Performance Analysis

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Review

- Algorithm and Program
- Recursive Functions
 - Direct
 - Indirect
 - Tail
 - Compared with non-recursive functions

Space and Time Complexities

- There are many criteria upon which we can judge a program/algorithm
 - Does it do what we want it to do?
 - Dos it work correctly?
 - Is there documentation for the program/algorithm?
 - Is the code readable?
- A reasonable way is to judge programs by considering their computing time and storage requirements

Definition: The *space complexity* of a program is the amount of memory it needs to run to completion. The *time complexity* of a program is the amount of computer time it needs to run to completion. \Box

Hard to Imagine?





技嘉 AORUS GTX 1080Ti 11G 顯示卡

◆ 顯示晶片: NVIDIA GeForce GTX 1080 Ti

◆ 記憶體: 11GB GDDR5X

◆ 核心時脈: 1569 MHz

◆ 記憶體時脈: 11010 MHz

◆ 記憶體介面: 352-bit

◆ 最高解析度:7680x4320

◆ 輸出端子: 3x DP / 3x HDMI / DVI

◆ 電源連結器: 2x 8-pin

◆ 體積(長x寬x高): 29.3 x 14.2 x 5.5cm



▼贈TOSHIBA 64GB 隨身碟▼

▼頂級日製商務機●新上市▼

Fujitsu 富士通LIFEBOOK S937-PB722黑 ★第七代Core i7-7500U+512GB SSD+掌靜脈辨 識器

處理器:第七代Intel® Core™ i7-7500U

• 螢幕:13.3"高解析FHD防眩光護眼螢幕

• (解析度1920 x 1080)

• 記憶體: 24GB DDR4 2133MHz

硬碟: 512GB SATAIII M.2 SSD

作業系統: Windows 10 Pro 64bits

• USB3.1 x3 (Gen 1; 關機充電技術x1);

• HDMI x1; VGA x1; 擴充底座介面

• 內建富士通專利掌靜脈辨識器

• 起始重量1.13kg起

3期0利率	<u>32家</u>	18期0利率	<u>17家</u>
6期0利率	<u>32家</u>	24期0利率	<u>15家</u>
10期0利率	<u>27家</u>	30期0利率	<u>3家</u>
12期0利率	<u>11家</u>	12期 分期	<u>7家</u>

建議售價 \$79999

網路價\$78800

VISA 😂 🗱 🚾 ATM 說明

信用卡紅利折抵刷卡金 多家銀行

PChome儲值

LINE Pay Pay

完售,諸參考其他商品

Space Complexity

- The space analysis can be classified into two parts
 - Fixed part
 - The instruction space, space for simple variables, space for constants, etc
 - Variable part
 - Space needed by referenced variables
 - The recursion stack space
- The space requirement S(P) of a program P can be defined

$$S(P) = c + S_p$$
fixed part variable part
usually a constant depend on the task

– We usually concentrate on S_p

Recursion Stack Space.

• Given an Ackerman's function A(m, n), please calculate A(1,2).

$$A(m,n) = \begin{cases} n+1, & if \ m=0 \\ A(m-1,1), & if \ n=0 \\ A(m-1,A(m,n-1)), & otherwise \end{cases}$$

$$A(1,2) = A(0,A(1,1))$$

$$A(1,1) = A(0,A(1,0))$$

$$A(1,0) = A(0,1)$$

$$A(0,1) = 2$$



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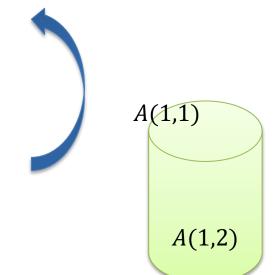
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Time Complexity

- The time, T(P), taken by a program P is the sum of the **compile time** and the **run (execution) time**
 - We mainly concentrate on the run time of a program

$$T(P) = c + T_p$$
compile time run time

- There are two ways to determine the run time
 - Measurement
 Execute the program

Record the CPU time

• Analysis

Count only the number of program steps

Count the number of instructions

Examples

• How many times does the function *call_fun()* execute?

$$\sum_{n=1}^{n} (a^2 - a) = \sum_{n=1}^{n} a^2 - \sum_{n=1}^{n} a = \frac{n(n+1)(2n+1)}{6} - \frac{n(n+1)}{2} = \frac{n(n+1)(n-1)}{3}$$

$$\sum_{n=1}^{n} a^2 = 1^2 + 2^2 + \dots + n^2 = \frac{n(n+1)(2n+1)}{6}$$

Asymptotic Notations

 We introduce some terminology that will enable is to make meaningful but inexact statements about the time and space complexities of a program

Definition [Big "oh"]: f(n) = O(g(n)) (read as "f of n is big oh of g of n") iff (if and only if) there exist positive constants c and n_0 such that $f(n) \le cg(n)$ for all $n, n \ge n_0$. \square

Definition: [Omega] $f(n) = \Omega(g(n))$ (read as "f of n is omega of g of n") iff there exist positive constants c and n_0 such that $f(n) \ge cg(n)$ for all $n, n \ge n_0$. \square

