PERFORMANCE EVALUATION

Office Hour 已公布在課程網站

Prof. Michael Tsai 2013/2/26

HW0 is due 23:59 on Thursday.

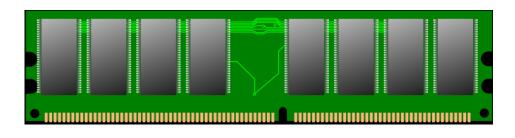
HW1 今天或明天 會公布



空間及時間複雜度

- 程式的空間複雜度:
 - 程式執行完畢所需使用的所有空間(記憶體)
- 程式的時間複雜度:
 - 程式執行完畢所需使用的(執行)時間
- Goal: 找出執行時間/使用空間"如何"隨著input size變長 (成長的有多快)
- 什麼是input size?
- · 問題給的input的"元素數量",如:
 - Array大小
 - 多項式最高項的次方
 - 矩陣的長寬
 - 二進位數的位元數目

空間複雜度



- •程式所需空間:
 - 1. 固定的空間
 - 和input/output的大小及內容無關
 - 2. 變動的空間
 - 和待解問題P的某個input instance I(某一個input)有關
 - 隨著input size變大而變大 ←我們需要注意的地方!
 - 跟recursive function會使用到的額外空間有關

$$\cdot S(P) = c + S_P(I)$$

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

- $S_{abc}(I) = ?$ (變動空間)
- 只有固定空間.
- $S_{abc}(I) = 0$.

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

- *S_{sum}(I)* =? (變動空間)
- $S_{sum}(I) = n$ (list陣列所占空間)

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

- $S_{rsum}(I) = ?$ (變動空間)
- list[]占n個sizeof(float)的大小
- ·另外每個recursive call需要紀錄:
- •float *list, int n, 還有return address. 假設這三樣東西需要k bytes (注意是list的指標, 非整個陣列)
- S(n) = S'(n) + sizeof(float) * n
- $\cdot S'(n) = S'(n-1) + k$
- $\bullet S'(0) = k$
- $\cdot S'(n) = (n+1)k$

時間複雜度

- 一個程式P所需使用的時間:
 - Compile所需時間
 - 執行時間 (execution time or run time)
 - How to do it on workstation?
- Compile時間: 固定的. (例外?)
 - C (and other compiled programming languages)
 - → One Compilation → Multiple Executions
- Run time:
 - 和input instance的特性有關!
 - 例如input size n 我們可以把run time寫成T(n)



如何得到T(n)?

- 1. 總執行時間
 - 花了多少時間?→有其他程式也在使用處理器!
 - 花了多少處理器時間? → 跟機器, 作業系統相關





- 2. 所執行的程式指令的數目 & 每個指令的時間
- 3. 比較數學的方法(使用function來代表程式執行的時間)

Is it good to use? (方法1)



一些小假設

- 使用一個處理器(processor). 循序一次執行一個指令. (無法 同時做超過一件事情)
- 常用的指令每個指令花固定時間(constant time)
 - +-*/%
 - 記憶體動作(讀取,修改,儲存)
 - 控制用的statement: 呼叫subroutine, branch等等

Insertion Sort

藍色: 已經排好的範圍

綠色:目前正在處理的數字

i	1	2	3	4	5	6
A[i]	5	2	4	6	1	3
i	1	2	3	4	5	6
A[i]	2	5	4	6	1	3
i	1	2	3	4	5	6
A[i]	2	4	5	6	1	3
i	1	2	3	4	5	6
A[i]	2	4	5	6	1	3
i	1	2	3	4	5	6
A[i]	1	2	4	5	6	3
i	1	2	3	4	5	6
A[i]	1	2	3	4	5	6

Insertion-Sort(A)

j=6

i	1	2	3	4	5	6
A[i]	1	2	4	5	6	3

執行時間=?

n=A.length

 t_i :j=i的時候那次 while執行次數

時間(cost) 做了幾次 Insertion-Sort (A) for j=2 to A.length key=A[j] \mathcal{C}_2 i=j-1while i>0 and A[i] > key C_5 A[i+1]=A[i]i=i-1 C_7 A[i+1]=keyn-1

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

Best case

• 最好的狀況?

i	1	2	3	4	5	6
A[i]	1	2	3	4	5	6

- Best case
- 全部已經排好了.

