

CSC 411
Design and Analysis of Algorithms

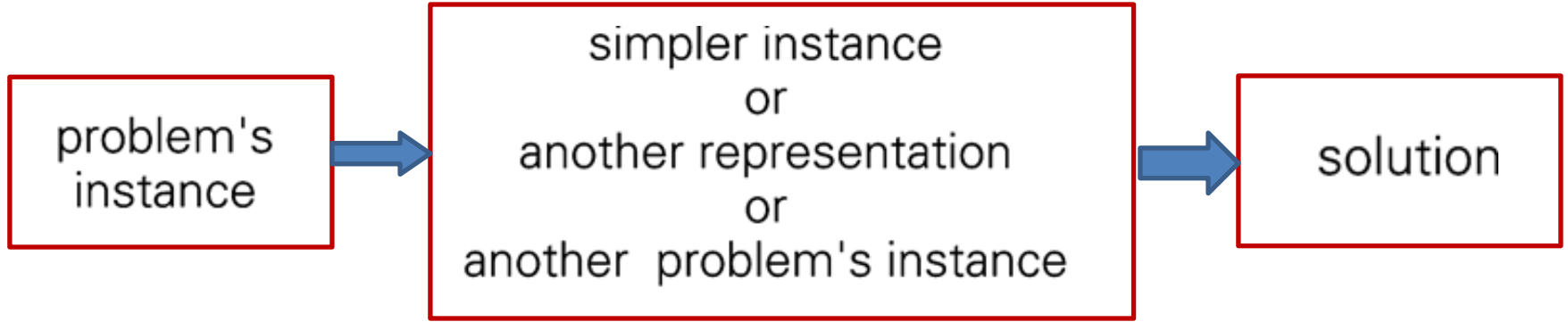
Chapter 6 Transfer and Conquer
- Part 1

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Transform and Conquer

- This group of techniques solves a problem by a ***transformation***
 - to a simpler/more convenient instance of the same problem (*instance simplification*)
 - to a different representation of the same instance (*representation change*)
 - to a different problem for which an algorithm is already available (*problem reduction*)

Transform and Conquer strategy



Transfer-and-Conquer Examples

- Presorting (6.1)
- Gaussian Elimination (6.2)
- Balanced Search Trees (6.3)
 - AVL Trees
 - 2-3 Trees
- Heaps and Heapsort (6.4)
- Horner's Rule and Binary Exponentiation (6.5)

6.1 Presorting

- Instance simplification
 - Solve a problem's instance by transforming it into another simpler or easier instance of the same problem
- Presorting
 - Many problems involving lists are easier when list is sorted.
 - searching
 - computing the median (selection problem)
 - checking if all elements are distinct (element uniqueness)
 - Also:
 - Topological sorting helps solving some problems for dags.
 - Presorting is used in many geometric algorithms.

How fast can we sort ?

- Efficiency of algorithms involving sorting depends on efficiency of sorting.
 - Selection sort, bubble sort, insertion sort:
 - $\Theta(n^2)$ in the worst and average cases.
 - Merge sort:
 - $\Theta(n \log n)$ always
 - Quick sort:
 - $\Theta(n \log n)$ in the average cases, $\Theta(n^2)$ in the worst
- No general comparison-based sorting algorithm can have a better efficiency than $\Theta(n \log n)$ in the worst case or average cases (Section 11.2)

Example 1: Finding Element Uniqueness

- Given a list A of n orderable elements, determine if there are any **duplicates** of any element

Brute Force:

```
for each  $x \in A$ 
  for each  $y \in \{A - x\}$ 
    if  $x = y$  return false
return true
```

Presorting:

```
Sort  $A$ 
for  $i \leftarrow 1$  to  $n-1$ 
  if  $A[i] = A[i+1]$  return false
return true
```

Runtime?

