Remembering the <u>edit distance</u> example we saw in class last time, we formulated the processing of the recursion as a table:

	ε	D	R	Ε	A	D
ε						
D						
Ε						
Ε						
D						

Is there an easier way to get the minimum cost alignment without having to calculate the value in each cell?

ECE-374-B: Lecture 14 - Graph search

Instructor: Abhishek Kumar Umrawal March 6, 2024

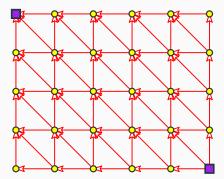
University of Illinois at Urbana-Champaign

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	ε	D	R	Ε	A	D
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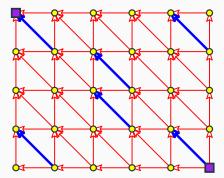
	ε	D	R	Ε	A	D
ε						
D						
Ε						
Ε						
D						



Look at the flow of the computation!

Remembering the edit distance example we saw in class last time, we formaluted the processing of the recursion as a table:

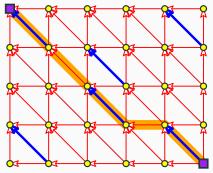
	ε	D	R	Ε	A	D
ε						
D						
Ε						
Ε						
D						



Look at the flow of the computation!

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	ε	D	R	E	A	D
ε						
D						
Ε						
Ε						
D						



We can solve the problem by turning it into a graph!



Why Graphs?

- Graphs help model networks which are ubiquitous:
 NP- tork
 transportation networks (rail, roads, airways), social networks
 (interpersonal relationships), information networks (web page links), and many problems that don't even look like graph problems.
- Fundamental objects in <u>Computer Science</u>, <u>Optimization</u>, <u>Combinatorics</u>.
- Many important and useful optimization problems are graph problems.
- Graph theory: elegant, fun and deep mathematics.

n: users k: budget

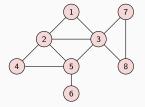
Social Infinence Maximization

"hest "subset of size "k"

Graph

An undirected (simple) graph G(V, E) is a 2-tuple:

- V is a set of vertices (also referred to as nodes/points)
- *E* is a set of edges where each edge $e \in E$ is a set of the form $\{u, v\}$ with $u, v \in V$ and $u \neq v$.



Example

In figure, G = (V, E) where $V = \{1, 2, 3, 4, 5, 6, 7, 8\}$ and $E = \{\{1, 2\}, \{1, 3\}, \{2, 3\}, \{2, 4\}, \{2, 5\}, \{3, 5\}, \{3, 7\}, \{3, 8\}, \{4, 5\}, \{5, 6\}, \{7, 8\}\}.$

State Space Search

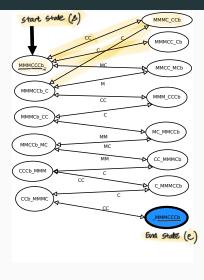
Many search problems can be modeled as search on a graph. The trick is figuring out what the vertices and edges are.

Missionaries and Cannibals

- (M) (c) (b) (_) • Three missionaries, three cannibals, one boat, one river
- Boat carries two people, must have at least one person
- Must all get across
- At no time can cannibals outnumber missionaries

How is this a graph search problem? What are the vertices? What are the edges?

Cannibals and Missionaries: Is the language empty?



Problems goes back to 800 CE

Versions with brothers and sisters.

Jealous Husbands.

Lions and buffalo

All bad names to a simple problem...

Find the shortest bath from s to e!

*Omitted states where cannibals out-

number missionaries

Problems on DFAs and NFAs sometimes are just problems on graphs

- M: DFA/NFA is L(M) empty? You can draw a DFA!
- *M*: DFA is $L(M) = \Sigma^*$?
- M: DFA, and a string w. Does M accepts w?
- N: NFA, and a string w. Does N accepts w?

Weighted Directed

Graph notation and representation

Notation

An edge in an undirected graphs is an <u>unordered pair</u> of nodes and hence it is a set. Conventionally we use uv for $\{u, v\}$ when it is clear from the context that the graph is undirected.

- u and v are the end points of an edge $\{u, v\}$
- Multi-graphs allow
 - <u>loops</u> which are edges with the same node appearing as both end points
 - multi-edges: different edges between same pairs of nodes
- In this class we will assume that a graph is a simple graph unless explicitly stated otherwise.

