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

	E	D	R	Ε	A	D
E						
D						
Е						
Ε						
D						

Is there an easier way to get the minimum alignment without having to calculate all the values in the cell?

ECE-374-B: Lecture 15 - Graph search

Instructor: Nickvash Kani

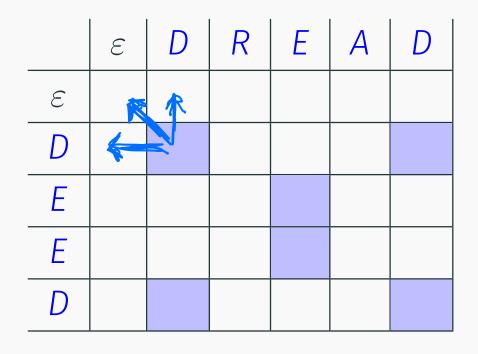
March 09, 2023

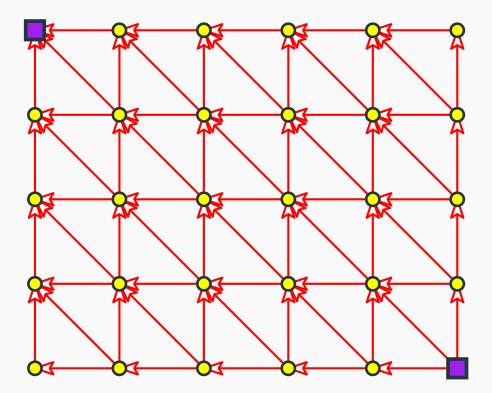
University of Illinois at Urbana-Champaign

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

	E	D	R	Ε	A	D
E						
D						
Ε						
Ε						
D						

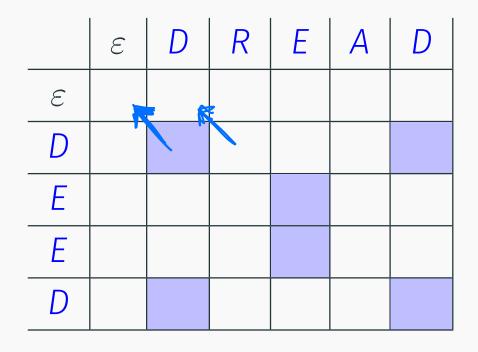
Is there an easier way to get the minimum alignment without having to calculate all the values in the cell? Remembering the edit distance example we saw in class last time, we formaluted the processing of the recursion as a table:

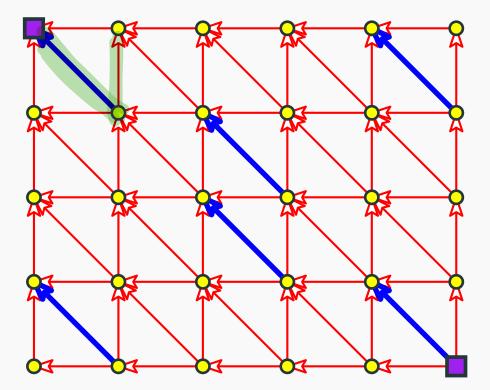




Look at the flow of the computation!

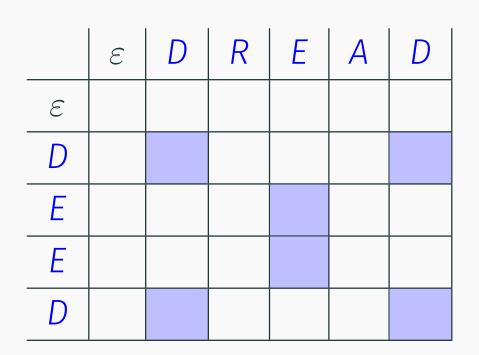
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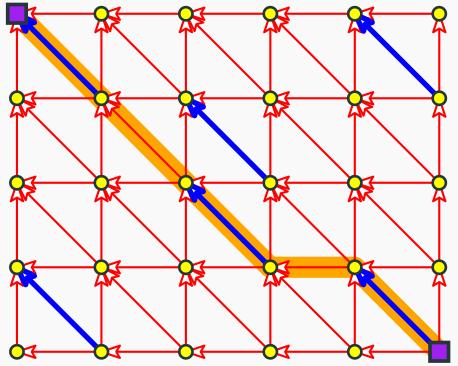




Look at the flow of the computation!

