $O(\log \log(n))$ ,  $O(n \log^2(n))$ ...

- **$O(1)$**  means that the execution time is independent of problem size
- E.g., time for retrieving the  $k^{\text{th}}$  entry of an array (of size  $n$ ) is  $O(1)$





# Omega

- $f(n) = \Omega(g(n))$  iff
  - there exist positive constants  $c$  and  $n_0$  such that  $f(n) \geq c \cdot g(n)$  for all  $n, n \geq n_0$

↕ Compare with Big O

such that  $f(n) \leq c \cdot g(n)$  for all  $n, n \geq n_0$

- Example

|                                  |                                  |                                                                                                |                                         |
|----------------------------------|----------------------------------|------------------------------------------------------------------------------------------------|-----------------------------------------|
| • <u><math>n+1</math></u>        | $= \Omega(\underline{n}),$       | <u><math>n+1</math></u> $\geq$ <u><math>1</math></u> $\cdot$ <u><math>n</math></u>             | $\forall n \geq$ <u><math>1</math></u>  |
| • <u><math>n+1000</math></u>     | $= \Omega(\underline{n}),$       | <u><math>n+1000</math></u> $\geq$ <u><math>1</math></u> $\cdot$ <u><math>n</math></u>          | $\forall n \geq$ <u><math>1</math></u>  |
| • <u><math>1000n</math></u>      | $= \Omega(\underline{n}),$       | <u><math>1000n</math></u> $\geq$ <u><math>1000</math></u> $\cdot$ <u><math>n</math></u>        | $\forall n \geq$ <u><math>1</math></u>  |
| • <u><math>1000n+1000</math></u> | $= \Omega(\underline{n}),$       | <u><math>1000n+1000</math></u> $\geq$ <u><math>1000</math></u> $\cdot$ <u><math>n</math></u>   | $\forall n \geq$ <u><math>1</math></u>  |
| • <u><math>\log(n)+1</math></u>  | $= \Omega(\underline{\log(n)}),$ | <u><math>\log(n)+1</math></u> $\geq$ <u><math>1</math></u> $\cdot$ <u><math>\log(n)</math></u> | $\forall n \geq$ <u><math>10</math></u> |



# Omega (Cont'd)

- More examples

- $\frac{2n^2+3n+4}{n^2}$  =  $\Omega(n^2)$ ,
- $\frac{2n^2+3n+4}{n^{1.9}}$  =  $\Omega(n^{1.9})$ ,
- $\frac{2n^2+3n+4}{n}$  =  $\Omega(n)$ ,
- $\frac{2n^2+3n+4}{1}$  =  $\Omega(1)$ ,
- $\frac{2n^2+3n+4}{n^{2.1}}$   $\neq \Omega(n^{2.1})$ ,

- **Theorem 1.3**

- $f(n) = a_m n^m + \dots + a_1 n + a_0, a_m > 0$   
 $\Rightarrow f(n) = \Omega(n^m)$



# Theta

- $f(n) = \Theta(g(n))$  iff
  - there exist positive constants  $c_1, c_2$  and  $n_0$  such that  $c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$  for all  $n, n \geq n_0$
  - i.e.,  $f(n)$  is  $O(g(n))$  and  $\Omega(g(n))$

- Example

- $\underline{n+1} = \Theta(\underline{n}), \quad \underline{1} \cdot \underline{n} \leq \underline{n+1} \leq \underline{2} \cdot \underline{n} \quad \forall n \geq \underline{1}$
- $\underline{n+1000} = \Theta(\underline{n}), \quad \underline{1} \cdot \underline{n} \leq \underline{n+1000} \leq \underline{1001} \cdot \underline{n} \quad \forall n \geq \underline{1}$
- $\underline{1000n} = \Theta(\underline{n}), \quad \underline{1000} \cdot \underline{n} \leq \underline{1000n} \leq \underline{1000} \cdot \underline{n} \quad \forall n \geq \underline{1}$
- $\underline{1000n+1000} = \Theta(\underline{n}), \quad \underline{1000} \cdot \underline{n} \leq \underline{1000n+1000} \leq \underline{2000} \cdot \underline{n} \quad \forall n \geq \underline{1}$
- $\underline{\log(n)+1} = \Theta(\underline{\log(n)}), \quad \underline{1} \cdot \underline{\log(n)} \leq \underline{\log(n)+1} \leq \underline{2} \cdot \underline{\log(n)} \quad \forall n \geq \underline{10}$