G1. no self loops 2. no multir edges

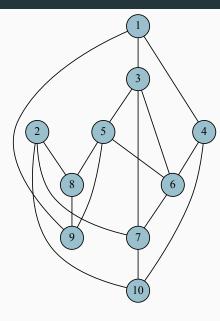
Adjacency Matrix

Represent G = (V, E) with *n* vertices and *m* edges using a $n \times n$ adjacency matrix *A* where

- A[i,j] = A[j,i] = 1 if $\{i,j\} \in E$ and A[i,j] = A[j,i] = 0 if $\{i,j\} \notin E$.
- Advantage: can check if $\{i, j\} \in E$ in O(1) time
- Disadvantage: needs $\Omega(n^2)$ space even when $\underline{m} \ll \underline{n}^2$

$$\frac{Max}{2} = \frac{n(n-1)}{2} = O(n^2)$$
(Handshake Lemma)

Graph adjacency matrix example [10 vertices]



	1	2	3	4	5	6	7	8	9	10
1	0	0	1	1	0	0	0	0	1	0
2	0	0	0	0	0	0	1	1	0	1
3	1	0	0	0	1	1	1	0	0	0
4	1	0	0	0	0	1	0	0	0	1
5	0	0	1	0	0	1	0	1	1	0
6	0	0	1	1	1	0	1	0	0	0
7	0	1	1	0	0	1	0	0	0	1
8	0	1	0	0	1	0	0	0	1	0
9	1	0	0	0	1	0	0	1	0	0
10	0	1	0	1	0	0	1	0	0	0

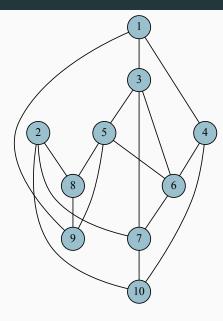
Adjacency Lists

Represent G = (V, E) with *n* vertices and *m* edges using adjacency lists:

- For each u ∈ V, Adj(u) = {v | {u, v} ∈ E}, that is neighbors of u. Sometimes Adj(u) is the list of edges incident to u.
- Advantage: space is O(m + n)
- Disadvantage: cannot "easily" determine in O(1) time whether $\{i, j\} \in E$
 - By sorting each list, one can achieve $O(\log n)$ time
 - By <u>hashing</u> "appropriately", one can achieve O(1) time

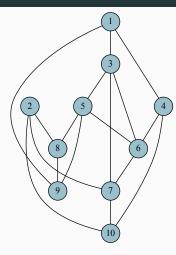
Note: In this class we will assume that by default, graphs are represented using plain vanilla (unsorted) adjacency lists.

Graph adjacency list example [10 vertices]



vertex	adjacency list
1	3, 4, 9
2	7, 8, 10
3	1, 5, 6, 7
4	1, 6, 10
5	3, 6, 8, 9
6	3, 4, 5, 7
7	2, 3, 6, 10
8	2, 5, 9
9	1, 5, 8
10	2, 4, 7

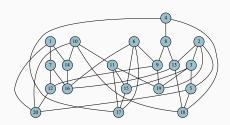
Graph adjacency matrix+list example [10 vertices]



vertex	adjacency list
1	3, 4, 9
2	7, 8, 10
3	1, 5, 6, 7
4	1, 6, 10
5	3, 6, 8, 9
6	3, 4, 5, 7
7	2, 3, 6, 10
8	2, 5, 9
9	1, 5, 8
10	2, 4, 7

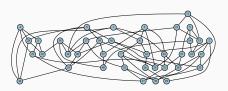
	1	2	3	4	5	6	7	8	9	10
1	0	0	1	1	0	0	0	0	1	0
2	0	0	0	0	0	0	1	1	0	1
3	1	0	0	0	1	1	1	0	0	0
4	1	0	0	0	0	1	0	0	0	1
5	0	0	1	0	0	1	0	1	1	0
6	0	0	1	1	1	0	1	0	0	0
7	0	1	1	0	0	1	0	0	0	1
8	0	1	0	0	1	0	0	0	1	0
9	1	0	0	0	1	0	0	1	0	0
10	0	1	0	1	0	0	1	0	0	0

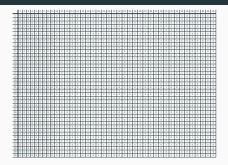
Graph adjacency matrix example [20 vertices]