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





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

Graph Basics

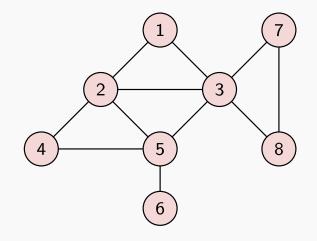
Why Graphs?

- Graphs help model networks which are ubiquitous: 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 Computer Science, Optimization, Combinatorics
- Many important and useful optimization problems are graph problems
- Graph theory: elegant, fun and deep mathematics

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

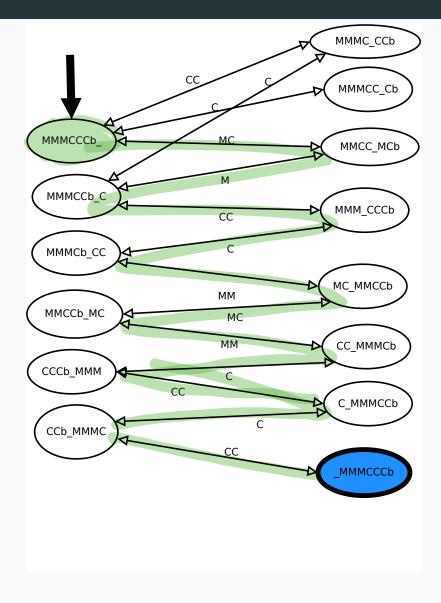
Many search problems can be modeled as search on a graph. The trick is figuring out what the vertices and edges are.

Missionaries and Cannibals

- 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...

*Omitted states where cannibals outnumber missionaries

Problems on DFAs and NFAs sometimes are just problems on graphs

- M: DFA/NFA is L(M) empty?
- 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?

Graph notation and representation

Notation

An edge in an undirected graphs is an <u>unordered</u> pair 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.

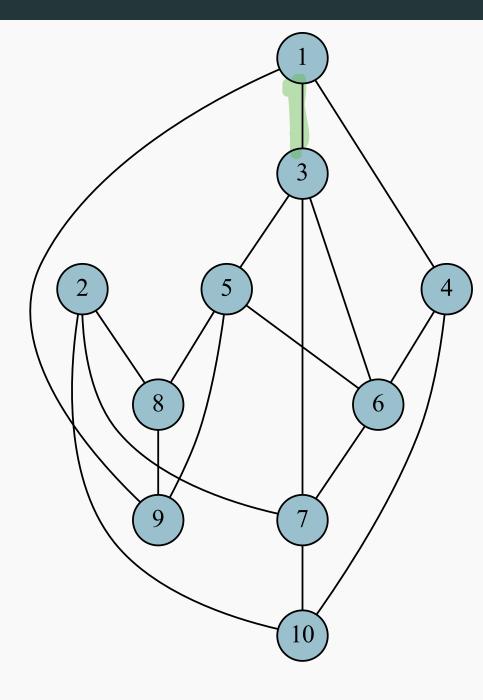
- *u* and *v* are the <u>end points</u> of an edge $\{u, v\}$
- <u>Multi-graphs</u> allow
 - <u>loops</u> which are edges with the same node appearing as both end points
 - <u>multi-edges</u>: different edges between same pairs of nodes
- In this class we will assume that a graph is a simple graph unless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

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 $m \ll n^2$

Graph adjacency matrix example [10 vertices]



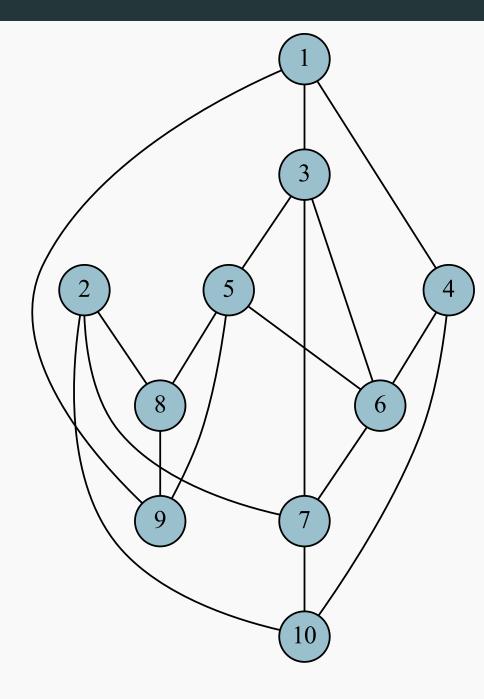
	1	2	3	4	5	6	7	8	9	10
1	0	0	1	1	0	0	0	0	1	0
2	0	0	0	0	0	0	1	1	0	1
3	1	0	0	0	1	1	1	0	0	0
4	1	0	0	0	0	1	0	0	0	1
5	0	0	1	0	0	1	0	1	1	0
6	0	0	1	1	1	0	1	0	0	0
7	0	1	1	0	0	1	0	0	0	1
8	0	1	0	0	1	0	0	0	1	0
9	1	0	0	0	1	0	0	1	0	0
10	0	1	0	1	0	0	1	0	0	0