Definition: [Theta] $f(n) = \Theta(g(n))$ (read as "f of n is theta of g of n") iff there exist positive constants c_1, c_2 , and n_0 such that $c_1g(n) \le f(n) \le c_2g(n)$ for all $n, n \ge n_0$. \square

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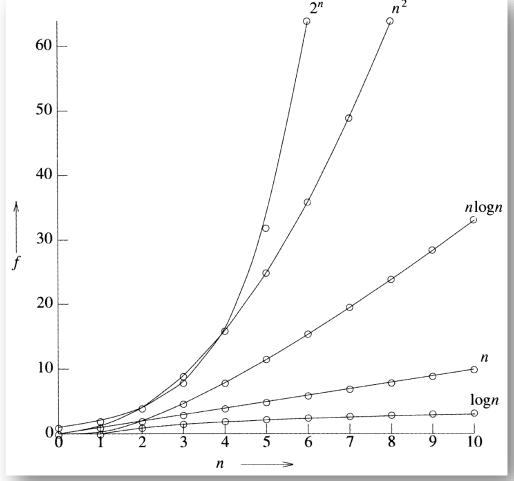
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- For the statement f(n) = O(g(n)) to be **informative**, g(n) should be as small a function of n as one can come up with
 - -3n + 3 = 0(n) vs. $3n + 3 = 0(n^2)$
- Fantastic names
 - O(1) mean a computing time that is a constant
 - O(n) is called linear
 - $O(n^2)$ is called quadratic
 - $O(n^3)$ is called cubic
 - $O(2^n)$ is called exponential
- Ordering
 - $0(1) < 0(\log n) < 0(n) < 0(n\log n) < 0(n^2) < 0(n^3) < 0(2^n)$

Big-Oh...

• $0(1) < 0(\log n) < 0(n) < 0(n\log n) < 0(n^2) < 0(n^3) < 0(n^c) < 0(2^n) < 0(3^n) < 0(c^n) < 0(n!) < 0(n^n) < 0(n^c)^n$



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• The function g(n) is a lower bound on f(n)

- For the statement $f(n) = \Omega(g(n))$ to be informative, g(n) should be as large a function of n as possible
 - $-3n + 3 = \Omega(n)$ vs. $3n + 3 = \Omega(1)$
 - $-6 \times 2^{n} + n^{2} = \Omega(2^{n}) \text{ vs. } 6 \times 2^{n} + n^{2} = \Omega(1)$

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- The theta is more precise than both big-oh and omega
 - g(n) is both an upper and lower bound on f(n)

Example 1.16: $3n + 2 = \Theta(n)$ as $3n + 2 \ge 3n$ for all $n \ge 2$, and $3n + 2 \le 4n$ for all $n \ge 2$, so $c_1 = 3$, $c_2 = 4$, and $n_0 = 2$. $3n + 3 = \Theta(n)$; $10n^2 + 4n + 2 = \Theta(n^2)$; $6*2^n + n^2 = \Theta(2^n)$; and $10*\log n + 4 = \Theta(\log n)$. $3n + 2 \ne \Theta(1)$; $3n + 3 \ne \Theta(n^2)$; $10n^2 + 4n + 2 \ne \Theta(n)$; $10n^2 + 4n + 2 \ne \Theta(1)$; $6*2^n + n^2 \ne \Theta(n^2)$; $6*2^n + n^2 \ne \Theta(n^{100})$; and $6*2^n + n^2 \ne \Theta(1)$. \square

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Example

• Given a recursive function $T(n) = 2T\left(\frac{n}{2}\right) + n$, where T(1) = 0, please write down the time complexity in big-oh for the function.

$$T(n) = 2 \times T\left(\frac{n}{2}\right) + n$$

$$= 2 \times \left[2 \times T\left(\frac{n}{4}\right) + \frac{n}{2}\right] + n = 4 \times T\left(\frac{n}{4}\right) + 2 \times n$$

$$= 4 \times \left[2 \times T\left(\frac{n}{8}\right) + \frac{n}{4}\right] + 2 \times n = 8 \times T\left(\frac{n}{8}\right) + 3 \times n$$

$$= \cdots$$

$$= n \times T\left(\frac{n}{n}\right) + (\log_2 n) \times n = n \log_2 n$$

$$\therefore T(n) = O(n \log_2 n)$$

Data Type

 All programming languages provide a set of predefined data types, and they also have the ability to construct new, and/or user-defined types

Definition: A *data type* is a collection of *objects* and a set of *operations* that act on those objects. \Box

- A data type should consider two aspects
 - Objects
 - Operations

Data Type	Objects	Operations	
integer	$0, \pm 1, \pm 2, \pm 3, \cdots$	+, -,×,÷,···	

Abstract Data Type

Definition: An abstract data type (ADT) is a data type that is organized in such a way that the specification of the objects and the specification of the operations on the objects is separated from the representation of the objects and the implementation of the operations. \Box

- ADT is implementation-independent
- Some programming languages provide explicit mechanisms to support the distinction between specification and implementation
 - Class in C++
 - Standard Template Library (STL) in
 C++
 - Map, Vector, List, ...
 - http://www.cplusplus.com/reference/stl



Questions?



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