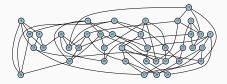
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
1	0	0	0	0	0	0	1	0	0	0	0	0	0	1	0	0	0	0	0	1
2	0	0	1	0	1	0	0	0	0	0	0	0	1	0	0	0	0	0	1	0
3	0	1	0	0	1	0	0	0	0	0	0	0	0	0	1	0	0	0	1	0
4	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	1	1	0	0
5	0	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	1
6	0	0	0	0	0	0	0	0	1	0	0	0	0	0	1	1	1	0	0	0
7	1	0	0	0	0	0	0	0	0	0	0	1	0	0	0	1	0	0	0	0
8	0	0	0	1	0	0	0	0	1	0	0	0	1	0	0	0	0	0	0	0
9	0	0	0	0	0	1	0	1	0	0	0	1	0	0	0	0	0	0	1	0
10	0	0	0	0	0	0	0	0	0	0	1	0	0	1	0	0	0	1	0	1
11	0	0	0	0	0	0	0	0	0	1	0	0	0	0	1	0	1	0	1	0
12	0	0	0	0	0	0	1	0	1	0	0	0	0	0	0	0	0	0	0	1
13	0	1	0	0	0	0	0	1	0	0	0	0	0	0	0	1	0	1	0	0
14	1	0	0	0	0	0	0	0	0	1	0	0	0	0	0	1	0	0	0	0
15	0	0	1	0	0	1	0	0	0	0	1	0	0	0	0	0	1	0	0	0
16	0	0	0	0	0	1	1	0	0	0	0	0	1	1	0	0	0	0	0	0
17	0	0	0	1	0	1	0	0	0	0	1	0	0	0	1	0	0	0	0	0
18	0	0	0	1	1	0	0	0	0	1	0	0	1	0	0	0	0	0	0	0
19	0	1	1	0	0	0	0	0	1	0	1	0	0	0	0	0	0	0	0	0
20	1	0	0	0	1	0	0	0	0	1	0	1	0	0	0	0	0	0	0	0

Graph adjacency matrix example [40 vertices]





Graph adjacency list example [40 vertices]

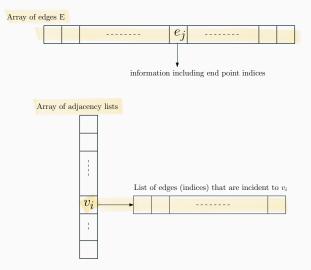


vertex	adjacency list
1	6, 24, 34, 36
2	12, 22, 23, 29
3	14. 15. 21
4	8, 19, 28, 36
5	6, 24, 25, 27
6	1, 5, 7, 23
7	6, 25, 32, 39
8	4, 19, 30
9	10, 16, 28, 35
10	9, 25, 27, 35
	13, 15, 33, 34
12	2, 33, 37, 38
13	11, 15, 17, 25
14	3, 22, 40
15	3, 11, 13, 22
16	9, 20, 23, 33
17	
18	20, 30, 34, 40
19	4, 8, 31, 37
20	16, 17, 18, 35
21	
22	2, 14, 15
	2, 6, 16, 26
24	1, 5, 31, 38
25	5, 7, 10, 13
26	
27	5, 10, 40 4, 9, 30, 36
28	4, 9, 30, 36
29	2, 26
	8, 18, 28
31	19, 21, 24, 37
32	7, 17, 37, 39
33	11, 12, 16, 39
34	1, 11, 17, 18
	9, 10, 20, 36
	1, 4, 28, 35
	12, 19, 31, 32
38	12, 21, 24, 39
39	7, 32, 33, 38
40	14, 18, 27

A Concrete Representation

- Assume vertices are numbered arbitrarily as $\{1, 2, \dots, n\}$.
- Edges are numbered arbitrarily as $\{1, 2, \dots, m\}$.
- Edges stored in an array/list of size *m*. *E*[*j*] is *j*th edge with info on end points which are integers in range 1 to *n*.
- Array Adj of size n for adjacency lists. Adj[i] points to adjacency list of vertex i. Adj[i] is a list of edge indices in range 1 to m.

