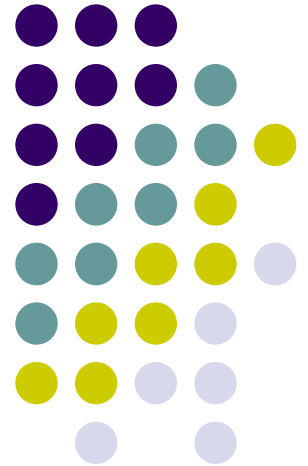


Analysis of Algorithms

Chapter 6.1, 6.5, 6.6





ROAD MAP

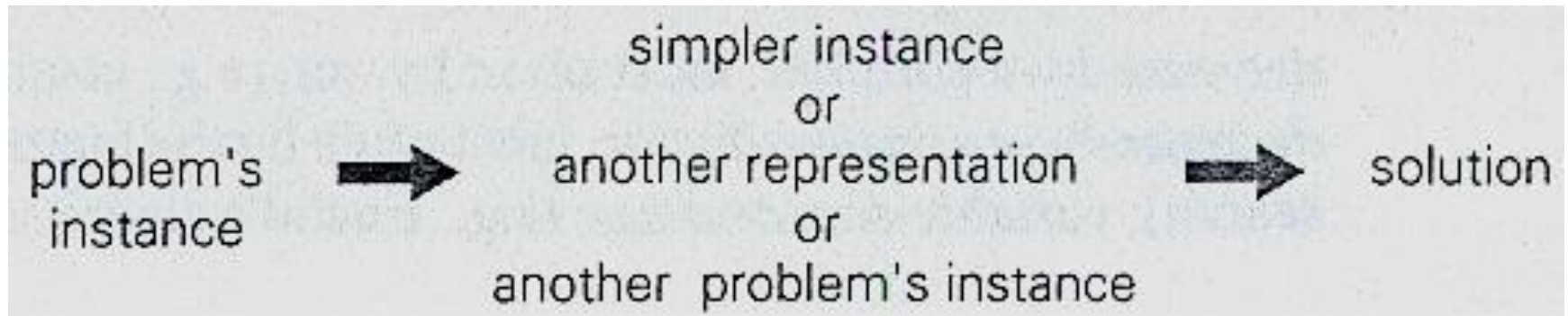
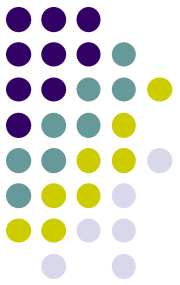
- **Transform And Conquer**
 - Instance simplification
 - Representation change
 - Problem reduction



Transform And Conquer

- *Transform and conquer technique* is based on idea of transformation
- This method works in two stages
 - Transformation stage
 - The problem is modified to another problem
 - more amenable to solution
 - Conquering stage
 - It is solved

Transform And Conquer Strategy



- Instance simplification
 - Transformation to a simpler instance problem
- Representation change
 - Transformation to a different representation of same instance
- Problem reduction
 - Transformation to an instance of a different problem for which an algorithm is already available



ROAD MAP

- Transform And Conquer
 - Instance simplification
 - Presorting
 - Element Uniqueness
 - Computing Mode
 - Searching
 - Representation change
 - Problem reduction



Presorting

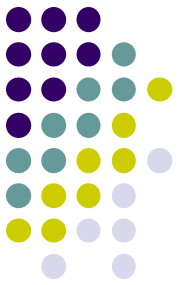
- Presorting is an old idea in computer science
- Many questions about a list are easier to answer if the list is sorted
- Efficiency of sorting algorithms is important
 - The benefits of a sorted list should more than the time spend for sorting.
 - Otherwise, use unsorted list directly
- We will assume that lists are implemented as *arrays*



Sorting

- We discussed three elementary sorting algorithms
 - Selection sort
 - Bubble sort
 - Insertion sort

