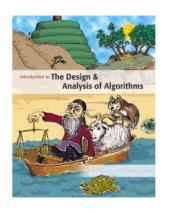




#### Introduction to

#### Algorithm Design and Analysis

[2] Asymptotics



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#### In the Last Class...

- Algorithm the spirit of computing
  - Model of computation
- Algorithm design and analysis
  - o Design
    - Correctness proof by induction
  - o Analysis
    - Worst-case / average-case complexity



### **Asymptotic Behavior**

- Asymptotic growth rate of functions
  - o Basic idea
- Key notations
  - $\circ$   $O, \Omega, \Theta$
  - $00,\omega$
- Brute force enumeration
  - o By iteration
  - o By recursion

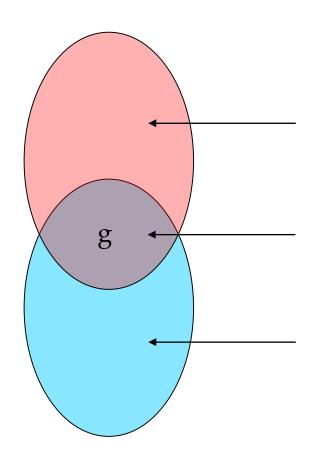


## How to Compare Two Algorithms

- Algorithm analysis, with simplifications
  - Measuring the cost by the number of critical operations
  - o Large input size only
  - o Essential part only
    - Only the leading term in f(n) is considered
    - Constant coefficients are ignored
- Capturing the essential part in the cost in a mathematical way
  - o Asymptotic growth rate of f(n)



#### Relative Growth Rate



 $\Omega$ (g):functions that grow at least as fast as g

 $\Theta$ (g):functions that grow at the same rate as g

O(g):functions that grow no faster as g

## "Big Oh"

- Basic idea  $f(n) \in O(g(n))$ 
  - For sufficiently large input size, g(n) is an upper bound for f(n)
- Definition " $\varepsilon N$ "
  - o Giving g: N→R<sup>+</sup>, then O(g) is the set of f:N→R<sup>+</sup>, such that for some c∈R<sup>+</sup> and some  $n_0$ ∈N, f(n)≤cg(n) for all n≥ $n_0$
- Definition " $lim_{n\to\infty}$ "

o 
$$f \in O(g)$$
 if  $\lim_{n \to \infty} \frac{f(n)}{g(n)} = c < \infty$ 

The limit may not exist, though it usually does.

## Example

• Let  $f(n)=n^2$ ,  $g(n)=n\log n$ , then:



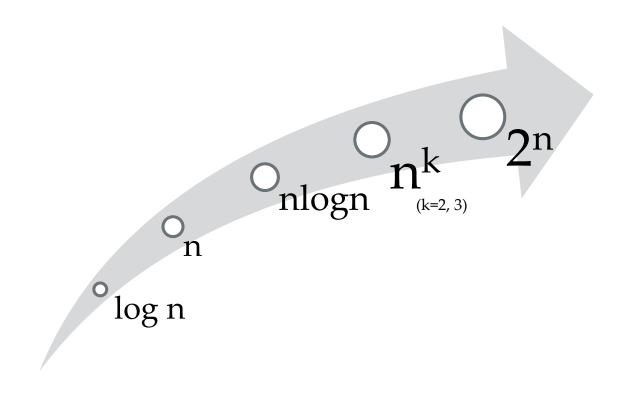
o f∉O(g), since

$$\lim_{n \to \infty} \frac{n^2}{n \log n} = \lim_{n \to \infty} \frac{n}{\log n} = \lim_{n \to \infty} \frac{1}{\frac{1}{n \ln 2}} = +\infty$$

o g∈O(f), since

$$\lim_{n \to \infty} \frac{n \log n}{n^2} = \lim_{n \to \infty} \frac{\log n}{n} = \lim_{n \to \infty} \frac{1}{n \ln 2} = 0$$

### **Asymptotic Growth Rate**





### **Asymptotic Order**

- Logarithm log n $log n \in O(n^{\alpha})$  for any  $\alpha > 0$
- Power  $n^k$

$$n^k \in O(c^n)$$
 for any  $c>1$ 

• Factorial *n!* 

$$n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$$
 (Stirling's formula)