•
$$t_j = 1$$
, $for j = 2, ..., n$

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

$$= (c_1 + c_2 + c_4 + c_5 + c_8)n + (c_2 + c_4 + c_5 + c_8)$$

$$\bullet = an + b$$

Worst case

• 最糟的狀況?

i	1	2	3	4	5	6
A[i]	6	1	2	3	4	5

- Worst case
- 全部已經排好了.

•
$$t_i = j$$
, $for j = 2, ..., n$

•
$$\sum_{j=2}^{n} t_j = \frac{n(n+1)}{2} - 1$$

•
$$\sum_{j=2}^{n} t_j - 1 = \frac{n(n-1)}{2}$$

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

• =
$$\left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2}\right)n^2 + \left(c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8\right)n - (c_2 + c_4 + c_5 + c_8)$$

$$\bullet = an^2 + bn + c$$

Worst case, best case, and average case

- · Average case: 把所有狀況所花的時間(空間)平均起來
- · Worst case: 我們通常最常分析的
 - 最糟的狀況會花的時間(最多時間) →不可能更糟了(確定)
 - Worst case不代表不常發生 (常常發生): 例如 binary search找不到!
 - 很多時候, Average case可能也跟worst case差不了多少
- 例如: 如果隨意選擇n個數字來做insertion sort

 - 最後算出來還是一個二次方多項式 (quadratic function), 和worst case 相同

比較數學的方法—Asymptotic analysis

- "預測"當input的性質改變(通常是input size改變)時, 執行時間的成長速度(growth of the running time)
- 比較兩個做相同事情的程式的時間複雜度
- "程式步驟"不是那麼精確:
 - "3n+3" 比"3n+5" 快?
 - 通常 "3n+3", "7n+2", or "2n+15" 執行時間都相差不遠.
 - →我們將使用一個比較不準確的方式來描述執行時間...
- 相對的, 如果是 $3n^2 + 5n + 7$ 和10n + 50, n^2 和n就非常重要!
- "Asymptote" 是漸近線。

- Program P and Q
- $T_P(n) = c_1 n^2 + c_2 n$
- $T_Q(n) = c_3 n$

Usually no (if the constants are close).



"小時候胖不算胖"

- N很大的時候, Q 會比 P 快, 不管 c_1, c_2, c_3 的數值是什麼.
- Example:
 - $c_1 = 1$, $c_2 = 2$, $c_3 = 100$, then $c_1 n^2 + c_2 n^2 > c_3 n$ for n > 98.
 - $c_1 = 1, c_2 = 2, c_3 = 1000$, then $c_1 n^2 + c_2 n^2 > c_3 n$ for n > 998

Break even point

- 需要知道 c_1, c_2, c_3 的數值嗎?
- No.

$$T(n) = an^2 + bn + c = \Theta(n^2)$$

Asymptotic Notation – Big OH

- Definition [Big "oh"]:
- O(g(n)) ={f(n): there exist positive constants c and n_0 such that $0 \le f(n) \le cg(n)$ for all $n \ge n_0$ }

- "f of n is big oh of g of n" (是集合的成員)
- "=" is "is" not "equal" ("∈"的意思)
- -O(g(n)) = f(n)
- 可以想成是Upper Bound (王先生打了太多全壘打,所以是upper bound)



- 3n + 2 = O(n) ?
- Yes, since $3n + 2 \le 4n$ for all $n \ge 2$.
- 3n + 3 = O(n) ?
- Yes, since $3n + 3 \le 4n$ for all $n \ge 3$.
- 100n + 6 = O(n) ?
- Yes, since $100n + 6 \le 101n$ for all $n \ge 10$.
- $10n^2 + 4n + 2 = O(n^2)$?
- Yes, since $10n^2 + 4n + 2 \le 11n^2$ for all $n \ge 5$.