- Theorem 1.4

- $f(n) = amn^m + \dots + a_1n + a_0, a_m > 0$   
 $\Rightarrow f(n) = \Theta(n^m)$



# Step Counting — Asymptotic Notation

| <b>float sum (float *a, const int n)</b> | <b>s/e</b> | <b>freq.</b>  | <b>subtotal</b>               |
|------------------------------------------|------------|---------------|-------------------------------|
| {                                        | 0          |               |                               |
| <b>float s = 0;</b>                      | 1          | $\Theta(1)$   | $\Theta(1)$                   |
| <b>for (int i = 0; i &lt; n; i++)</b>    | 1          | $\Theta(n)$   | $\Theta(n)$                   |
| <b>s += a[i];</b>                        | 1          | $\Theta(n)$   | $\Theta(n)$                   |
| <b>return s;</b>                         | 1          | $\Theta(1)$   | $\Theta(1)$                   |
| }                                        | 0          |               |                               |
|                                          |            | <b>total:</b> | <b><math>\Theta(n)</math></b> |

**s/e**: number of steps per execution



# Step Counting — Asymptotic Notation

| (recursion of sum())<br><b>float Rsum (float *a, const int n)</b> | <b>s/e</b> | <b>freq.</b> |               | <b>subtotal</b>               |                                      |
|-------------------------------------------------------------------|------------|--------------|---------------|-------------------------------|--------------------------------------|
|                                                                   |            | <b>n=0</b>   | <b>n&gt;0</b> | <b>n=0</b>                    | <b>n&gt;0</b>                        |
| {                                                                 | 0          |              |               |                               |                                      |
| <b>if (n &lt;= 0)</b>                                             | 1          | $\Theta(1)$  | $\Theta(1)$   | $\Theta(1)$                   | $\Theta(1)$                          |
| <b>return 0;</b>                                                  | 1          | $\Theta(1)$  | 0             | $\Theta(1)$                   | 0                                    |
| <b>else</b>                                                       | 0          |              |               |                               |                                      |
| <b>return (Rsum(a, n-1) + a[n-1]);</b>                            | $1+t(n-1)$ | 0            | $\Theta(1)$   | 0                             | $\Theta(1+t(n-1))$                   |
| }                                                                 | 0          |              |               |                               |                                      |
|                                                                   |            |              | <b>total</b>  | <b><math>\Theta(1)</math></b> | <b><math>\Theta(1+t(n-1))</math></b> |

**s/e**: number of steps per execution



# Step Counting — Asymptotic Notation

```
void MatAdd (int **a, int **b, int **c, int m, int n)
```

```
{  
  for (int i = 0; i < m; i++)       $\Theta(m)$   
  |   for (int j = 0; j < n; j++)  }  $\Theta(mn)$   
  |     c[i][j] = a[i][j] + b[i][j];  
return;  
}
```

**total:**  $\Theta(mn)$

# Recursive Permutation Generator



```
void Permutations(int *a, const int k, const int m)
{
    // one element between k and m means one possible permutation
    if(k == m) {
        for(int i=0; i<=m; i++)
            cout << a[i] << " ";
        cout << endl;
        return;
    }

    for(int i=k; i<=m; i++) {
        swap(a[k], a[i]);
        Permutations(a, k+1, m);
        swap(a[k], a[i]);
    }
}
```

$k=m$   
 $\rightarrow \Theta(t(k, m)) = \Theta(m)$

$\Theta(t(k, m)) =$   
 $(m-k+1) \times \Theta(t(k+1, m)) + \Theta(1)$

$\Theta(1)$  comes from the if statement



# Recursive Permutation Generator

## Solve the recurrence

$$\Theta(t(k, m)) = (m-k+1) \times \Theta(t(k+1, m)) + \Theta(1) \quad \text{Eq. (1)}$$