These algorithms are *quadratic* in worst and average case
- Also discussed two advanced algorithms
 - Merge sort
 - $\Theta(n \log n)$ in worst and average case
 - Quick sort
 - $\Theta(n \log n)$ in average case
 - $\Theta(n^2)$ in worst case
- Are there faster algorithms ?
 - There is no general comparison-based sorting algorithm can have better efficiency than $\Theta(n \log n)$



Element Uniqueness

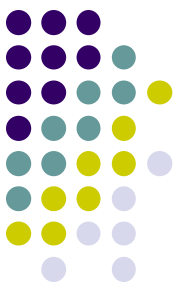
- Example 1 : *Checking element uniqueness in an array*
 - Brute force algorithm compare pairs of array's elements until either two equal elements were found or no pairs were left
 - Its worst case efficiency was $\Theta(n^2)$
 - Alternatively, what can we do ?



Element Uniqueness

- Approach :
 1. sort the array
 2. check only its consecutive elements

If the array has equal elements, a pair of them must be next to each other



Element Uniqueness

ALGORITHM *PresortElementUniqueness*($A[0..n - 1]$)

//Solves the element uniqueness problem by sorting the array first

//Input: An array $A[0..n - 1]$ of orderable elements

//Output: Returns “true” if A has no equal elements, “false” otherwise

Sort the array A

for $i \leftarrow 0$ to $n - 2$ do

 if $A[i] = A[i + 1]$ return false

return true

- What is the running time of the algorithm ?



Element Uniqueness

- Analysis :

$$T(n) = T_{sort}(n) + T_{scan}(n)$$

$$T(n) \in \Theta(n \log n) + \Theta(n)$$

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

More efficient than brute-force algorithm



Computing Mode

- Example 2 : Computing mode

A mode is value that occurs most often in a given list of numbers

For 5, 1, 5, 7, 6, 5, 7 the mode is 5

- In brute-force approach
 - Scan the list
 - Compute the frequencies of all distinct values
 - Find the value with largest frequency
- How to implement this idea?



Computing Mode

- Method:
 - Store values already encountered, along with their frequencies in a separate list
 - On each iteration, the i th element of original list is compared with values encountered
 - If a matching value is found, its frequency is incremented
 - Otherwise, current element is added to the list of distinct values seen so far with a frequency of 1

What about analysis?



Computing Mode

- Number of comparisons depends on the input.
 - In the best case: (all the elements are same)

$$C(n) \in \Theta(n)$$

- In worst case: (all the elements are different)

$$C(n) = \sum_{i=1}^n (i-1) = 0 + 1 + \dots + (n-1)$$

$$C(n) = \frac{n(n-1)}{2}$$

$$C(n) \in \Theta(n^2)$$

What can we do as an alternative ?



Computing Mode

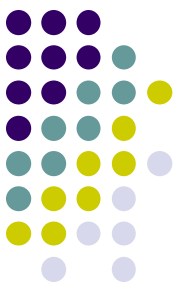
- Approach :

1. Sort the input

Then all equal values will be adjacent to each other

2. Find the longest run of adjacent equal values in the sorted array

Computing Mode



ALGORITHM *PresortMode*($A[0..n-1]$)

//Computes the mode of an array by sorting it first

//Input: An array $A[0..n-1]$ of orderable elements

//Output: The array's mode

Sort the array A

$i \leftarrow 0$ //current run begins at position i

$modefrequency \leftarrow 0$ //highest frequency seen so far

while $i \leq n-1$ **do**

$runlength \leftarrow 1$; $runvalue \leftarrow A[i]$

while $i+runlength \leq n-1$ **and** $A[i+runlength] = runvalue$

$runlength \leftarrow runlength+1$

if $runlength > modefrequency$

$modefrequency \leftarrow runlength$; $modevalue \leftarrow runvalue$

$i \leftarrow i+runlength$

return $modevalue$



Computing Mode

- Analysis :
 - Running time of algorithm depends on the time spent on sorting
 - remainder of the algorithm takes linear time (why ?)
 - So, with an $\Theta(n \log n)$ sort, worst case efficiency will be $\Theta(n \log n)$