- $1000n^2 + 100n 6 = O(n^2)$?
- Yes, since $1000n^2 + 100n 6 \le 1001n^2$ for all $n \ge 100$.
- $6 * 2^n + n^2 = 0(2^n)$?
- Yes, since $6 * 2^n + n^2 \le 7 * 2^n$ for all $n \ge 4$.
- $3n + 3 = O(n^2)$?
- Yes, since $3n + 3 \le 3n^2$ for all $n \ge 2$.
- $10n^2 + 4n + 2 = O(n^4)$?
- Yes, since $10n^2 + 4n + 2 \le 10n^4$ for all $n \ge 2$.
- 3n + 2 = 0(1)?
- No. Cannot find c and n_0 . 3n < c 2 無法永遠成立.

 $f(n) \le cg(n)$ for all $n, n \ge n_0$

The World of Big Oh

- $0(1) \rightarrow$ constant
- $O(n) \rightarrow linear$
- $O(n^2) \rightarrow$ quadratic
- $O(n^3) \rightarrow \text{cubic}$
- $O(2^n) \rightarrow \text{exponential}$
- $O(1), O(\log n), O(n), O(n\log n), O(n^2), O(n^3), O(2^n)$

Faster



Slower



On a 1 billion-steps-per-sec computer

n	n	$n \log_2 n$	n^2	n^3	n^4	n^{10}	2^n
10	$.01 \mu s$.03µs	$.1\mu s$	$1\mu s$	$10\mu s$	10 <i>s</i>	$1\mu s$
20	.02µs	.09µs	.4μs	8µs	160μs	2.84h	1ms
30	.03µs	.15 <i>μs</i>	.9µs	27μs	810μs	6.83 <i>d</i>	1 <i>s</i>
40	$.04\mu s$.21µs	1.6µs	64 <i>μs</i>	2.56 <i>ms</i>	121 <i>d</i>	18 <i>m</i>
50	.05 <i>µs</i>	.28µs	2.5 <i>μs</i>	125 <i>μs</i>	6.25 <i>ms</i>	3.1 <i>y</i>	13 <i>d</i>
100	.10µs	.66µs	10μs	1ms	100ms	3171 <i>y</i>	4 * 10 ¹³ y
10 ³	1μs	9.96μs	1ms	1 <i>s</i>	16.67 <i>m</i>	3.17 * 10 ¹³ y	32 * 10 ²⁸³ y
10 ⁴	10μs	130μs	100 <i>ms</i>	16.67 <i>m</i>	115.7 <i>d</i>	$3.17 * 10^{23}y$	
10 ⁵	100μs	1.66 <i>ms</i>	10 <i>s</i>	11.57 <i>d</i>	3171 <i>y</i>	3.17 * 10 ³³ y	
10 ⁶	1ms	19.92 <i>ms</i>	16.67 <i>m</i>	31.71 <i>y</i>	$3.17 * 10^7 y$	3.17 * 10 ⁴³ y	

Big Oh 是 Upper Bound

· 但是沒有說它是多好的upper bound

$$\cdot n = O(n)$$

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

•
$$n = O(n^{2.5})$$

•
$$n = O(2^n)$$

• 通常我們會把它取得越小(緊)越好.

•
$$3n + 3 = O(n^2)$$

•
$$3n + 3 = O(n)$$





一些條件

- 1. 正式的定義下, O(g(n))裡面的n為自然數={0,1,2,...}
- O(g(n))的member f(n)必須為asymptotically nonnegative (也就是當n很大的時候, f(n)必須不為負)
- 3. g(n)本身也必須為asymptotically nonnegative

以上也適用於其他的asymptotic notation!

A Useful Theorem

- Theorem: If $f(n) = a_m n^m + \dots + a_1 n + a_0$, then $f(n) = 0(n^m)$.
- Proof:

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

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

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

, for all $n \ge 1$. So, $f(n) = O(n^m)$.

Asymptotic Notation – Omega

- Definition [Omega]:
- $\Omega(g(n)) = \{f(n): \text{ there exist positive }$ constants c and n_0 such that $0 \le$ $cg(n) \le f(n)$ for all $n \ge n_0\}$
- $f(n) = \Omega(g(n))$
- "f of n is omega of g of n"
- •可以想成是Lower Bound





- $3n + 2 = \Omega(n)$
- since $3n + 2 \ge 3n$ for all $n \ge 1$.
- $3n + 3 = \Omega(n)$
- since $3n + 3 \ge 3n$ for all $n \ge 1$.
- $100n + 6 = \Omega(n)$
- since $100n + 6 \ge 100n$ for all $n \ge 1$.
- $10n^2 + 4n + 2 = \Omega(n^2)$
- since $10n^2 + 4n + 2 \ge n^2$ for all $n \ge 1$.
- $6 * 2^n + n^2 = \Omega(2^n)$
- since $6 * 2^n + n^2 \ge 2^n$ for all $n \ge 1$.