$$\Theta(t(m, m)) = \Theta(m) \quad \text{Eq. (2)}$$

Let  $k=0$  and  $m=(n-1)$

$$\begin{aligned} \Theta(t(0, n-1)) &= n \times \Theta(t(1, n-1)) + \Theta(1) \\ &= n \times (n-1) \times \Theta(t(2, n-1)) + \Theta(1) + \Theta(1) \\ &= \dots \\ &= \underbrace{n \times (n-1) \times (n-2) \dots \times 2}_{n-1 \text{ terms}} \times \Theta(t(n-1, n-1)) + (n-1) \times \Theta(1) \\ &= n! \times \Theta(t(n-1, n-1)) + \Theta(n-1) \\ &= n! \times \Theta(n-1) + \Theta(n-1) \quad \dots \text{ because of Eq. (2)} \\ &= \Theta(n \times n!) \end{aligned} \quad \left. \vphantom{\begin{aligned} \Theta(t(0, n-1)) &= n \times \Theta(t(1, n-1)) + \Theta(1) \\ &= n \times (n-1) \times \Theta(t(2, n-1)) + \Theta(1) + \Theta(1) \\ &= \dots \\ &= \dots \end{aligned}} \right\} n-1 \text{ equations}$$

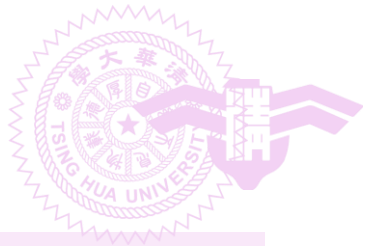




# Binary Search

```
int BinarySearch(int *a, const int x, const int n)
{ //Search the sorted array a[0], ... , a[n-1] for x
  int left = 0, right = n-1;
  while(left <= right)
  { //there are more elements
    int middle =(left+right)/2;
    if(x<a[middle])      right=middle-1;
    else if(x>a[middle]) left = middle+1;
    else return middle;
  } //end of while
  return -1;
}
```