Searching Problem

- Example 3 : Searching Problem
 - Searching for a given value v in a given array of n sortable items
 - Brute force solution is sequential search
 - needs n comparisons in worst case
 - If the array is sorted, we apply binary search
 - requires only $\lfloor \log_2 n \rfloor + 1$ comparisons in worst case



Searching Problem

- Assume the most efficient $\Theta(n \log n)$ sort is used
- Total running time in worst case and also average case will be

$$\begin{aligned} T(n) &= T_{\text{sort}}(n) + T_{\text{search}}(n) \\ &= \Theta(n \log n) + \Theta(\log n) = \Theta(n \log n) \end{aligned}$$

- Worst than sequential search!...
- What if the search will be done several times?...

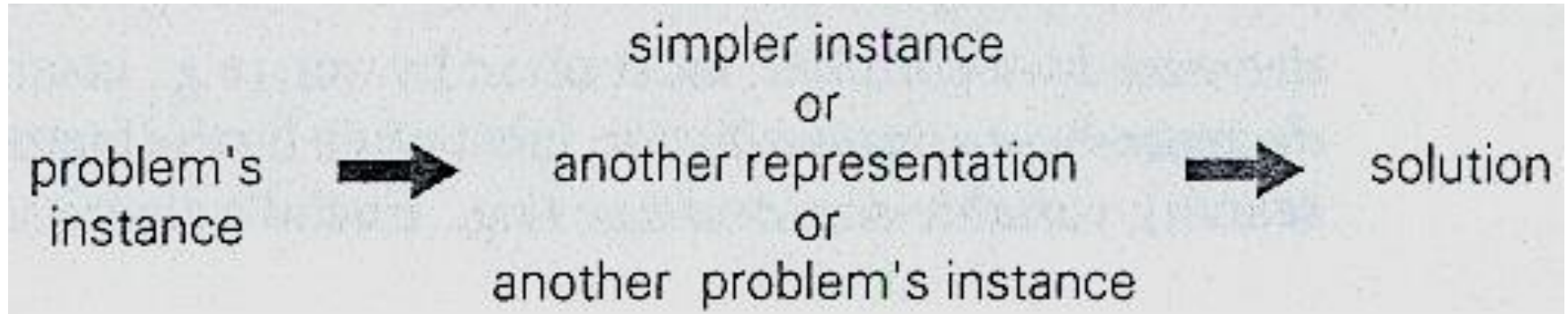
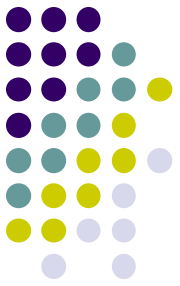


Presorting

Discussion:

- Geometric algorithms dealing with sets of points use presorting in one way or another
 - Presorting is used in divide and conquer for closest pair problem and convex-hull problem
- Some problems for directed acyclic graphs can be solved more easily after topologically sorting the digraph
 - Finding the shortest and longest paths

Transform And Conquer Strategy



- Instance simplification
 - Transformation to a simpler instance problem
- **Representation change**
 - Transformation to a different representation of same instance
- Problem reduction
 - Transformation to an instance of a different problem for which an algorithm is already available



ROAD MAP

- **Transform And Conquer**
 - Instance simplification
 - Representation change
 - **Horner's Rule and Binary Exponentiation**
 - Problem Reduction



Horner's Rule

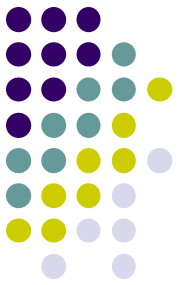
Problem Definition:

- Compute the value of a polynomial

$$p(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$$

at a given point x

- Polynomials constitute the most important class of functions
 - They possess a wealth of good properties
 - Can be used for approximating other types of functions
- Manipulating polynomials efficiently is an important problem



Horner's Rule

- Horner's rule provides elegant method for evaluating a polynomial
- It is a good example of representation change technique since it is based on representing $P(x)$ by a formula

$$p(x) = (... (a_n x + a_{n-1}) x + ...) x + a_0$$



Horner's Rule

- Example :