A Concrete Representation



A Concrete Representation: Advantages

(RIY)

- Edges are explicitly represented/numbered. Scanning/processing all edges easy to do.
- Representation easily supports multigraphs including self-loops.
- Explicit numbering of vertices and edges allows use of arrays: O(1)-time operations are easy to understand.
- Can also implement via pointer based lists for certain dynamic graph settings.

Given a graph G = (V, E):

• path: sequence of distinct vertices v_1, v_2, \ldots, v_k such that $v_i v_{i+1} \in E$ for $1 \le i \le k-1$. The length of the path is k-1 (the number of edges in the path) and the path is from v_1 to v_k . Note: a single vertex u is a path of length 0.

$$p_1 : V_1 - V_2 - \cdots - V_{k-1} - V_k \qquad p_2: \bigvee_1$$

$$length(p_1) = k-1 \qquad length(p_2) = 0$$

Given a graph G = (V, E):

- <u>path</u>: sequence of <u>distinct</u> vertices v_1, v_2, \ldots, v_k such that $v_i v_{i+1} \in E$ for $1 \le i \le k-1$. The length of the path is k-1 (the number of edges in the path) and the path is from v_1 to v_k . <u>Note</u>: a single vertex u is a path of length 0.
- cycle: sequence of distinct vertices $v_1, v_2, ..., v_k$ such that $\{v_i, v_{i+1}\} \in E$ for $1 \le i \le k 1$ and $\{v_1, v_k\} \in E$. Single vertex not a cycle according to this definition. Caveat: Some times people use the term cycle to also allow vertices to be repeated; we will use the term tour.

Tour :

0

21

$$cycle: \quad v_1 - v_2 - \cdots - v_{k-1} - v_k$$

Given a graph G = (V, E):

- <u>path</u>: sequence of <u>distinct</u> vertices v_1, v_2, \ldots, v_k such that $v_i v_{i+1} \in E$ for $1 \le i \le k-1$. The length of the path is k-1 (the number of edges in the path) and the path is from v_1 to v_k . <u>Note</u>: a single vertex u is a path of length 0.
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 {v_i, v_{i+1}} ∈ E for 1 ≤ i ≤ k − 1 and {v₁, v_k} ∈ E. Single
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• A vertex u is connected to v if there is a path from u to v.

Given a graph G = (V, E):

- <u>path</u>: sequence of <u>distinct</u> vertices v_1, v_2, \ldots, v_k such that $v_i v_{i+1} \in E$ for $1 \le i \le k-1$. The length of the path is k-1 (the number of edges in the path) and the path is from v_1 to v_k . <u>Note</u>: a single vertex u is a path of length 0.
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 vertex not a cycle according to this definition.
 <u>Caveat</u>: Some times people use the term cycle to also allow
 vertices to be repeated; we will use the term tour.
- A vertex *u* is connected to *v* if there is a path from *u* to *v*.
- The connected component of <u>u</u>, con(<u>u</u>), is the set of all vertices connected to <u>u</u>. Is <u>u</u> ∈ con(<u>u</u>)? YES!

Connectivity contd

if UCV = True : n is Define a relation C on $V \times V$ as uCv if μ is connected to ν In undirected graphs, connectivity 9 is a reflexive, symmetric, and 3 transitive relation. Connected 5 8 components are the equivalence classes. Graph is connected if there is only \Rightarrow Con(1) = $\frac{1}{2}$ 1, 2, 3, 4, 5, 6, 7, 8 one connected component. $= Con(u) + u \in Con(1)$ C is reflexive: if UCU is True -> Equivalence relation symmetric: if ucv = vcu "transitive: if nCr, VCW then NCW Con(9) = 29,10} = Con(u) + u \in Con(9)

Algorithmic Problems

- Given graph G and nodes u and v, is u connected to v?
- Given G and node u, find all nodes that are connected to u.
- Find all connected components of G.

Algorithmic Problems

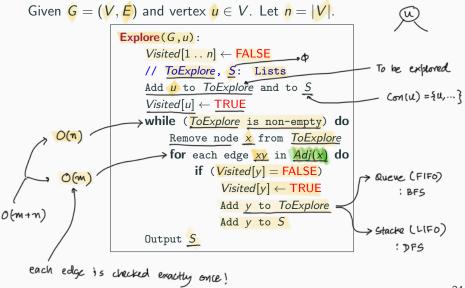
- Given graph G and nodes u and v, is u connected to v?
- Given G and node u, find all nodes that are connected to u.
- Find all connected components of G.