- $3n + 3 = \Omega(1)$
- $10n^2 + 4n + 2 = \Omega(1)$
- $6 * 2^n + n^2 = \Omega(n^{100})$
- $6 * 2^n + n^2 = \Omega(n^{50.2})$
- $6 * 2^n + n^2 = \Omega(n^2)$
- $6 * 2^n + n^2 = \Omega(n)$
- $6 * 2^n + n^2 = \Omega(1)$

Discussion

- Omega is a lower bound.
- Should be as large a function as possible.
- Theorem: If $f(n)=a_mn^m+\cdots+a_1n+a_0$ and $a_m>0$, then $f(n)=\Omega(n^m)$.
- (證明: 請自己證證看!)

Asymptotic Notation – Theta

- Definition [Theta]:
- $\Theta(g(n)) =$

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

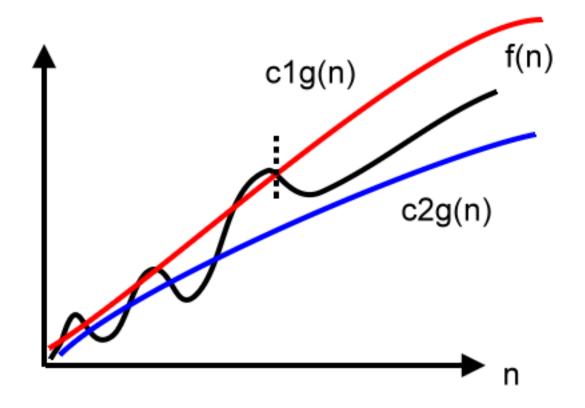
- $f(n) = \Theta(g(n))$
- "f of n is theta of g of n"
- 三明治夾心, 同時具有O(g(n)和 $\Omega(g(n))$
- (Asymptotically tight)



GMC Terrain

用圖表示

- Big Oh
- 紅色
- Omega
- 藍色
- Theta
- 紅色藍色都要



- $3n + 2 = \Theta(n)$
- since $3n + 2 \ge 3n$ for all $n \ge 2$ and $3n + 2 \le 4n$ for all $n \ge 2$.
- $3n + 3 = \Theta(n)$
- $10n^2 + 4n + 2 = \Theta(n^2)$
- $6 * 2^n + n^2 = \Theta(2^n)$
- $10 * \log n + 4 = \Theta(\log n)$
- $3n + 2 \neq \Theta(1)$
- $3n + 3 \neq \Theta(n^2)$
- $10n^2 + 4n + 2 \neq \Theta(n)$
- $10n^2 + 4n + 2 \neq \Theta(1)$
- $6 * 2^n + n^2 \neq \Theta(n^2)$
- $6 * 2^n + n^2 \neq \Theta(n^{100})$
- $6 * 2^n + n^2 \neq \Theta(1)$

Discussion

- More precise than both the "big oh" and omega notations
- It is true if and only g(n) is both an upper and lower bound on f(n).
- g(n) is an asymptotically tight bound for f(n) if $f(n)=\Theta(n)$
- 意思是說, f(n) equals g(n) within a constant factor (差常數倍而已).
- Theorem: If $f(n) = a_m n^m + \dots + a_1 n + a_0$ and $a_m > 0$, then $f(n) = \Theta(n^m)$.
- (證明: 請自己證證看!)