For example, for the polynomial

$$p(x) = 2x^4 - x^3 + 3x^2 + x - 5 \quad \text{we get}$$

$$\begin{aligned} p(x) &= 2x^4 - x^3 + 3x^2 + x - 5 \\ &= x(2x^3 - x^2 + 3x + 1) - 5 \\ &= x(x(2x^2 - x + 3) + 1) - 5 \\ &= x(x(x(2x - 1) + 3) + 1) - 5 \end{aligned}$$



Horner's Rule

- The pen-and-pencil calculation can be conveniently organized with a two row table
 - First row contains the polynomial's coefficients listed from the highest a_n to the lowest a_0
 - Second row is filled from left to right as follows (except its first entry which is a_n)
 - Next entry is computed as the x 's value times the last entry in the second row plus the next coefficient from first row
 - Final entry is the value being sought



Horner's Rule

EXAMPLE 1 Evaluate $p(x) = 2x^4 - x^3 + 3x^2 + x - 5$ at $x = 3$.

coefficients	2	-1	3	1	-5
$x = 3$	2	$3 \cdot 2 + (-1) = 5$	$3 \cdot 5 + 3 = 18$	$3 \cdot 18 + 1 = 55$	$3 \cdot 55 + (-5) = 160$

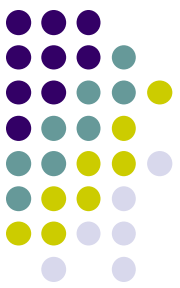
$$P(3) = 160$$

$$3 \cdot 2 + (-1) \rightarrow 2x - 1 \text{ at } x=3$$

$$3 \cdot 5 + 3 = 18 \rightarrow x(2x - 1) + 3 \text{ at } x=3$$

$$3 \cdot 18 + 1 = 55 \rightarrow x(x(2x - 1) + 3) + 1 \text{ at } x=3$$

$$3 \cdot 55 + (-5) = 160 \rightarrow x(x(x(2x - 1) + 3) + 1) - 5 = p(x)$$



Horner's Rule

ALGORITHM *Horner*($P[0..n]$, x)

//Evaluates a polynomial at a given point by Horner's rule

//Input: An array $P[0..n]$ of coefficients of a polynomial of degree n

// (stored from the lowest to the highest) and a number x

//Output: The value of the polynomial at x

$p \leftarrow P[n]$

for $i \leftarrow n - 1$ **downto** 0 **do**

$p \leftarrow x * p + P[i]$

return p



Horner's Rule

- Analysis :
 - Number of multiplications and number of additions

$$M(n) = A(n) = \sum_{i=0}^{n-1} 1 = n$$

- So how efficient is Horner's rule ?



Horner's Rule

- Analysis :
 - Consider only the first term of a polynomial of degree n : $a_n x^n$
 - Just computing this term with brute force approach requires n multiplications
 - Horner's rule computes $n-1$ other terms in addition to this and still uses the same number of multiplications
 - So it is an optimal algorithm for polynomial evaluation



Horner's Rule

- Discussion:

- Horner's rule also has some useful by-products
- The intermediate numbers generated by the algorithm in the process of evaluating $P(x)$ at some point x_0 turn out to be the coefficient to the quotient of the division of $P(x)$ by $x - x_0$

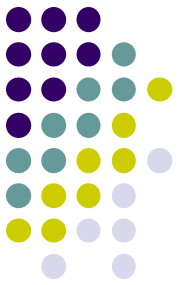
- While the final result, in addition to being $P(x_0)$ is equal to the remainder of this division of

$$P(x) = P'(x) (x - x_0) + P(x_0)$$

$$2x^4 - x^3 + 3x^2 + x - 5 \quad \text{by} \quad x - 3$$

$$2x^3 + 5x^2 + 18x + 55 \quad \text{and} \quad 160$$

- This division algorithm is known as *synthetic division*
 - It is more convenient than long division