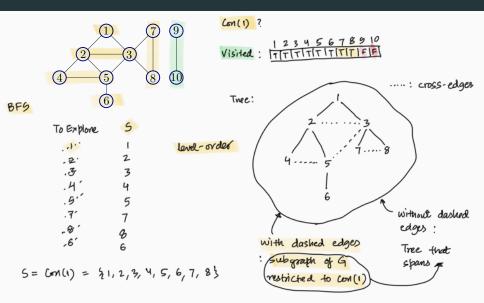
Can be accomplished in $\overline{O(m+n)}$ time using **BFS** or **DFS**. **BFS** and **DFS** are refinements of a basic search procedure which is good to understand on its own.

$$|E| = m$$
: $\#$ edges = $O(n^2)$ Define $m + n$: Size of the graph $|V| = n$: $\#$ vertices

Computing connected components in undirected graphs using basic graph search

Basic Graph Search in Undirected Graphs





Properties of Basic Search

Running Time: O(m + n)

Running Time: O(m + n)

BFS and DFS are special case of BasicSearch.

- Breadth First Search (BFS): use <u>queue</u> data structure to implementing the list *ToExplore*
- Depth First Search (DFS): use stack data structure to implement the list ToExplore

DIY

One can create a natural search tree T rooted at u during search.

```
Explore(G, u):
    array Visited[1..n]
    Initialize: Visited[i] \leftarrow FALSE for i = 1,...,n
    List: ToExplore, S
    Add u to ToExplore and to S, Visited[u] \leftarrow TRUE
    Make tree T with root as \mu
    while (ToExplore is non-empty) do
        Remove node x from ToExplore
        for each edge (x, y) in Adj(x) do
             if (Visited[y] = FALSE)
                  Visited[y] \leftarrow TRUE
                  Add y to ToExplore
                  Add v to S
                  Add y to T with x as its parent
    Output S
```

T is a spanning tree of con(u) rooted at u

Finding all connected components

· · -

Modify Basic Search to find all connected components of a given graph G in O(m + n) time.

While $\exists x \in V$ where visited (x) = = False

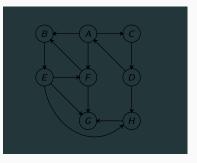
Directed Graphs and Directed Connectivity

Directed Graphs

Definition

A directed graph G = (V, E) consists of

- set of vertices/nodes V and
- a set of edges/arcs $E \subseteq V \times V.$



An edge is an <u>ordered</u> pair of vertices. (u, v) different from (v, u).

In many situations relationship between vertices is asymmetric:

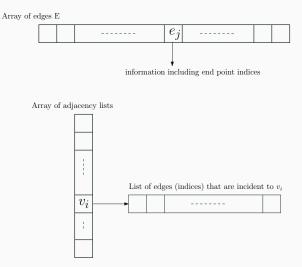
- Road networks with one-way streets.
- Web-link graph: vertices are web-pages and there is an edge from page p to page p' if p has a link to p'. Web graphs used by Google with PageRank algorithm to rank pages.
- Dependency graphs in variety of applications: link from x to y if y depends on x. Make files for compiling programs.
- Program Analysis: functions/procedures are vertices and there is an edge from x to y if x calls y.

Graph G = (V, E) with *n* vertices and *m* edges:

- Adjacency Matrix: n × n asymmetric matrix A. A[u, v] = 1 if (u, v) ∈ E and A[u, v] = 0 if (u, v) ∉ E. A[u, v] is not same as A[v, u].
- Adjacency Lists: for each node u, <u>Out(u)</u> (also referred to as Adj(u)) and <u>In(u)</u> store out-going edges and in-coming edges from u.

Default representation is adjacency lists.

Concrete representation discussed previously for undirected graphs easily extends to directed graphs.



