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6.006 Introduction to Algorithms
Spring 2008

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Lecture 12: Searching I: Graph Search and Representations

Lecture Overview: Search 1 of 3

- Graph Search
- Applications
- Graph Representations
- Introduction to breadth-first and depth-first search

Readings

CLRS 22.1-22.3, B.4

Graph Search

Explore a graph e.g., find a path from start vertices to a desired vertex

Recall: graph $G = (V, E)$

- V = set of vertices (arbitrary labels)
- E = set of edges i.e. vertex pairs (v, w)
 - ordered pair \implies *directed* edge of graph
 - unordered pair \implies *undirected*

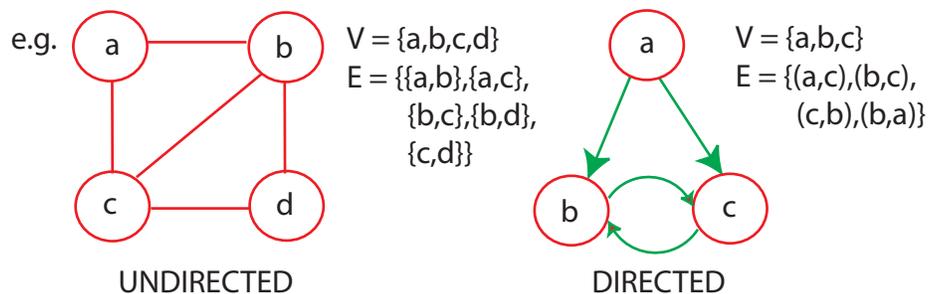


Figure 1: Example to illustrate graph terminology

Applications:

There are many.

- web crawling (How Google finds pages)
- social networking (Facebook friend finder)
- computer networks (Routing in the Internet)
shortest paths [next unit]
- solving puzzles and games
- checking mathematical conjectures

Pocket Cube:

Consider a $2 \times 2 \times 2$ Rubik's cube

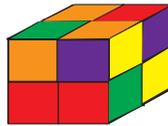


Figure 2: Rubik's Cube

- **Configuration Graph:**
 - vertex for each possible state
 - edge for each basic move (e.g., 90 degree turn) from one state to another
 - undirected: moves are reversible
- **Puzzle:** Given initial state s , find a path to the solved state
- $\#$ vertices = $8! \cdot 3^8 = 264,539,520$ (because there are 8 cubelets in arbitrary positions, and each cubelet has 3 possible twists)



Figure 3: Illustration of Symmetry

- can factor out 24-fold symmetry of cube: fix one cubelet

$$8^{11} \cdot 3 \implies 7! \cdot 3^7 = 11,022,480$$

- in fact, graph has 3 connected components of equal size \implies only need to search in one

$$\implies 7! \cdot 3^6 = 3,674,160$$

“Geography” of configuration graph

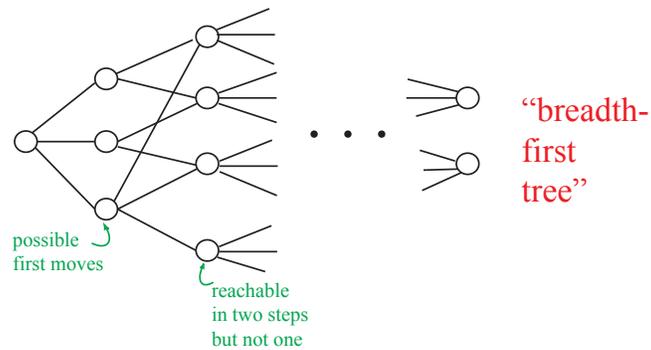


Figure 4: Breadth-First Tree

‡ reachable configurations

<u>distance</u>	<u>90° turns</u>	<u>90° & 180° turns</u>
0	1	1
1	6	9
2	27	54
3	120	321
4	534	1,847
5	2,256	9,992
6	8,969	50,136
7	33,058	227,536
8	114,149	870,072
9	360,508	1,887,748
10	930,588	623,800
11	1,350,852	2,644 ← diameter
12	782,536	
13	90,280	
14	276 ← diameter	
	3,674,160	3,674,160

Wikipedia Pocket Cube

Cf. $3 \times 3 \times 3$ Rubik’s cube: ≈ 1.4 trillion states; diameter is unknown! ≤ 26

Representing Graphs: (data structures)

Adjacency lists:

Array Adj of $|V|$ linked lists

- for each vertex $u \in V$, $Adj[u]$ stores u 's neighbors, i.e., $\{v \in V \mid (u, v) \in E\}$. $\text{colorBlue}(u, v)$ are just outgoing edges if directed. (See Fig. 5 for an example)
- in Python: $Adj =$ dictionary of list/set values vertex = any hashable object (e.g., int, tuple)
- advantage: multiple graphs on same vertices