Exponentiation

- Problem Definition :
 - Compute a^n
 - Computing a^n is an essential operation in *primality-testing* and *encryption methods*
 - The brute-force algorithm takes linear time
 - Designing other algorithms for computing a^n is important
 - For example, based on the representation change idea



Binary Exponentiation

- We will consider two algorithms for computing a^n
- Both of them exploit the binary representation of exponent n
 - One of them processes this binary string left to right
 - The second does it right to left



Binary Exponentiation

- Let

$$n = b_I \dots b_i \dots b_0$$

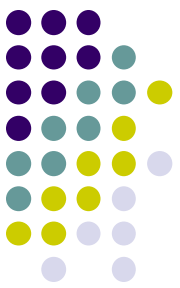
be the string representation of a positive integer n in binary system

The value of n can be computed as the value of polynomial at $x = 2$

$$P(x) = b_I x^I + \dots + b_i x^i + \dots + b_0$$

If $n = 13$ its binary representation is 1101 and

$$13 = 1.2^3 + 1.2^2 + 0.2^1 + 1.2^0$$



Binary Exponentiation

- If we compute the value of $P(x)$ with Horner's rule

$$a^n = a^{p(2)} = a^{b_I 2^I + \dots + b_i 2^i + \dots + b_0}$$

Horner's rule for the binary polynomial $p(2)$

$p \leftarrow 1$ //the leading digit is always 1 for $n \geq 1$
for $i \leftarrow I - 1$ **downto** 0 **do**
 $p \leftarrow 2p + b_i$

Implications for $a^n = a^{p(2)}$

$a^p \leftarrow a^1$
for $i \leftarrow I - 1$ **downto** 0 **do**
 $a^p \leftarrow a^{2p+b_i}$

$$a^{2p+b_i} = a^{2p} \cdot a^{b_i} = (a^p)^2 \cdot a^{b_i} = \begin{cases} (a^p)^2 & \text{if } b_i = 0 \\ (a^p)^2 \cdot a & \text{if } b_i = 1 \end{cases}$$



Binary Exponentiation

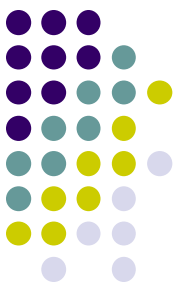
- After initializing the accumulator's value to a ,
 - the bit string representing the exponent is always square the last value of accumulator
 - if the current binary digit is 1 , also multiply it by a
- These observations lead to *left-to-right exponentiation* method of computing an



Left-to-right binary exponentiation

- Example :
 - Compute a^{13} by left-right binary exponentiation
 - Here $n = 13 = (1101)_2$
 - So

binary digits of n	1	1	0	1
product accumulator	a	$a^2 \cdot a = a^3$	$(a^3)^2 = a^6$	$(a^6)^2 \cdot a = a^{13}$



Left-to-right binary exponentiation

ALGORITHM *LeftRightBinaryExponentiation*($a, b(n)$)

//Computes a^n by the left-to-right binary exponentiation algorithm

//Input: A number a and a list $b(n)$ of binary digits b_1, \dots, b_0

// in the binary expansion of a positive integer n

//Output: The value of a^n

$product \leftarrow a$

for $i \leftarrow 1 - 1$ **downto** 0 **do**

$product \leftarrow product * product$

if $b_i = 1$ $product \leftarrow product * a$

return $product$



Left-to-right binary exponentiation

- Analysis :

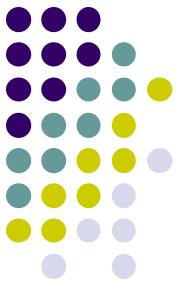
Total number of multiplications $M(n)$

$$I \leq M(n) \leq 2I$$

- $I + 1$ is the length of bit string representing exponent n
- $I = \lfloor \log_2 n \rfloor$

So efficiency is $\Theta(\log n)$

Left-to-right binary exponentiation



- Discussion :
 - This algorithm is better efficiency class than brute-force exponentiation
 - requires $n-1$ multiplications