O(小歐)&ω(小歐美加)

- o(g(n)) = {f(n): for any positive constant c, there exists a constant n_0 such that $0 \le f(n) < cg(n)$ for all $n \ge n_0$ }
- 意思是g(n)是f(n)不緊的upper bound (長得比f(n)快)

• 也就是,
$$\lim_{n \to \infty} \frac{f(n)}{g(n)} = 0$$

- $\omega(g(n)) = \{f(n): \text{ for any positive constant } c, \text{ there exists a constant } n_0 \text{ such that } 0 \le cg(n) < f(n) \text{ for all } n \ge n_0 \}$
- 意思是g(n)是f(n)不緊的lower bound (長得比f(n)慢)
- 也就是, $\lim_{n\to\infty} \frac{f(n)}{g(n)} = \infty$

等式與不等式

- $n = O(n^2)$ (這個是什麼意思我們懂)
- $2n^2 + 3n + 1 = 2n^2 + \Theta(n)$ (????)
- 意思是 $2n^2 + 3n + 1 = 2n^2 + f(n)$, 然後 $f(n) = \Theta(n)$
- $2n^2 + \Theta(n) = \Theta(n^2)$ (???)
- 不管左邊asymptotic notation代表的function怎麼選擇,一定 有一種方法選擇右邊的asymptotic notation代表的function使 得等號成立.
- $2n^2 + 3n + 1 = 2n^2 + \Theta(n) = \Theta(n^2)$ (???)

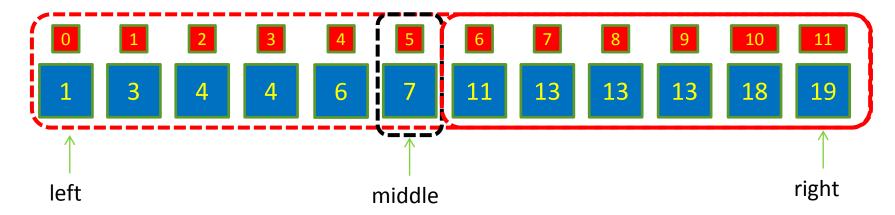
Another Example

```
int binsearch (int list[], int searchnum, int
left, int right) {
  int middle;
  while(left<=right) {</pre>
      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;
```

Worst case: $\Theta(\log n)$

Best case: $\Theta(1)$

searchnum=13;



middle=(left+right)/2;
left=middle+1;

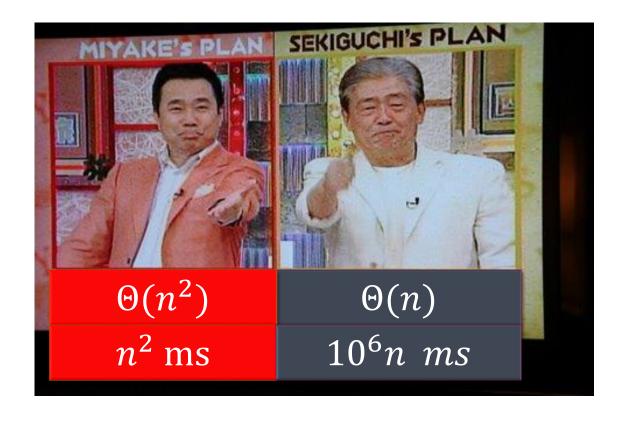
Half the searching window every iteration.

Another Example: recursive binsearch

```
int binsearch(int list[], int searchnum, int left, int right)
    int middle;
    if (left<=right) {</pre>
        middle=(left+right)/2;
         switch(COMPARE(list[middle], searchnum)) {
                  case -1:
                           return binsearch(list, searchnum, middle+1, right)
                  case 0:
                           return middle;
                  case 1:
                           return binsearch(list, searchnum, left, middle-1)
     else
                            T(n) = T\left(\frac{n}{2}\right) + O(1)
         return -1;
                                                    T(1) = O(1), T(0) = O(1)
                  T(n) = T\left(\frac{n}{2}\right) + c = \left(T\left(\frac{n}{4}\right) + c\right) + c = \dots = T(1) + c \ (n \log n)
```

 $T(n) = O(n \log n)$

演算法東西軍: 兜基??



哪一個好...

$$\Theta(n^2)$$
 $\Theta(n)$ $n^2 \text{ ms}$ $10^6 n \text{ ms}$

- n很大的時候右邊比左邊好
- · 但是n會很大嗎?
- 有時候不會
- · 如果n永遠小於106??
- · 結論:要看n大小(實際上寫程式的時候適用)

Reading Assignments

• Cormen 2.1, 2.2, 3.1