## "Big $\Omega$ "

- Basic idea of  $f(n) \in \Omega(g(n))$ 
  - o Dual of "O"
- Definition " $\varepsilon N$ "
  - o Giving g: N→R<sup>+</sup>, then  $\Omega(g)$  is the set of f:N→R<sup>+</sup>, such that for some  $c \in R^+$  and some  $n_0 \in N$ ,  $f(n) \ge cg(n)$  for all  $n \ge n_0$
- Definition " $\lim_{n\to\infty}$ "
  - o  $f \in \Omega(g)$  if  $\lim_{n \to \infty} \frac{f(n)}{g(n)} = c > 0$  (the limit may be  $\infty$ )



#### 

- Basic idea of  $f(n) \in \Theta(g(n))$ 
  - o Roughly the same

$$\circ \Theta(g) = O(g) \cap \Omega(g)$$

- Definition " $\varepsilon N$ "
  - o Giving  $g: N \to R^+$ , then  $\Theta(g)$  is the set of  $f: N \to R^+$ , such that for some  $c_1, c_2 \in R^+$  and some  $n_0 \in N$ ,

$$0 \le c_1 g(n) \le f(n) \le c_2 g(n)$$
, for all  $n \ge n_0$ 

• Definition – " $\lim_{n\to\infty}$ "

o 
$$f \in \Theta(g)$$
 if  $\lim_{n \to \infty} \frac{f(n)}{g(n)} = c \ (0 < c < \infty)$ 



## Some Empirical Data

algorithm		1	2	3	4
Run time in <i>ns</i>		1.3 <i>n</i> <sup>3</sup>	10 <i>n</i> <sup>2</sup>	47nlogn	48 <i>n</i>
time for size	10 <sup>3</sup> 10 <sup>4</sup> 10 <sup>5</sup> 10 <sup>6</sup> 10 <sup>7</sup>	1.3s 22m 15d 41yrs 41mill	10ms 1s 1.7m 2.8hrs 1.7wks	0.4ms 6ms 78ms 0.94s 11s	0.05 <i>ms</i> 0.5 <i>ms</i> 5 <i>ms</i> 48ms 0.48s
max Size in time	sec min hr day	920 3,600 14,000 41,000	10,000 77,000 6.0×10 <sup>5</sup> 2.9×10 <sup>6</sup>	1.0×10 <sup>6</sup> 4.9×10 <sup>7</sup> 2.4×10 <sup>9</sup> 5.0×10 <sup>10</sup>	$ \begin{array}{c} 2.1 \times 10^{7} \\ 1.3 \times 10^{9} \\ 7.6 \times 10^{10} \\ 1.8 \times 10^{12} \end{array} $
time for 10 times size		×1000	×100	×10+	×10

on 400Mhz Pentium II, in C

from: Jon Bentley: *Programming Pearls* 



### Properties of O, $\Omega$ and $\Theta$

- Transitive property:
  - o If f∈O(g) and g∈O(h), then f∈O(h)
- Symmetric properties
  - o  $f \in O(g)$  if and only if  $g \in \Omega(f)$
  - $\circ f \in \Theta(g)$  if and only if  $g \in \Theta(f)$
- Order of sum function
  - $\circ O(f+g)=O(\max(f,g))$



#### "Little Oh"

- Basic idea of  $f(n) \in o(g(n))$ 
  - o Non-ignorable gap between f and its upper bound g
- Definition –" $\varepsilon N$ "
  - o Giving  $g:N \to R^+$ , then o(g) is the set of  $f:N \to R^+$ , such that for any c∈ $R^+$ , there exists some  $n_0 \in N$ ,

$$0 \le f(n) < cg(n)$$
, for all  $n \ge n_0$ 

• Definition – " $lim_{n\to\infty}$ "

o 
$$f \in o(g)$$
 if  $\lim_{n \to \infty} \frac{f(n)}{g(n)} = 0$ 



#### "Little ω"

- Basic idea of  $f(n) \in \omega(g(n))$ 
  - o Dual of "o"
- Definition " $\varepsilon N$ "
  - o Giving  $g:N \to R^+$ , then  $\omega(g)$  is the set of  $f:N \to R^+$ , such that for any c∈ $R^+$ , there exists some  $n_0 \in N$ ,

$$0 \le cg(n) < f(n)$$
, for all  $n \ge n_0$ 

• Definition – " $lim_{n\to\infty}$ "