Right-to-left binary exponentiation

- Definition:
 - Right-to-left binary exponentiation uses same binary polynomial $p(2)$ yielding value of n
 - But it does not apply Horner's rule
 - Exploits it differently

$$a^n = a^{b_I 2^I + \dots + b_i 2^i + \dots + b_0} = a^{b_I 2^I} \cdot \dots \cdot a^{b_i 2^i} \cdot \dots \cdot a^{b_0}$$



Right-to-left binary exponentiation

- Thus a^n can be computed as the product of terms

$$a^{b_i 2^i} = \begin{cases} a^{2^i} & \text{if } b_i = 1 \\ 1 & \text{if } b_i = 0 \end{cases}$$

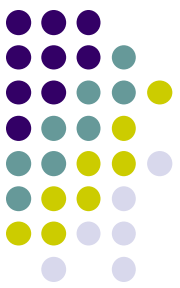
- The product of consecutive terms a^{2^i} , skipping those for which the binary digit b_i is zero
- We can compute a^{2^i} by simply squaring the same term we computed for the previous value of i since
$$a^{2^i} = (a^{2^{i-1}})^2$$
- We compute powers of a right to left (smallest to largest)



Right-to-left binary exponentiation

- Example :
 - Compute a^{13} by right-to-left binary exponentiation
 - Here $n = 13 = 1101$
 - So

1	1	0	1	binary digits of n
a^8	a^4	a^2	a	terms a^{2^i}
$a^5 \cdot a^8 = a^{13}$	$a \cdot a^4 = a^5$		a	product accumulator



Right-to-left binary exponentiation

ALGORITHM *RightLeftBinaryExponentiation*($a, b(n)$)

//Computes a^n by the right-to-left binary exponentiation algorithm

//Input: A number a and a list $b(n)$ of binary digits b_1, \dots, b_0

// in the binary expansion of a nonnegative integer n

//Output: The value of a^n

$term \leftarrow a$ //initializes a^{2^i}

if $b_0 = 1$ $product \leftarrow a$

else $product \leftarrow 1$

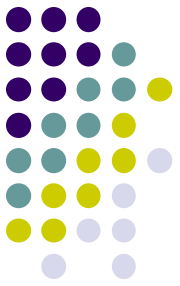
for $i \leftarrow 1$ **to** I **do**

$term \leftarrow term * term$

if $b_i = 1$ $product \leftarrow product * term$

return $product$

Right-to-left binary exponentiation



- Analysis :
 - Efficiency is *logarithmic*
 - Same as left-to-right binary multiplications



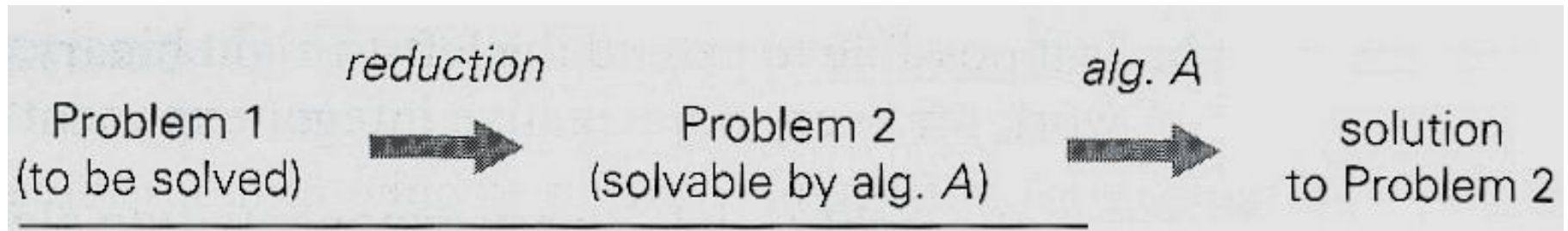
ROAD MAP

- **Transform And Conquer**
 - Instance simplification
 - Representation change
 - **Problem Reduction**
 - Computing The Least Common Multiple
 - Counting Paths in A Graph