}  $\Theta(\log(n))$

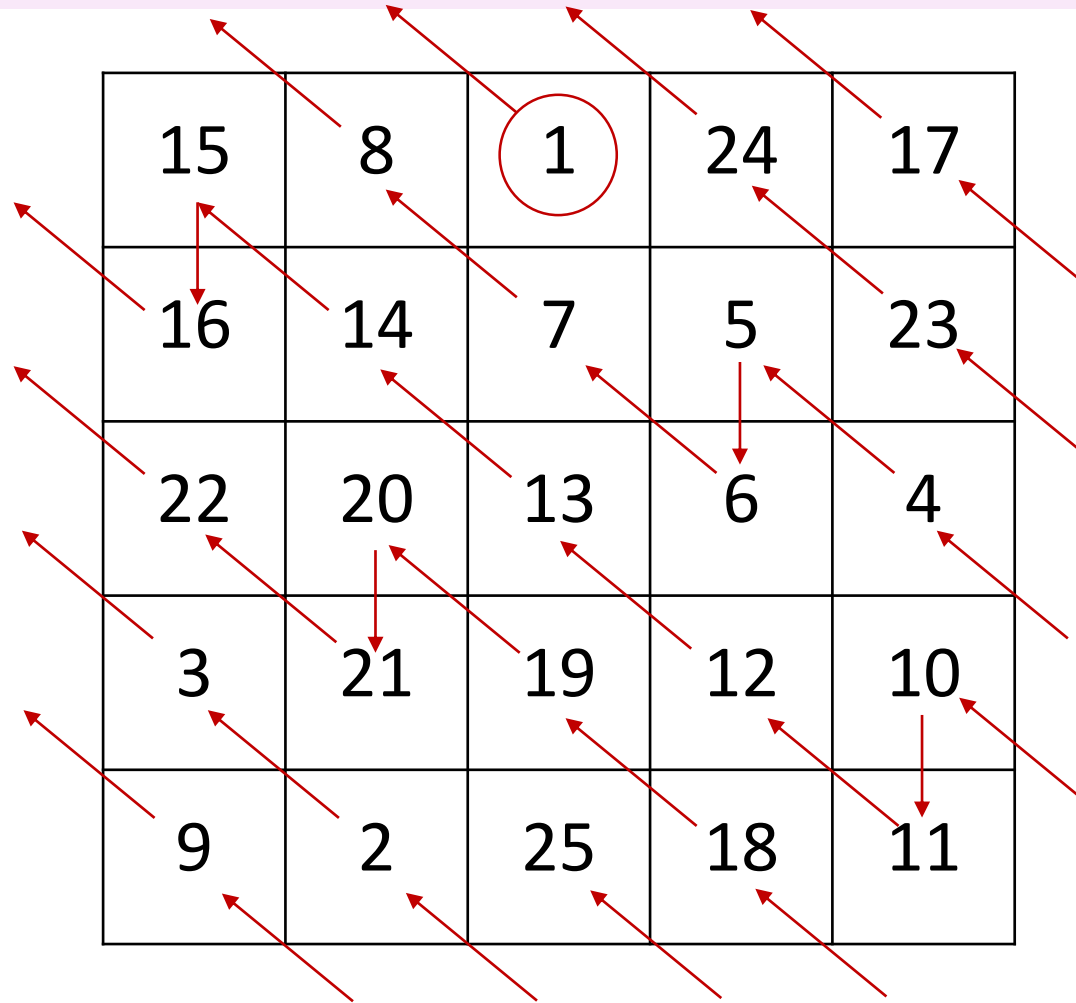


# Magic Square

|      |      |      |      |      |      |
|------|------|------|------|------|------|
| 15   | 8    | 1    | 24   | 17   | = 65 |
| 16   | 14   | 7    | 5    | 23   | = 65 |
| 22   | 20   | 13   | 6    | 4    | = 65 |
| 3    | 21   | 19   | 12   | 10   | = 65 |
| 9    | 2    | 25   | 18   | 11   | = 65 |
| = 65 | = 65 | = 65 | = 65 | = 65 | = 65 |



# Generate the Magic Square





# Magic Square Algorithm

```
void magic (int n)
// create a magic square of size n, n is odd
{
    const int MaxSize = 51; // maximum square size
    int square[MaxSize][MaxSize], k, l;
} }  $\Theta(1)$ 

// check correctness of n
if ((n > MaxSize) || (n < 1)) {
    cerr << "Error!..n out of range \n";
    return;
} else if (!(n%2)) {
    cerr << "Error!..n is even \n";
    return;
} }  $\Theta(1)$ 

// n is odd. Coxeter's rule can be used
for (int i = 0; i < n; i++) // initialize square to 0
    for (int j = 0; j < n; j++)
        square[i][j] = 0;
} }  $\Theta(n^2)$ 

square[0][(n-1)/2] = 1; // middle of first row
// please continue to the next slide.. }  $\Theta(1)$ 
```

```
// i and j are current position
```

```
int key = 2; i = 0;
```

```
int j = (n-1)/2;
```

```
while (key <= n*n) {
```

```
    // move up and left
```

```
    if (i-1 < 0) k = n-1; else k = i-1;
```

```
    if (j-1 < 0) l = n-1; else l = j-1;
```

```
    if (square[k][l]) i = (i+1)%n;
```

```
    else { // square[k][l] is unoccupied
```

```
        i = k;
```

```
        j = l;
```

```
    }
```

```
    square[i][j] = key;
```

```
    key++;
```

```
} // end of while
```

```
// output the magic square
```

```
cout << "magic square of size " << n << endl;
```

```
for ( i = 0; i < n; i++) {
```

```
    for ( j = 0; j < n; j++)
```

```
        cout << square[i][j] << " ";
```

```
    cout << endl;
```

```
}
```

```
}
```

}  $\Theta(1)$

}  $\Theta(n^2)$

}  $\Theta(1)$

}  $\Theta(n^2)$



# Magic Square (Cont'd)

---

- We just show how can we quickly analyze the complexity of an algorithm without knowing all the details
- $\Theta(n^2)$  is the optimal one we can achieve (in terms of asymptotic complexity) to generate an  $n^2$  magic square
  - Since there are  $n^2$  positions the algorithm must place a number



