

Outline: Search I (of 3)

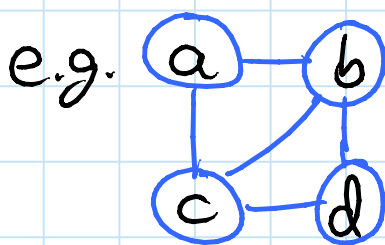
- graph search
- applications
- graph representations
- introduction to breadth-first & depth-first search

Reading: CLRS 22.1-22.3, B.4

Graph search: explore a graph
e.g. find a path from start vertex s
to a desired vertex

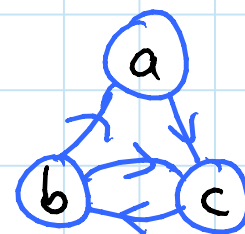
Recall: graph $G=(V, E)$

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



UNDIRECTED

$V = \{a, b, c, d\}$
 $E = \{\{a, b\}, \{a, c\},$
 $\{b, c\}, \{b, d\},$
 $\{c, d\}\}$



DIRECTED

$V = \{a, b, c\}$
 $E = \{(a, b),$
 $(a, c),$
 $(b, c), (c, b),$
 $(c, a)\}$

Applications: *many*

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

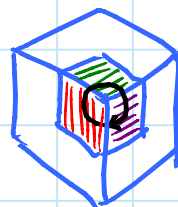
Pocket Cube: $2 \times 2 \times 2$ Rubik's cube



- configuration graph:
 - vertex for each possible state
 - edge for each basic move (e.g., 90° turn) from one state to another
 - undirected: moves are reversible
- puzzle: given initial state s_1 , find a path to the solved state
- #vertices = $8! \cdot 3^8 = 264,539,520$

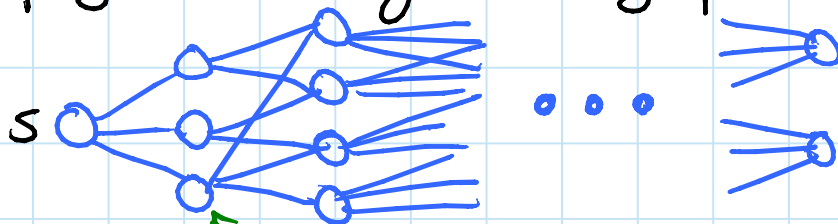
8 cubelet in
arbitrary positions

each cubelet
has 3 possible twists



- can factor out 24-fold symmetry of cube:
fix one cubelet $8! \cdot 3$
 $\Rightarrow 7! \cdot 3^7 = 11,022,480$
- in fact, graph has 3 connected components of equal size \Rightarrow only need to search in one
 $\Rightarrow 7! \cdot 3^6 = 3,674,160$

"Geography" of configuration graph:



"breadth-first tree"

possible first moves

reachable in two steps but not one

distance	# reachable configurations	
	90° turns	90° & 180° turns
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	
	<u>3,674,160</u>	<u>3,674,160</u>

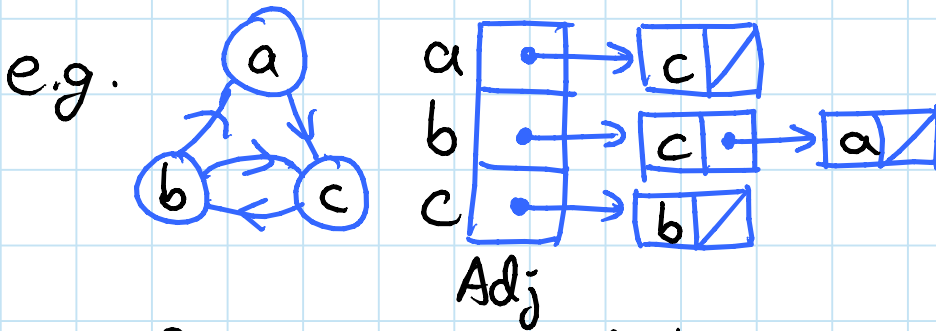
http://en.wikipedia.org/wiki/Pocket_Cube