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 hashing "appropriately", one can achieve O(1) time

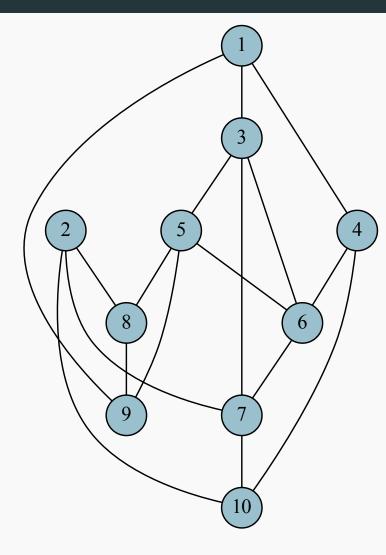
Note: In this class we will assume that by default, graphs are represented using plain vanilla (unsorted) adjacency lists.

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

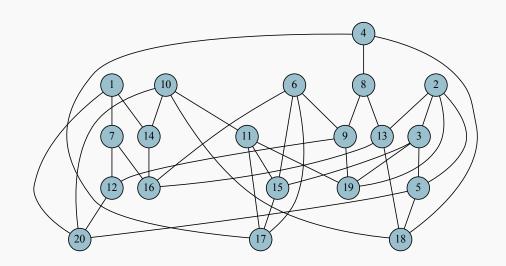
Graph adjacency matrix+list example [10 vertices]



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

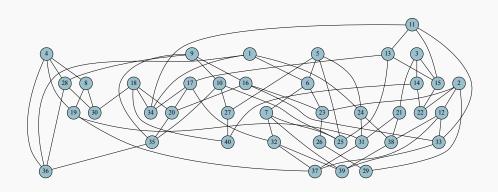
	1	2	3	4	5	6	7	8	9	10
1	0	0	1	1	0	0	0	0	1	0
2	0	0	0	0	0	0	1	1	0	1
3	1	0	0	0	1	1	1	0	0	0
4	1	0	0	0	0	1	0	0	0	1
5	0	0	1	0	0	1	0	1	1	0
6	0	0	1	1	1	0	1	0	0	0
7	0	1	1	0	0	1	0	0	0	1
8	0	1	0	0	1	0	0	0	1	0
9	1	0	0	0	1	0	0	1	0	0
10	0	1	0	1	0	0	1	0	0	0

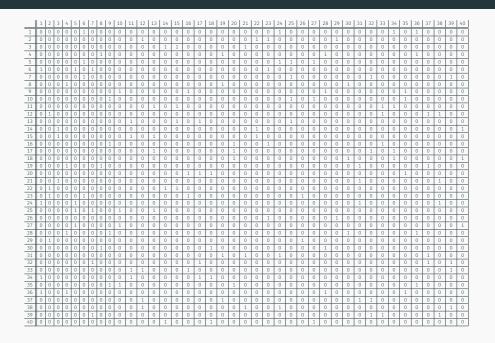
Graph adjacency matrix example [20 vertices]



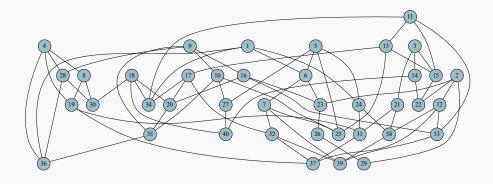
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
1	0	0	0	0	0	0	1	0	0	0	0	0	0	1	0	0	0	0	0	1
2	0	0	1	0	1	0	0	0	0	0	0	0	1	0	0	0	0	0	1	0
3	0	1	0	0	1	0	0	0	0	0	0	0	0	0	1	0	0	0	1	0
4	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	1	1	0	0
5	0	1	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	1
6	0	0	0	0	0	0	0	0	1	0	0	0	0	0	1	1	1	0	0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