Directed Connectivity

Given a graph G = (V, E):

A (directed) path is a sequence of distinct vertices
v₁, v₂,..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1. The length of the path is k − 1 and the path is from v₁ to v_k. By convention, a single node u is a path of length 0.

$$V_1 \rightarrow V_2 \rightarrow \cdots \rightarrow V_{k-1} \rightarrow V_k \qquad V_1$$

length: $k-1$ length: 0

Given a graph G = (V, E):

- A (directed) path is a sequence of distinct vertices v₁, v₂,..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1. The length of the path is k − 1 and the path is from v₁ to v_k. By convention, a single node u is a path of length 0.
- A cycle is a sequence of distinct vertices v₁, v₂,..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1 and (v_k, v₁) ∈ E. By convention, a single node u is not a cycle.

$$V_1 \rightarrow V_2 \rightarrow \cdots \rightarrow V_{K-1} \rightarrow V_K$$

Given a graph G = (V, E):

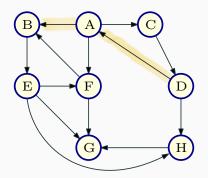
- A (directed) path is a sequence of distinct vertices v₁, v₂,..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1. The length of the path is k − 1 and the path is from v₁ to v_k. By convention, a single node u is a path of length 0.
- A cycle is a sequence of distinct vertices v₁, v₂,..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1 and (v_k, v₁) ∈ E. By convention, a single node u is not a cycle.
- A vertex u can reach v if there is a path from u to v.
 Alternatively v can be reached from u.

Given a graph G = (V, E):

- A (directed) path is a sequence of distinct vertices v₁, v₂,..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1. The length of the path is k − 1 and the path is from v₁ to v_k. By convention, a single node u is a path of length 0.
- A cycle is a sequence of distinct vertices v₁, v₂,..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1 and (v_k, v₁) ∈ E. By convention, a single node u is not a cycle.
- A vertex *u* can reach *v* if there is a path from *u* to *v*. Alternatively *v* can be reached from *u*.
- Let rch(u) be the set of all vertices reachable from u.

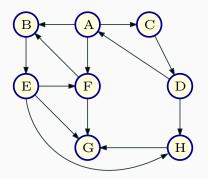
Connectivity contd

Asymmetricity: *D* can reach *B* but *B* cannot reach *D*



Connectivity contd

Asymmetricity: D can reach B but B cannot reach D



Questions:

- Is there a notion of connected components?
- How do we understand connectivity in directed graphs?

Strong connected components

Given a directed graph G, u is strongly connected to v if u can reach v and v can reach u. In other words $v \in \operatorname{rch}(u)$ and $u \in \operatorname{rch}(v)$.

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Define relation C where uCv if u is (strongly) connected to v.

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Define relation \underline{C} where uCv if u is (strongly) connected to v.

Proposition

C is an equivalence relation, that is <u>reflexive</u>, <u>symmetric</u> and <u>transitive</u>.

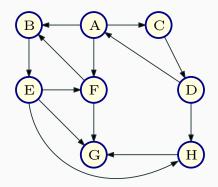
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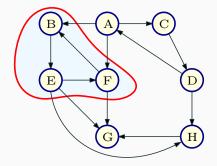
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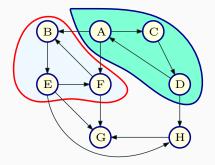
Proposition

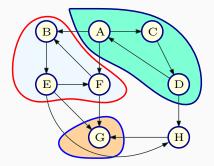
C is an equivalence relation, that is reflexive, symmetric and transitive.

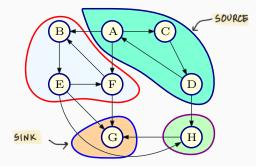
Equivalence classes of C: strong connected components of G. They partition the vertices of G. SCC(u): strongly connected component containing u.











4 strongly connected components 2 B, E, F}, 2 A, C, D3, 263, 2H3

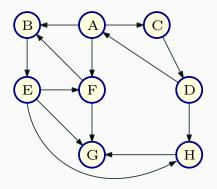
Directed Graph Connectivity Problems

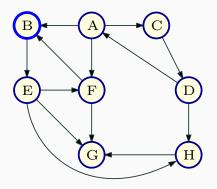
- Given G and nodes u and v, can u reach v?
- Given *G* and *u*, compute rch(*u*).
- Given G and u, compute all v that can reach u, that is all v such that u ∈ rch(v).
- Find the strongly connected component containing node u, that is SCC(u).
- Is *G* strongly connected (a single strong component)?
- Compute all strongly connected components of G.