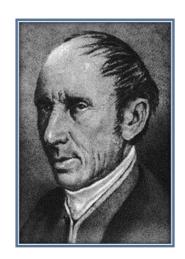
o 
$$f \in \omega(g)$$
 if  $\lim_{n \to \infty} \frac{f(n)}{g(n)} = \infty$ 

## Do You Know Infinity

Mathematical analysis

(differentiation / integration)

o Firm foundation



Cauchy

- How to talk about infinity?
  - o  $(\varepsilon N)$ -definition
  - $\circ (\varepsilon \delta)$ -definition



Weierstrass

# Brute Force Enumeration by Iteration

#### Swapping array elements

- o <time, space>
  - From  $<O(n^2)$ , O(1)>
  - To <O(n), O(n)>
  - To <O(n), O(1)>

#### Maximum subsequence sum

- o Time
  - From O(n<sup>3</sup>)
  - To O(n<sup>2</sup>)
  - To O(n log n)
  - To O(n)



## **Swapping Array Elements**

- E.g.,  $1,2,3,4 \mid 5,6,7 \Rightarrow 5,6,7,1,2,3,4$
- Brute force

Space sensitive

Time sensitive

	Time	Space
BF1	O(n²)	O(1)
BF2	O(n)	O(n)
Your Task	O(n)	O(1)

Your task

o Both time and space efficient

### Max-sum Subsequence

• The problem: Given a sequence *S* of integer, find the largest sum of a consecutive subsequence of *S*. (0, if all negative items)

```
o An example: -2, 11, -4, 13, -5, -2; the result 20: (11, -4, 13)
A brute-force algorithm:
                                                                    the sequence
MaxSum = 0;
 for (i = 0; i < N; i++)
  for (j = i; j < N; j++)
                                   i=0
   ThisSum = 0;
                                        i=1
   for (k = i; k \le j; k++)
   ThisSum += A[k];
                                            i=2
   if (ThisSum > MaxSum)
    MaxSum = ThisSum;
                                in O(n^3)
                                                                     i=n-1
 return MaxSum;
```



## **More Precise Complexity**

The total cost is: 
$$\sum_{i=0}^{n-1} \sum_{j=i}^{n-1} \sum_{k=i}^{j} 1$$

$$\sum_{i=1}^{j} 1 = j - i + 1$$

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

$$\sum_{i=0}^{n-1} \frac{(n-i+1)(n-i)}{2} = \sum_{i=1}^{n} \frac{(n-i+2)(n-i+1)}{2}$$

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

$$=\frac{n^3+3n^2+2n}{6}$$

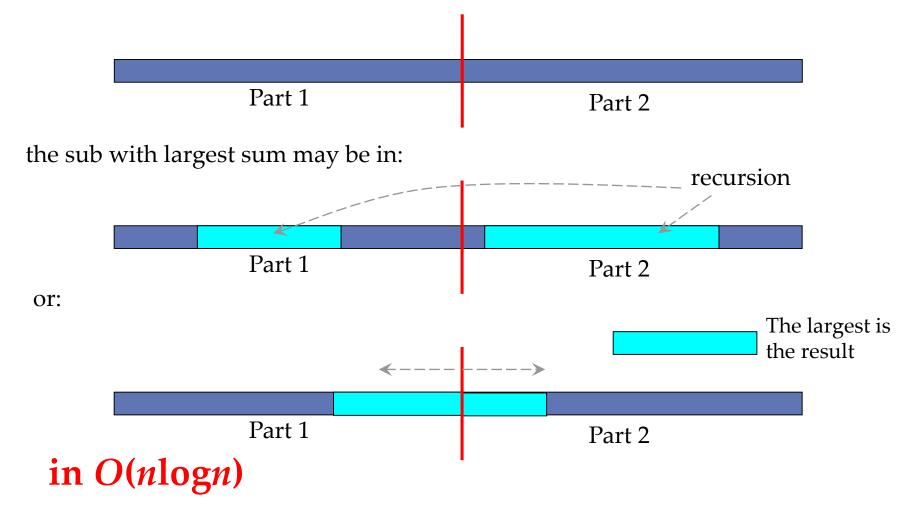


## Decreasing the Number of Loops

```
An improved algorithm
MaxSum = 0;
 for (i = 0; i < N; i++)
                                                             the sequence
  ThisSum = 0;
                              i=0
  for (j = i; j < N; j++)
                                   i=1
                                        i=2
   ThisSum += A[j];
   if (ThisSum > MaxSum)
    MaxSum = ThisSum;
                             in O(n^2)
                                                                     i=n-1
 return MaxSum;
```