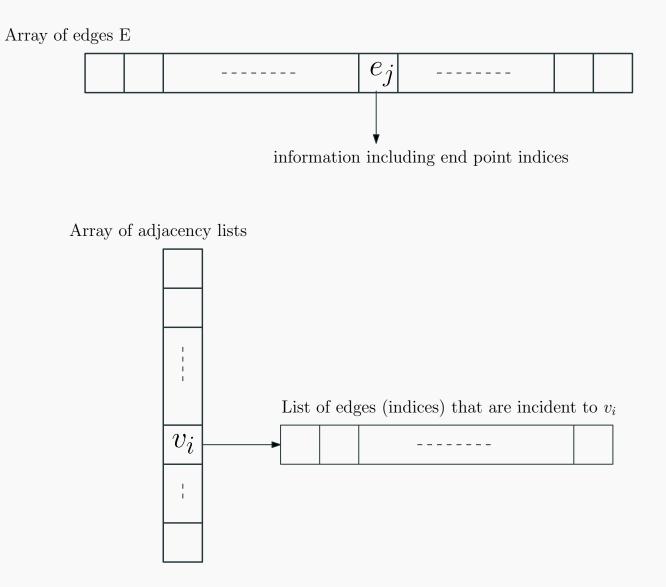


vertex	adjacency list
1	6, 24, 34, 36
2	12, 22, 23, 29
3	14, 15, 21
4	8, 19, 28, 36
5	6, 24, 25, 27
6	1, 5, 7, 23
7	6, 25, 32, 39
8	4, 19, 30
9	10, 16, 28, 35
10	9, 25, 27, 35
11	13, 15, 33, 34
12	2, 33, 37, 38
13	11, 15, 17, 25
14	3, 22, 40
15	3, 11, 13, 22
16	9, 20, 23, 33
17	13, 20, 32, 34 20, 30, 34, 40
18	
19	4, 8, 31, 37
20	16, 17, 18, 35
21	3, 31, 38
22	2, 14, 15
23	2, 6, 16, 26
24	1, 5, 31, 38
25	5, 7, 10, 13
26	23, 29
27	5, 10, 40
28	4, 9, 30, 36
29	2, 26
30	8, 18, 28
31	19, 21, 24, 37
32	7, 17, 37, 39
33	11, 12, 16, 39
34	1, 11, 17, 18
35	9, 10, 20, 36
36	1, 4, 28, 35
37	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 $\{1, 2, \ldots, n\}$.
- Edges are numbered arbitrarily as $\{1, 2, \ldots, m\}$.
- 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

- 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.

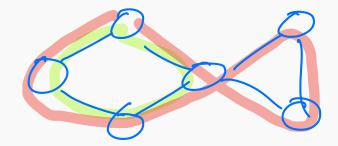
Given a graph G = (V, E):



• <u>path</u>: sequence of <u>distinct vertices</u> 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. <u>Base</u> Case

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.
- <u>cycle</u>: sequence of <u>distinct</u> vertices v_1, v_2, \ldots, v_k such that $\{v_i, v_{i+1}\} \in E$ for $1 \le i \le k 1$ and $\{v_1, v_k\} \in E$. Single vertex not a cycle according to this definition. Simple scale only allow vertices to be repeated; we will use the term tour.



Cycle Tour

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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- A vertex u is <u>connected</u> to v if there is a path from u to v.

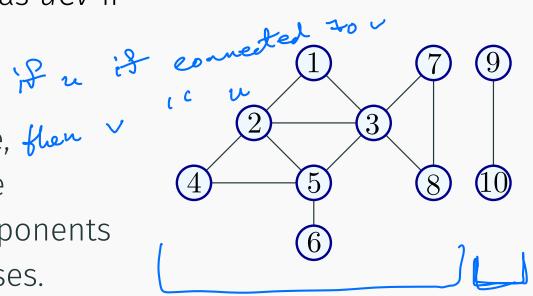
Given a graph G = (V, E):

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- A vertex *u* is <u>connected</u> to *v* if there is a path from *u* to *v*.
- The <u>connected component</u> of u, con(u), is the set of all vertices connected to u. Is $u \in con(u)$?

Connectivity contd

Define a relation C on $V \times V$ as uCv if u is connected to v

- In undirected graphs,
 connectivity is a reflexive, flow
 symmetric, and transitive
 relation. Connected components
 are the equivalence classes.
- Graph is <u>connected</u> if there is only one connected component.



