

## CSC 8301- Design and Analysis of Algorithms

### Lecture 4 Brute Force, Exhaustive Search, Graph Traversal Algorithms

#### Brute-Force Approach

*Brute force* is a straightforward approach to solving a problem, usually directly based on the problem's statement and definitions of the concepts involved.

Example 1: Computing  $a^n$  ( $a > 0$ ,  $n$  is a positive integer)

Example 2: Searching for a given value in a list



- *Summation for  $C(n)$*

$$C(n) = \sum_{0 \leq i \leq n-1} \sum_{i+1 \leq j \leq n-1} 1$$

## **Closest-Pair Problem**

The closest-pair problem is to find the two closest points in a set of  $n$  points in the Cartesian plane (with the distance between two points measured by the standard Euclidean distance formula).

Brute-force algorithm for the closest-pair problem

Compute the distance between every pair of distinct points and return the indexes of the points for which the distance is the smallest.

## Closest-Pair Brute Force Algorithm (cont.)

**ALGORITHM** *BruteForceClosestPoints(P)*

//Input: A list  $P$  of  $n$  ( $n \geq 2$ ) points  $P_1 = (x_1, y_1), \dots, P_n = (x_n, y_n)$

//Output: Indices  $index1$  and  $index2$  of the closest pair of points

$dmin \leftarrow \infty$

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

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

$d \leftarrow \text{sqr}t((x_i - x_j)^2 + (y_i - y_j)^2)$  //sqr is the square root function

**if**  $d < dmin$

$dmin \leftarrow d; index1 \leftarrow i; index2 \leftarrow j$

**return**  $index1, index2$

- Time efficiency:
- Noteworthy improvement:

## General Notes on Brute-Force Approach

### Strengths:

- wide applicability
- simplicity
- yields reasonable algorithms for some important problems (e.g., sorting, searching, matrix multiplication)

### Weaknesses:

- yields efficient algorithms very rarely
- some brute-force algorithms are unacceptably slow
- not as constructive as some other design techniques

Note: Brute force can be a legitimate alternative in view of the human time vs. computer time costs

## Exhaustive search

*Exhaustive search* is a brute force approach to solving a problem that involves searching for an element with a special property, usually among *combinatorial objects* such as permutations, combinations, or subsets of a set.

Method:

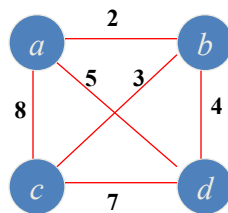
- systematically construct all potential solutions to the problem (often using standard algorithms for generating combinatorial objects such as those in Sec. 4.3)
- evaluate solutions one by one, disqualifying infeasible ones and, for optimization problems, keeping track of the best solution found so far
- when search ends, return the (best) solution found

### Example 1: Traveling salesman problem (TSP)

Given  $n$  cities with known distances between each pair, find the shortest tour that passes through all the cities exactly once before returning to the starting city.

Alternatively: Find shortest *Hamiltonian circuit* in a weighted connected graph.

Example:



## TSP by exhaustive search

Tour	Cost
a→b→c→d→a	2+3+7+5 = 17
a→b→d→c→a	2+4+7+8 = 21
a→c→b→d→a	8+3+4+5 = 20
a→c→d→b→a	8+7+4+2 = 21
a→d→b→c→a	5+4+3+8 = 20
a→d→c→b→a	5+7+3+2 = 17

Fewer tours?

Efficiency:

## Example 2: Knapsack Problem

Given  $n$  items with

weights:  $w_1 \ w_2 \ \dots \ w_n$

values:  $v_1 \ v_2 \ \dots \ v_n$

and a knapsack of capacity  $W$ ,

find most valuable subset of the items that fit into the knapsack

Example: Knapsack capacity  $W=16$

item	weight	value
1	2	\$20
2	5	\$30
3	10	\$50
4	5	\$10

## Knapsack by exhaustive search

Subset	Total weight	Total value	
{1}	2	\$20	
{2}	5	\$30	
{3}	10	\$50	
{4}	5	\$10	
{1,2}	7	\$50	
{1,3}	12	\$70	
{1,4}	7	\$30	
{2,3}	15	\$80	
{2,4}	10	\$40	
{3,4}	15	\$60	
{1,2,3}	<b>17</b>	not feasible	
{1,2,4}	12	\$60	
{1,3,4}	<b>17</b>	not feasible	
{2,3,4}	<b>20</b>	not feasible	
{1,2,3,4}	<b>22</b>	not feasible	

Efficiency:

## Comments on Exhaustive Search

- Typically, exhaustive search algorithms run in a realistic amount of time *only on very small instances*
- For some problems, there are much better alternatives
  - shortest paths
  - minimum spanning tree
  - assignment problem
- In many cases, exhaustive search (or variation) is the only known way to solve problem exactly for all its possible instances
  - TSP
  - knapsack problem

## Graph Traversal

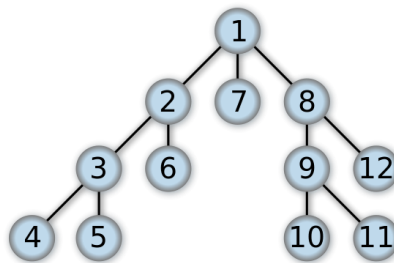
Many problems require processing all graph vertices (and edges) in systematic fashion, which can be considered “exhaustive search” algorithms

### Graph traversal algorithms:

- Depth-first search (DFS)
- Breadth-first search (BFS)

### Depth-First Search (DFS)

- Visits graph’s vertices by always moving away from last visited vertex to unvisited one, backtracks if no adjacent unvisited vertex is available.