Efficiency of Element Uniqueness Algorithms

- Brute force algorithm
 - Compare all pairs of elements
 - Efficiency: $O(n^2)$
- Presorting-based algorithm
 - Stage 1: sort by efficient sorting algorithm (e.g. mergesort)
 - Stage 2: scan array to check pairs of adjacent elements
 - Efficiency: $\Theta(n \log n) + O(n) = \Theta(n \log n)$
- Another algorithm?
 - Hashing (do you own research if you are interested.)

Example 2: Searching with presorting

- Problem: Search for a given K in $A[0..n-1]$
- Presorting-based algorithm:
 - Stage 1 Sort the array by an efficient sorting algorithm
 - Stage 2 Apply binary search
- **Efficiency?**
 - $\Theta(n \log n) + O(\log n) = \Theta(n \log n)$
- Good or bad?
 - Why do we have our dictionaries, telephone directories, etc. sorted?
 - If we are to search in the same list more than once, the time spent on sorting is justified.

6.2 Gaussian Elimination

- **Given:** A system of n linear equations in n unknowns with an arbitrary coefficient matrix.
- **Transform to:**
 - An equivalent system of n linear equations in n unknowns with an upper triangular coefficient matrix.
 - An example of transform and conquer through **representation change**
- Backward substitution
 - Solve the latter by substitutions starting with the last equation and moving up to the first one.

Gaussian Elimination

- Consider a system of two linear equations:

$$a_{11}x + a_{12}y = b_1$$

$$a_{21}x + a_{22}y = b_2.$$

- To solve this we can rewrite the first equation to solve for x:

$$a_{11}x = (b_1 - a_{12}y) \rightarrow x = (b_1 - a_{12}y)/a_{11}$$

- And then substitute x in the second equation to solve for y.

After we solve for y, we can then solve for x:

$$a_{21} \cdot (b_1 - a_{12}y)/a_{11} + a_{22}y = b_2 \rightarrow y = ?$$

Gaussian Elimination

- In many applications we need to solve a system of n equations with n unknowns, e.g.:

$$a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2$$

$$\vdots$$

$$a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n = b_n$$

- If n is a large number it is very cumbersome to solve these equations using the substitution method.

Gaussian Elimination

- **Gaussian elimination**

- Fortunately there is a more elegant algorithm to solve such systems of linear equations
- The idea is to transform the system of linear equations into an equivalent one that **eliminates coefficients** so we end up with a triangular matrix.

$$\begin{array}{l} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2 \\ \vdots \\ a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n = b_n \end{array} \quad \Rightarrow \quad \begin{array}{l} a'_{11}x_1 + a'_{12}x_2 + \cdots + a'_{1n}x_n = b'_1 \\ a'_{22}x_2 + \cdots + a'_{2n}x_n = b'_2 \\ \vdots \\ a'_{nn}x_n = b'_n. \end{array}$$

Gaussian Elimination

$$\begin{array}{lcl} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1 & & a'_{11}x_1 + a'_{12}x_2 + \cdots + a'_{1n}x_n = b'_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2 & & a'_{22}x_2 + \cdots + a'_{2n}x_n = b'_2 \\ \vdots & \Rightarrow & \vdots \\ a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n = b_n & & a'_{nn}x_n = b'_n. \end{array}$$

- The matrix with zeros in the lower triangle (it is called an **upper triangular matrix**) is easier to solve.
- We can solve the last equation first, substitute into the second to last, etc. working our way back to the first one.