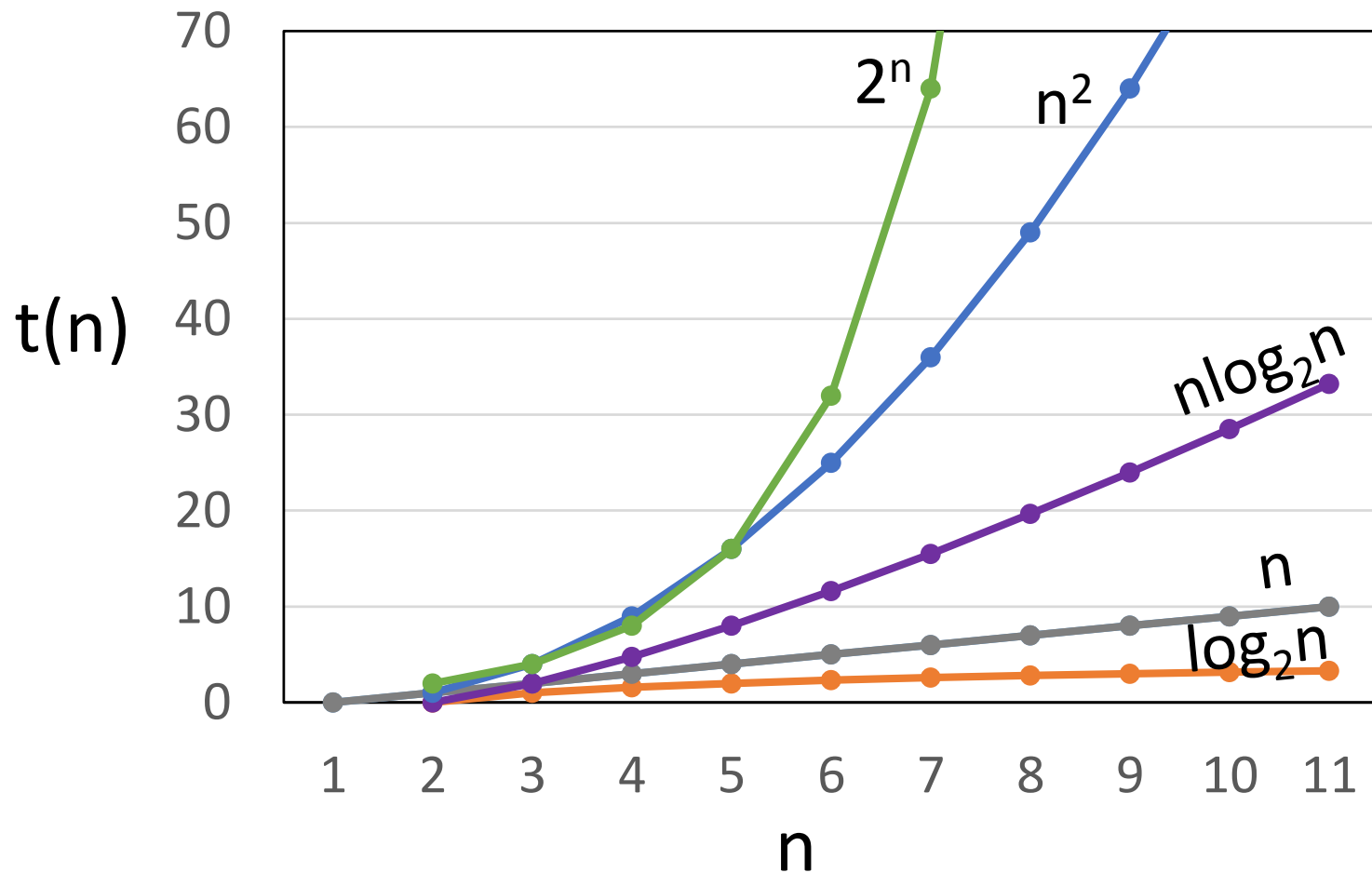
# Practical Complexities

| Prob. size | $n$        | $n \log(n)$ | $n^2$  | $n^3$ | $n^4$             | $2^n$                   |
|------------|------------|-------------|--------|-------|-------------------|-------------------------|
| $10^3$     | 1 $\mu$ s  | 10 $\mu$ s  | 1 ms   | 1 s   | 17 min            | $3.2 \times 10^{283}$ y |
| $10^4$     | 10 $\mu$ s | 130 $\mu$ s | 100 ms | 17 m  | 116 d             |                         |
| $10^5$     | 0.1 ms     | 1.7 ms      | 10 s   | 12 d  | 3171 y            |                         |
| $10^6$     | 1 ms       | 20 ms       | 17 m   | 32 y  | $3 \times 10^7$ y |                         |