Connectivity Problems

Algorithmic Problems

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

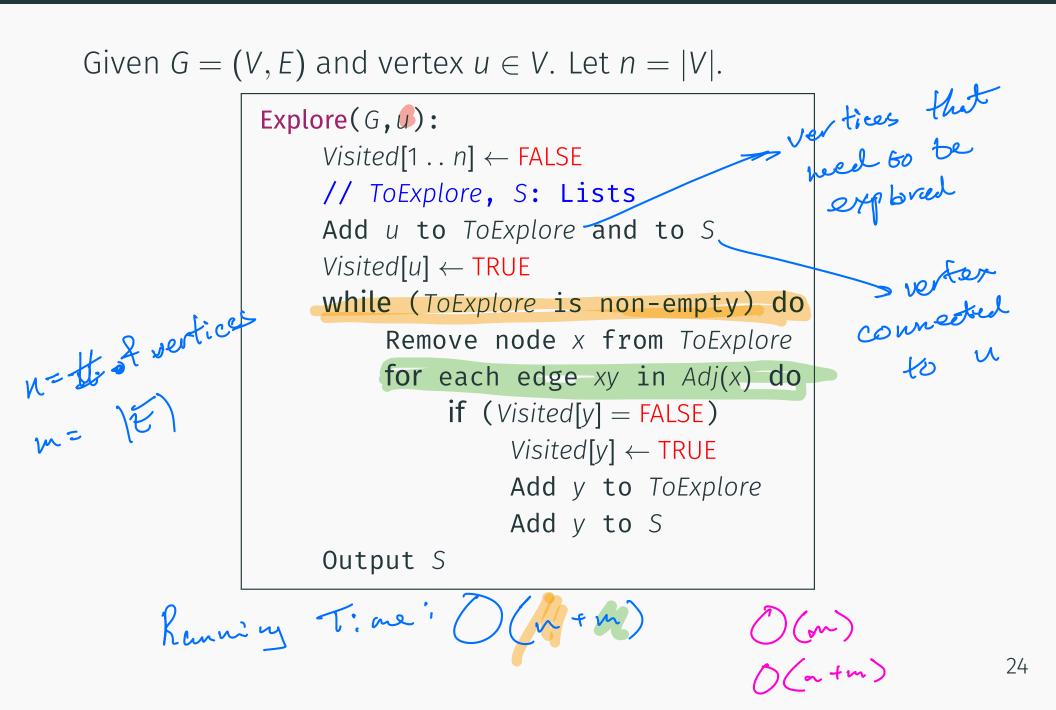
Connectivity Problems

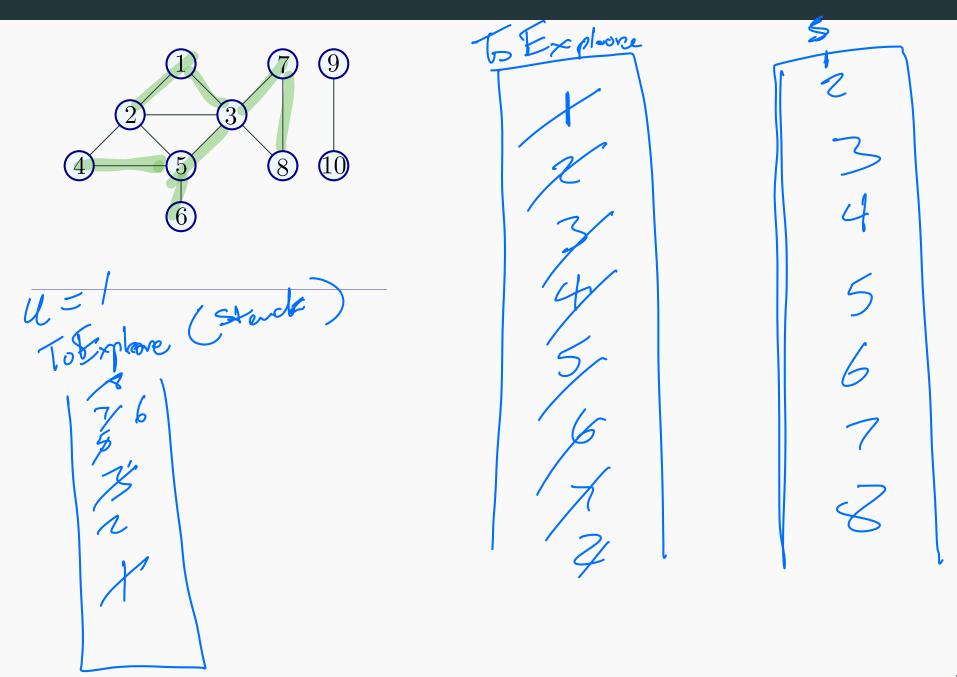
Algorithmic Problems

- Given graph G and nodes u and v, is u <u>connected</u> 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 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. Computing connected components in undirected graphs using basic graph search