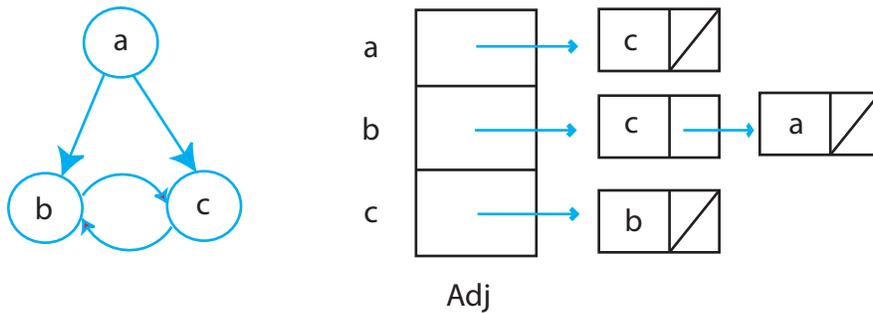


Figure 5: Adjacency List Representation

Object-oriented variations:

- object for each vertex u
- $u.\text{neighbors} =$ list of neighbors i.e., $Adj[u]$

Incidence Lists:

- can also make edges objects (see Figure 6)
- $u.\text{edges} =$ list of (outgoing) edges from u .
- advantage: storing data with vertices and edges without hashing

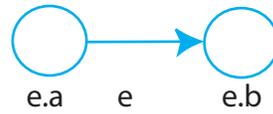


Figure 6: Edge Representation

Representing Graphs: contd.

The above representations are good for sparse graphs where $|E| \ll (|V|)^2$. This translates to a space requirement $= \Theta(V + E)$ (Don't bother with $|\cdot|$'s inside O/Θ).

Adjacency Matrix:

- assume $V = \{1, 2, \dots, |v|\}$ (number vertices)
- $A = (a_{ij}) = |V| \times |V|$ matrix where $i = \text{row}$ and $j = \text{column}$, and

$$a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E \\ \phi & \text{otherwise} \end{cases}$$

See Figure 7.

- good for dense graphs where $|E| \approx (|V|)^2$
- space requirement $= \Theta(V^2)$
- cool properties like A^2 gives length-2 paths and Google PageRank $\approx A^\infty$
- but we'll rarely use it **Google couldn't**; $|V| \approx 20 \text{ billion} \implies (|V|)^2 \approx 4.10^{20}$ [50,000 petabytes]

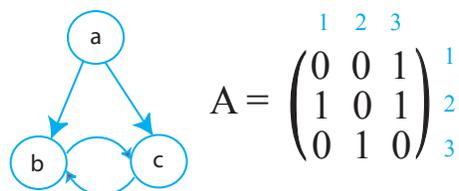


Figure 7: Matrix Representation

Implicit Graphs:

$\text{Adj}(u)$ is a function or $u.\text{neighbors}/\text{edges}$ is a method \implies “no space” (just what you need now)

High level overview of next two lectures:**Breadth-first search**

Levels like “geography”

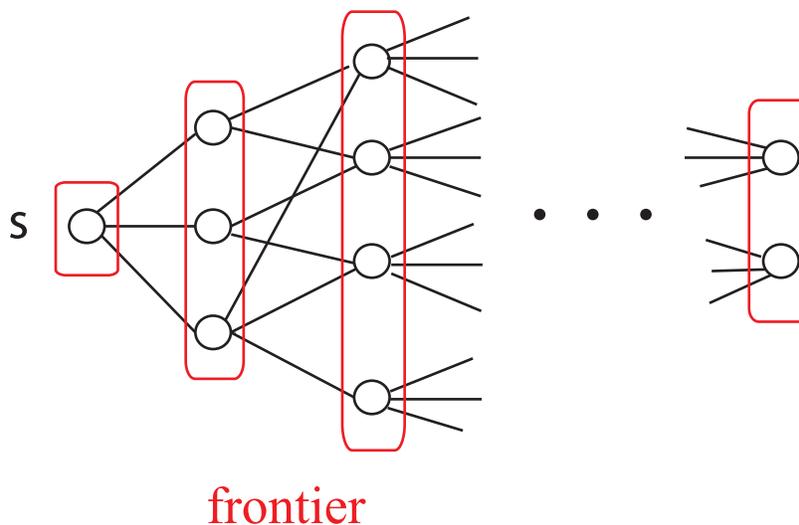


Figure 8: Illustrating Breadth-First Search

- frontier = current level
- initially $\{s\}$
- repeatedly advance frontier to next level, careful not to go backwards to previous level
- actually find shortest paths i.e. fewest possible edges

Depth-first search

This is like exploring a maze.

- e.g.: (left-hand rule) - See Figure 9
- follow path until you get stuck
- backtrack along breadcrumbs until you reach an unexplored edge

- recursively explore it
- careful not to repeat a vertex

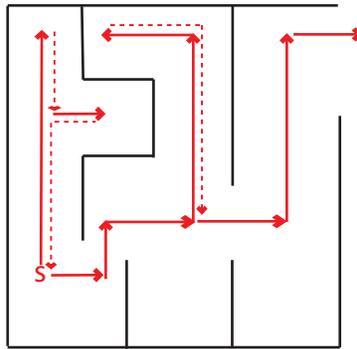


Figure 9: Illustrating Depth-First Search