Assume a 1-billion-steps-per-second computer



# Practice Complexity







# Performance Measurement

---

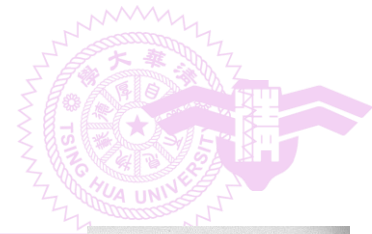
- Techniques
  - Use time-related library functions
    - `gettimeofday()`
    - `clock()`
    - `time()`
  - Repeatedly measure a program to reduce noises
  - Use randomized inputs to obtain best-case, average, and worst-case execution time
  - Prediction
    - Regression (curve fitting)
    - Interpolation
    - Extrapolation
- Please read Section 1.7.2 for details



# Performance Measurement

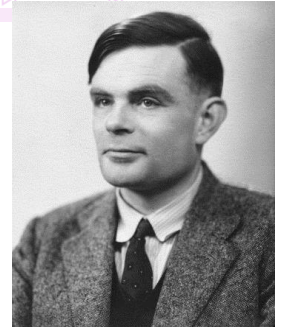
---

- Benefits
  - Provide actual execution time
- Limitations of asymptotic analysis
  - For two programs that are both  $O(n^2)$  time complexity
    - We cannot tell which is faster
  - For one program that is  $O(n)$  and the other is  $O(n^2)$ 
    - The  $O(n)$  one can be slower for a practical size of  $n$



# Alan Turing

- One of the greatest **computer scientists** and **computational theorists**
  - Complexity analysis is part of computational theory
- Often called the father of modern computing
- Some famous things
  - Turing award
    - Nobel Prize of computing
  - Turing machine
    - Theoretical computer model
    - <http://www.google.com/doodles/alan-turings-100th-birthday>
  - Turing test
    - Test of a computer's ability to exhibit behavior equivalent to human





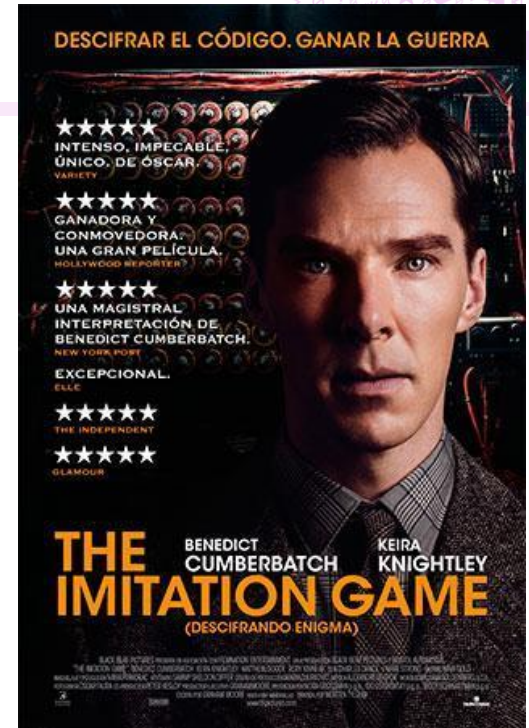
# Alan Turing (Cont'd)

- The Imitation Game
  - A movie about Alan Turing trying to crack the enigma code during World War II
- IMDB 8.2

## User Reviews

★★★★★★★★★ **Compelling and Enthralling from start to finish.**  
16 October 2014 | by [fruitbat00](#) (United Kingdom) – [See all my reviews](#)

Truly excellent film and definitely Oscar worthy material for both the film and the actors. The entire cast are amazing.





# Complexity of Learning DS

- $\Theta(1)$ 
  - Number of weeks in the semester  
= 18 =  $\Theta(1)$
  - Number of chapters covered in the semester  
= 8 =  $\Theta(1)$
  - Time(read these chapters twice)  
=  $2 \times 8 \times \text{Time}_{\text{read\_one\_chapter}}$   
=  $\Theta(1)$