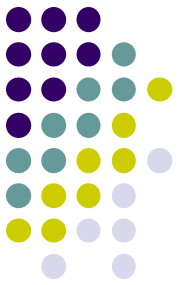


Problem Reduction

- **Definition:**
 - Problem reduction is to reduce a problem you need to solve to *another* problem that you know how to solve
 1. Find a problem to reduce onto
 2. Perform reduction



- The reduction worth if the reduction operations and algorithm A takes less time than solving the original problem directly



Least Common Multiple

- Definition :
 - Computing the least common multiple of two integers m and n is denoted $lcm(m,n)$
 - lcm is defined as the smallest integer that is divisible by both m and n
 - $lcm(24, 60) = 120$
 - $lcm(11,5) = 55$
 - It is an important notion in arithmetic and algebra



Computing the Least Common Multiple

- Approach :
 - Given the prime factorizations of m and n , $lcm(m, n)$ can be computed as the product of all the common prime factors of m and n times the product of m 's prime factors that are not in n times n 's prime factors that are not in m

$$24 = 2 . 2 . 2 . 3$$

$$60 = 2 . 2 . 3 . 5$$

$$lcm(24, 60) = (2 . 2 . 3) . 2 . 5 = 120$$



Computing the Least Common Multiple

- As a computational procedure, this algorithm has the same drawbacks as middle-school algorithm for computing greatest-common-divisor

How can we design a more efficient algorithm by using problem reduction ?



Computing the Least Common Multiple

- Product of $\text{lcm}(m,n)$ and $\text{gcd}(m,n)$ includes every factor of m and n exactly once
- So,

$$\text{lcm}(m, n) = \frac{m \cdot n}{\text{gcd}(m, n)}$$

- This formula reduces lcm calculation to gcd calculation
- $\text{gcd}(m,n)$ can be computed with Euclid's algorithm efficiently

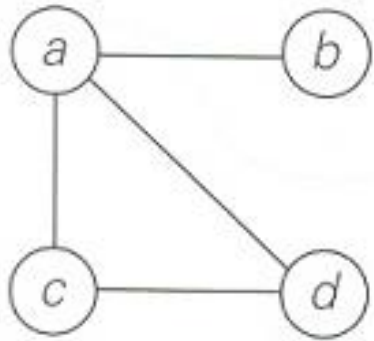


Counting Paths in a Graph

- Definition:
 - Counting different paths between two vertices in a graph
 - It is easy to prove that number of different paths of length $k > 0$ from the i th vertex to the j th vertex of a graph equals the (i,j) th element of A^k where A is the adjacency matrix of the graph



Counting Paths in a Graph



a graph

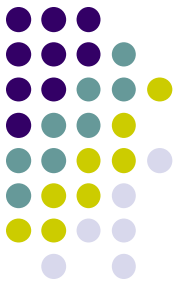
$$A = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix} \end{matrix}$$

its adjacency matrix A

$$A^2 = \begin{matrix} & \begin{matrix} a & b & c & d \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \end{matrix} & \begin{bmatrix} 3 & 0 & 1 & 1 \\ 0 & 1 & 1 & 1 \\ 1 & 1 & 2 & 1 \\ 1 & 1 & 1 & 2 \end{bmatrix} \end{matrix}$$

its square A^2

Elements of A and A^2 indicate the number of paths of lengths 1 and 2



Counting Paths in a Graph

- So, the problem can be solved with an algorithm for computing an appropriate power of its adjacency matrix
- Problem is reduced to matrix exponentiation
 - How to calculate A^k



Problem Reduction

- Discussion:
 - Plays a central role in theoretical computer science
 - where it is used to classify problems according to their complexity
 - The practical difficulty is finding a problem to which the problem at hand should be reduced
 - If we want our efforts to be of practical value, we need our reduction-based algorithm to be more efficient than solving the original problem directly