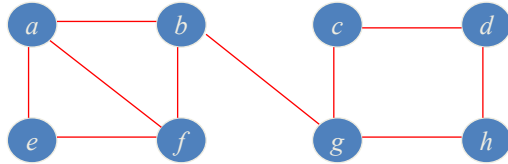
Source: Wikipedia,  
[https://en.wikipedia.org/wiki/Depth-first\\_search](https://en.wikipedia.org/wiki/Depth-first_search)

- “Redraws” graph in tree-like fashion (with tree edges and *back edges* for undirected graph)



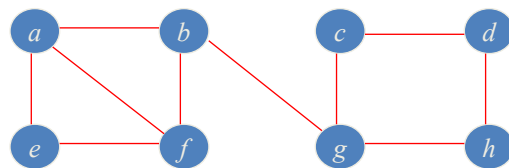
## Depth-First Search (DFS)

- Uses a stack
  - a vertex is pushed onto the stack first time is reached
  - a vertex is popped off the stack when it becomes a dead end, i.e., when there is no adjacent unvisited vertex



### Example: DFS traversal of undirected graph

- Assumption: ties broken alphabetically



**DFS traversal stack:**



**DFS forest:**

## Pseudocode of DFS

### ALGORITHM $DFS(G)$

```

//Implements a depth-first search traversal of a given graph
//Input: Graph  $G = \langle V, E \rangle$ 
//Output: Graph  $G$  with its vertices marked with consecutive integers
//in the order they've been first encountered by the DFS traversal
mark each vertex in  $V$  with 0 as a mark of being "unvisited"
count  $\leftarrow 0$ 
for each vertex  $v$  in  $V$  do
    if  $v$  is marked with 0
         $dfs(v)$ 

 $dfs(v)$ 
//visits recursively all the unvisited vertices connected to vertex  $v$  by a path
//and numbers them in the order they are encountered
//via global variable  $count$ 
count  $\leftarrow count + 1$ ; mark  $v$  with  $count$ 
for each vertex  $w$  in  $V$  adjacent to  $v$  do
    if  $w$  is marked with 0
         $dfs(w)$ 

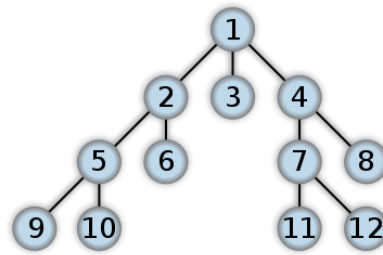
```

## Notes on DFS

- DFS can be implemented with graphs represented as:
  - adjacency matrices:  $\Theta(|V|^2)$
  - adjacency lists:  $\Theta(|V|+|E|)$
- Yields two distinct ordering of vertices:
  - order in which vertices are first encountered (pushed onto stack)
  - order in which vertices become dead-ends (popped off stack)
- Applications:
  - checking connectivity, finding connected components
  - checking acyclicity
  - searching state-space of problems for solution (AI)

## Breadth-first search (BFS)

- Visits graph vertices by moving across to all the neighbors of last visited vertex



Source: Wikipedia,  
[https://en.wikipedia.org/wiki/Breadth-first\\_search](https://en.wikipedia.org/wiki/Breadth-first_search)

- “Redraws” graph in tree-like fashion (with tree edges and *cross edges* for undirected graph)

## Pseudocode of BFS

### ALGORITHM *BFS*(*G*)

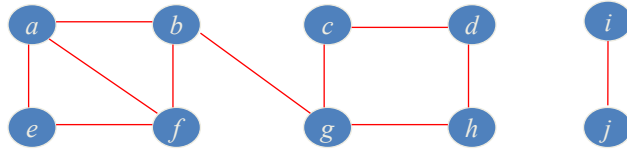
```

//Implements a breadth-first search traversal of a given graph
//Input: Graph  $G = \{V, E\}$ 
//Output: Graph  $G$  with its vertices marked with consecutive integers
//in the order they have been visited by the BFS traversal
mark each vertex in  $V$  with 0 as a mark of being “unvisited”
count ← 0
for each vertex  $v$  in  $V$  do
    if  $v$  is marked with 0
        bfs( $v$ )

bfs( $v$ )
//visits all the unvisited vertices connected to vertex  $v$  by a path
//and assigns them the numbers in the order they are visited
//via global variable count
count ← count + 1; mark  $v$  with count and initialize a queue with  $v$ 
while the queue is not empty do
    for each vertex  $w$  in  $V$  adjacent to the front vertex do
        if  $w$  is marked with 0
            count ← count + 1; mark  $w$  with count
            add  $w$  to the queue
    remove the front vertex from the queue
  
```

## Example of BFS traversal of undirected graph

- Instead of a stack, BFS uses a queue
- Assumption: neighbors visited in alphabetical order



**BFS traversal queue:**

**BFS forest:**

## Notes on BFS

- BFS has same efficiency as DFS and can be implemented with graphs represented as:
  - adjacency matrices:  $\Theta(|V|^2)$
  - adjacency lists:  $\Theta(|V|+|E|)$
- Yields single ordering of vertices (order added/deleted from queue is the same)
- Applications: same as DFS, but can also find paths from a vertex to all other vertices with the smallest number of edges

## Homework

Exercises 3.1: 4, 5, 8, 11

Exercises 3.4: 1, 5, 6

Exercises 3.5: 1, 2, 4, 6

Reading:

□ Sec. 3.1, 3.4 and 3.5

Next: Decrease-and-conquer (Chapter 4)