Pseudocode of Gaussian Elimination

ALGORITHM *ForwardElimination*($A[1..n, 1..n]$, $b[1..n]$)

//Applies Gaussian elimination to matrix A of a system's coefficients,

//augmented with vector b of the system's right-hand side values

//Input: Matrix $A[1..n, 1..n]$ and column-vector $b[1..n]$

//Output: An equivalent upper-triangular matrix in place of A with the

//corresponding right-hand side values in the $(n + 1)$ st column

for $i \leftarrow 1$ **to** n **do** $A[i, n + 1] \leftarrow b[i]$ //augments the matrix

for $i \leftarrow 1$ **to** $n - 1$ **do**

for $j \leftarrow i + 1$ **to** n **do**

for $k \leftarrow i$ **to** $n + 1$ **do**

$A[j, k] \leftarrow A[j, k] - A[i, k] * A[j, i] / A[i, i]$

Example: Gaussian Elimination

$$2x_1 - x_2 + x_3 = 1$$

$$4x_1 + x_2 - x_3 = 5$$

$$x_1 + x_2 + x_3 = 0.$$

$$\begin{bmatrix} 2 & -1 & 1 & 1 \\ 4 & 1 & -1 & 5 \\ 1 & 1 & 1 & 0 \end{bmatrix} \begin{array}{l} \\ \text{row 2} - \frac{4}{2} \text{ row 1} \\ \text{row 3} - \frac{1}{2} \text{ row 1} \end{array}$$

$$\begin{bmatrix} 2 & -1 & 1 & 1 \\ 0 & 3 & -3 & 3 \\ 0 & \frac{3}{2} & \frac{1}{2} & -\frac{1}{2} \end{bmatrix} \begin{array}{l} \\ \\ \text{row 3} - \frac{1}{2} \text{ row 2} \end{array}$$

$$\begin{bmatrix} 2 & -1 & 1 & 1 \\ 0 & 3 & -3 & 3 \\ 0 & 0 & 2 & -2 \end{bmatrix}$$

Efficiency of Gaussian Elimination

Efficiency: $\Theta(n^3) + \Theta(n^2) = \Theta(n^3)$

- Stage 1: Reduction to the upper-triangular matrix

$$\Theta(n^3)$$

- Stage 2: Backward substitution

$$\Theta(n^2)$$

LU Decomposition

$$2x_1 - x_2 + x_3 = 1$$

$$4x_1 + x_2 - x_3 = 5$$

$$x_1 + x_2 + x_3 = 0.$$

$$A = \begin{bmatrix} 2 & -1 & 1 \\ 4 & 1 & -1 \\ 1 & 1 & 1 \end{bmatrix}.$$

$$A = L \cdot U$$

$$Ax = b \rightarrow L \cdot Ux = b$$

$$U = \begin{bmatrix} 2 & -1 & 1 \\ 0 & 3 & -3 \\ 0 & 0 & 2 \end{bmatrix}$$

$$y = Ux \rightarrow Ly = b$$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ \frac{1}{2} & \frac{1}{2} & 1 \end{bmatrix}$$

Example: LU Decomposition

$$A = L \cdot U$$

$$Ax = b \rightarrow L \cdot Ux = b$$

$$y = Ux \rightarrow Ly = b$$

$$A = \begin{bmatrix} 2 & -1 & 1 \\ 4 & 1 & -1 \\ 1 & 1 & 1 \end{bmatrix}.$$

$$U = \begin{bmatrix} 2 & -1 & 1 \\ 0 & 3 & -3 \\ 0 & 0 & 2 \end{bmatrix}$$

$$L = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ \frac{1}{2} & \frac{1}{2} & 1 \end{bmatrix}$$

$$1. \quad Ly = b \quad \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ \frac{1}{2} & \frac{1}{2} & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 5 \\ 0 \end{bmatrix}$$

$$y_1 = 1, \quad y_2 = 5 - 2y_1 = 3, \quad y_3 = 0 - \frac{1}{2}y_1 - \frac{1}{2}y_2 = -2.$$

$$1. \quad y = Ux \quad \begin{bmatrix} 2 & -1 & 1 \\ 0 & 3 & -3 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \\ -2 \end{bmatrix},$$

$$x_3 = (-2)/2 = -1,$$

$$x_2 = (3 - (-3)x_3)/3 = 0, \quad x_1 = (1 - x_3 - (-1)x_2)/2 = 1.$$