Cf. 3x3x3 Rubik's cube:

≈ 1.4 trillion states
diameter 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. $\{v \in V \mid (u,v) \in E\}$
just outgoing edges if directed ↗



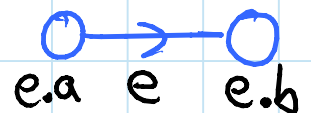
- in Python: Adj = dictionary of list/set values
vertex = any hashable object (e.g., int, tuple)
- advantage: multiple graphs on same vertices

Object-oriented variations:

- object for each vertex u
- $u.neighbors =$ list of neighbors i.e. Adj[u]

"Incidence lists:"

- can also make edges objects



- $u.edges =$ list of (outgoing) edges from u
- advantage: storing data with vertices & edges without hashing

Representing graphs: (cont'd)

above representations are good for sparse graphs where $|E| \ll |V|^2$ —
space requirement = $\Theta(V+E)$

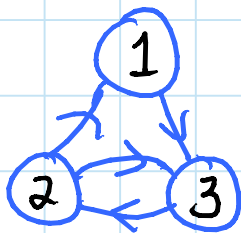
↖ ↗ don't bother with 1's inside O/Θ

Adjacency matrix:

- assume $V = \{1, 2, \dots, |V|\}$ (number vertices)
- $A = (a_{ij}) = |V| \times |V|$ matrix
where $a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E \\ \emptyset & \text{otherwise} \end{cases}$

i = row
 j = column

e.g.



$$A = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{pmatrix} \emptyset & \emptyset & 1 \\ 1 & \emptyset & 1 \\ \emptyset & 1 & \emptyset \end{pmatrix} \end{matrix} \begin{matrix} 1 \\ 2 \\ 3 \end{matrix}$$

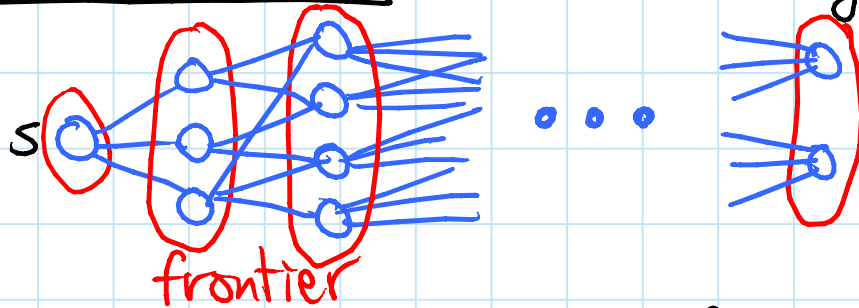
- good for dense graphs where $|E| \approx |V|^2$
- space requirement = $\Theta(V^2)$
- cool properties like A^2 gives length-2 paths & Google PageRank $\approx A^{100}$
- but we'll rarely use it

(Google couldn't: $|V| \approx 20$ billion $\Rightarrow |V|^2 \approx 4 \cdot 10^{20}$)
[50,000 petabytes]

Implicit graphs: $\text{Adj}(u)$ is a function
or $u.\text{neighbors/edges}$ is a method
 \Rightarrow "no space" (just what you need now)

High-level overview of next two lectures:

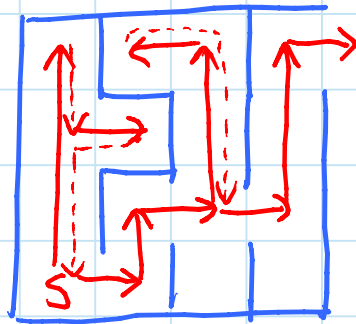
Breadth-first search: levels like "geography"



- 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: like exploring a maze

- e.g: (left-hand rule)



- follow path until you get stuck
- backtrack along breadcrumbs until you reach an unexplored edge
- recursively explore it
- careful not to repeat a vertex