## Power of Divide and Conquer





## Power of Divide and Conquer

```
Center = (Left + Right) / 2;
MaxLeftSum = MaxSubSum(A, Left, Center); MaxRightSum = MaxSubSum(A, Center + 1,
Right);
 MaxLeftBorderSum = 0; LeftBorderSum = 0;
for (i = Center; i >= Left; i--)
 LeftBorderSum += A[i];
 if (LeftBorderSum > MaxLeftBorderSum) MaxLeftBorderSum = LeftBorderSum;
                                                     Note: this is the core part of
 MaxRightBorderSum = 0; RightBorderSum = 0;
                                                     the procedure, with base case
for (i = Center + 1; i <= Right; i++)
                                                     and wrap omitted.
  RightBorderSum += A[i];
 if (RightBorderSum > MaxRightBorderSum) MaxRightBorderSum = RightBorderSum;
return Max3(MaxLeftSum, MaxRightSum,
     MaxLeftBorderSum + MaxRightBorderSum);
```



## A Linear Algorithm

First scan the array to eliminate the case of "all negative integers"

```
ThisSum = MaxSum = 0;
 for (j = 0; j < N; j++)
  ThisSum += A[i];
  if (ThisSum > MaxSum)
   MaxSum = ThisSum;
  else if (ThisSum < 0)
   ThisSum = 0;
```

the sequence



This is an example of "online algorithm"

return MaxSum;



Negative item or subsequence cannot be a prefix of the subsequence we want.

# Brute Force Enumeration by Recursion

#### Job scheduling

- o Problem definition
- o Brute force recursion
- o Further improvements

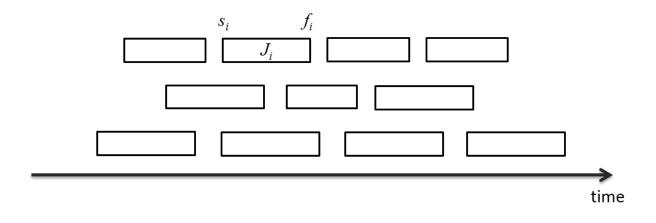
#### • Matrix chain multiplication

- o Problem definition
- Brute force recursion(s)
- o Further improvements



## Job Scheduling

- Jobs:  $J_i = [s_i, f_i]$
- Max number of compatible jobs





## Job Scheduling

#### Brute force recursion

- o Select job 'a'
- o Case 1: the result does not contain 'a'
  - Recursion on J\{a}
- o Case 2: the result contains 'a'
  - Recursion on J\{a}\{tasks overlapping with 'a'}

#### Further improvements

- o Dynamic programming (L16)
- o Greedy algorithms (L14)



## Matrix Chain Multiplication

#### • The task:

Find the product:  $A_1 \times A_2 \times ... \times A_{n-1} \times A_n$  $A_i$  is 2-dimentional array of different legal size

#### The Challenge:

- o Matrix multiplication is associative
- Different computing order results in great difference in the number of operations

#### The problem:

Which is the best computing order



## Cost of Matrix Multiplication

$$C_{i,j} = \sum_{k=1}^{q} a_{ik} b_{kj} \\ \begin{array}{c} \text{An example: } A_1 \times A_2 \times A_3 \times A_4 \\ 30 \times 1 & 1 \times 40 & 40 \times 10 & 10 \times 25 \\ ((A_1 \times A_2) \times A_3) \times A_4 : & 20700 \text{ multiplications} \\ A_1 \times (A_2 \times (A_3 \times A_4)) : & 11750 \\ (A_1 \times A_2) \times (A_3 \times A_4) : & 41200 \\ A_1 \times ((A_2 \times A_3) \times A_4) : & 1400 \\ \end{array}$$

C has  $p \times r$  elements as  $c_{i,j}$ 

So, pqr multiplications altogether



#### Solutions

- Brute force recursion (L16)
  - o BF1
  - o BF2

- Dynamic programming (L16)
  - o Based on brute force recursion 2



## Thank you!

Q & A

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