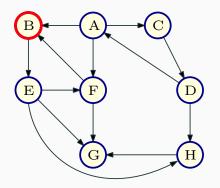
Graph exploration in directed graphs

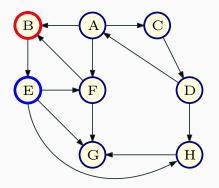
Basic Graph Search in Directed Graphs

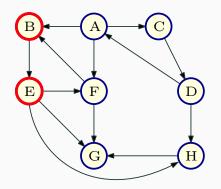
```
Given G = (V, E) a directed graph and vertex u \in V. Let n = |V|.
        Explore(G, u):
             array Visited[1..n]
             Initialize: Set Visited[i] \leftarrow FALSE for 1 \le i \le n
             List: ToExplore, S
             Add u to ToExplore and to S, Visited[u] \leftarrow TRUE
             Make tree T with root as \mu
             while (ToExplore is non-empty) do
                 Remove node x from ToExplore Out(x)
                 for each edge (x, y) in Adj(x) do
                      if (Visited[y] = FALSE)
                          Visited[y] \leftarrow TRUE
                          Add y to ToExplore
                          Add y to S
                          Add y to T with edge (x, y)
             Output S
```

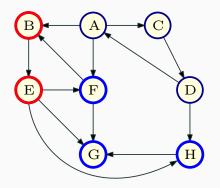


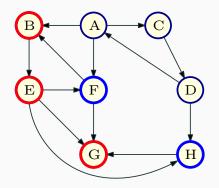


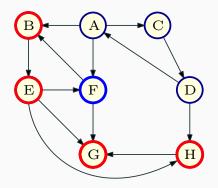


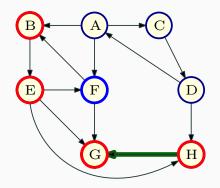


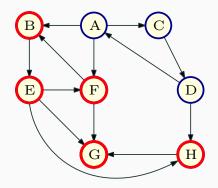


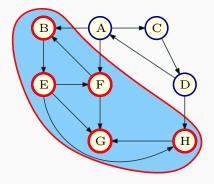












 $TCh(B) = 2B, E, F, G, H_{3}^{2}$

Properties of Basic Search

Proposition Explore(G, u) terminates with $S = \operatorname{rch}(u)$.

Proof Sketch.

- Once *Visited*[*i*] is set to *TRUE* it never changes. Hence a node is added only once to *ToExplore*. Thus algorithm terminates in at most *n* iterations of while loop.
- By induction on iterations, can show $v \in S \Rightarrow v \in \operatorname{rch}(u)$
- Since each node v ∈ S was in ToExplore and was explored, no edges in G leave S. Hence no node in V − S is in rch(u).
 <u>Caveat:</u> In directed graphs edges can enter S.
- Thus $S = \operatorname{rch}(u)$ at termination.

RIY

Directed Graph Connectivity Problems

- Given G and nodes u and v, can u reach v? if $v \in \underline{rch}(w)$: r_{u} !
- Given G and u, compute $\frac{\operatorname{rch}(u)}{u}$. Basic Search
- Given G and u, compute all \underline{v} that can reach \underline{u} , that is all \underline{v} such that $\underline{u} \in \operatorname{rch}(v)$. Naive: $O(\underline{n} \cdot (\underline{m+n}))$ Better: $O(\underline{m+n}) \checkmark$
- Find the strongly connected component containing node <u>u</u>, that is <u>SCC(u</u>). O(m+n) ✓
- Is <u>G</u> strongly connected (a single strong component)? O(m+n) ~
- Compute all strongly connected components of G.

Basic Seanch O(m+n)

Directed Graph Connectivity Problems

- Given G and nodes u and v, can u reach v?
- Given G and u, compute rch(u).
- Given G and u, compute all v that can reach u, that is all v such that u ∈ rch(v).
- Find the strongly connected component containing node *u*, that is *SCC*(*u*).
- Is G strongly connected (a single strong component)?
- Compute all strongly connected components of *G*.

First five problems can be solved in O(n + m) time by via Basic Search (or **BFS/DFS**). The last one can also be done in linear time but requires a rather clever **DFS** based algorithm (next lecture).

Algorithms via Basic Search

Algorithms via Basic Search - I

- Given G and nodes u and v, can u reach v?
- Given G and u, compute rch(u).

Algorithms via Basic Search - I

- Given G and nodes u and v, can u reach v?
- Given G and u, compute rch(u).

Use Explore(G, u) to compute rch(u) in O(n + m) time.

Algorithms via Basic Search - II

• Given G and u, compute all v that can reach u, that is all v such that $u \in \operatorname{rch}(v)$.

$$S = \xi$$
for $v \in V$

$$if \quad u \in \frac{v ch(v)}{s v \xi v}$$

$$O(m + n)$$

neturn S

Algorithms via Basic Search - II

Given G and u, compute all v that can reach u, that is all v such that u ∈ rch(v).
 Naive: O(n(n + m))

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Definition (Reverse graph.) Given G = (V, E), G^{rev} is the graph with edge directions reversed $G^{rev} = (V, E')$ where $E' = \{(y, x) \mid (x, y) \in E\}$ rch(1) = 212 in G runtime = 2 O(n+m) = O(n+m)

Given G and u, compute all v that can reach u, that is all v such that u ∈ rch(v). Naive: O(n(n + m))

Definition (Reverse graph.) Given G = (V, E), G^{rev} is the graph with edge directions reversed $G^{rev} = (V, E')$ where $E' = \{(y, x) \mid (x, y) \in E\}$

Compute rch(u) in G^{rev} !

Running time: O(n + m) to obtain G^{rev} from G and O(n + m) time to compute rch(u) via Basic Search. If both Out(v) and In(v) are available at each v then no need to explicitly compute G^{rev}. Can do Explore(G, u) in G^{rev} implicitly.

• Find the strongly connected component containing node *u*. That is, compute *SCC*(*G*, *u*).

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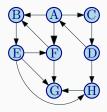
• Find the strongly connected component containing node *u*. That is, compute *SCC*(*G*, *u*).

 $SCC(G, u) = \operatorname{rch}(G, u) \cap \operatorname{rch}(G^{rev}, u)$

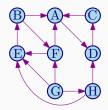
Hence, SCC(G, u) can be computed with Explore(G, u) and $Explore(G^{rev}, u)$. Total O(n + m) time.

Why can $\operatorname{rch}(G, u) \cap \operatorname{rch}(G^{rev}, u)$ be done in O(n) time?

Graph G and its reverse graph G^{rev}

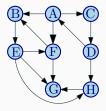


Graph G

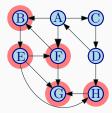


Reverse graph G^{rev}

Graph G a vertex F and its reachable set $\operatorname{reb}(G, F)$



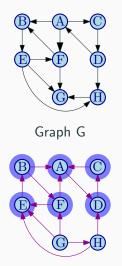
Graph G



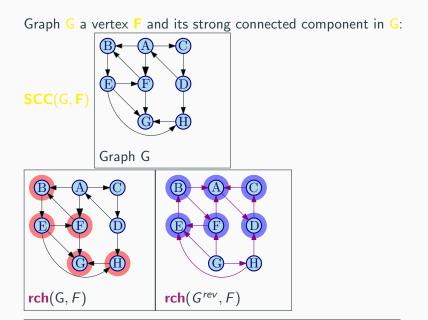
Reachable set of vertices from F

SCC III

Graph G a vertex F and the set of vertices that can reach it in $G:rch(G^{rev}, F)$



SCC IV: ...



• Is **G** strongly connected?

• Is *G* strongly connected?

Pick arbitrary vertex u. Check if SCC(G, u) = V.

Algorithms via Basic Search - V

• Find all strongly connected components of *G*.

Algorithms via Basic Search - V

• Find all strongly connected components of G.

While <mark>G</mark> is <mark>not empty do</mark>
Pick arbitrary node u
find $S = SCC(G, u)$
Remove <mark>S</mark> from G

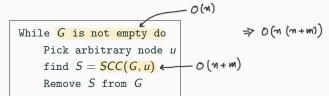
• Find <u>all</u> strongly connected components of *G*.

```
While G is not empty do
Pick arbitrary node u
find S = SCC(G, u)
Remove S from G
```

Question: Why doesn't removing one strong connected (1) components affect the other strong connected components?

Use contradiction

• Find all strongly connected components of G.



Question: Why doesn't removing one strong connected components affect the other strong connected components?

Running time: O(n(n+m)).

• Find <u>all</u> strongly connected components of *G*.

```
While G is not empty do
Pick arbitrary node u
find S = SCC(G, u)
Remove S from G
```

Question: Why doesn't removing one strong connected components affect the other strong connected components?

Running time: O(n(n+m)).

Question: Can we do it in O(n + m) time?

Find out next time.....