Basic Graph Search in Undirected Graphs





Running Time:

Running Time:

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 <u>stack</u> data structure to implement the list *ToExplore*

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 u
 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 y to S
                Add y to T with x as its parent
 Output S
```

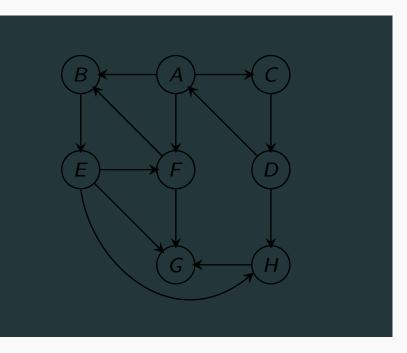
Modify Basic Search to find all connected components of a given graph G in O(m + n) time.

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:

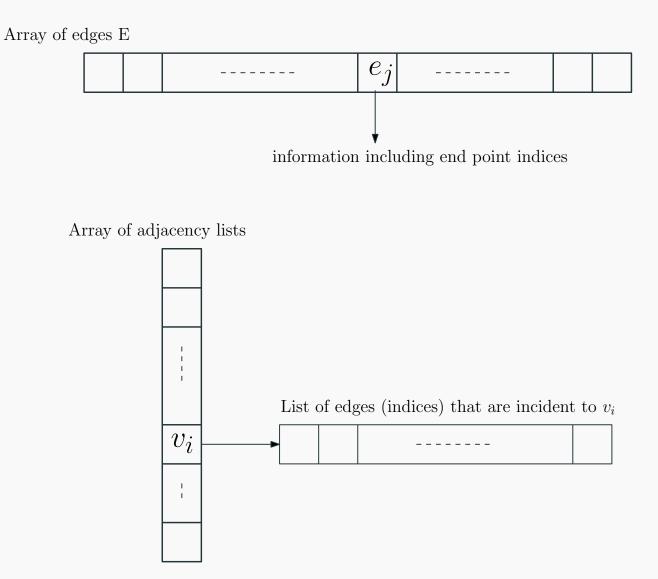
- Road networks with one-way streets.
- Web-link graph: vertices are web-pages and there is an edge from page p to page p' if p has a link to p'. Web graphs used by Google with PageRank algorithm to rank pages.
- Dependency graphs in variety of applications: link from *x* to *y* if *y* depends on *x*. 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:

- <u>Adjacency Matrix</u>: $n \times n$ <u>asymmetric</u> matrix A. A[u, v] = 1 if $(u, v) \in E$ and A[u, v] = 0 if $(u, v) \notin E$. A[u, v] is not same as A[v, u].
- <u>Adjacency Lists</u>: for each node u, Out(u) (also referred to as Adj(u)) and In(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.



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.

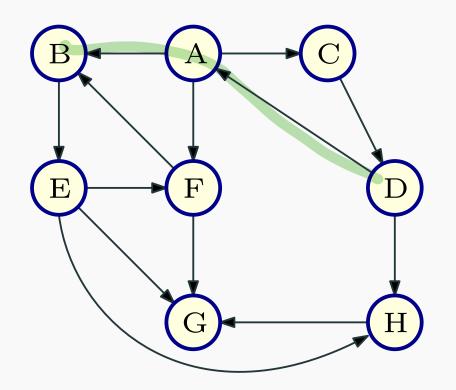
- A (directed) path is a sequence of distinct vertices
 v₁, v₂, ..., v_k such that (v_i, v_{i+1}) ∈ E for 1 ≤ i ≤ k − 1. The length of the path is k − 1 and the path is from v₁ to v_k. By convention, a single node u is a path of length 0.
- A <u>cycle</u> is a sequence of <u>distinct</u> vertices $v_1, v_2, ..., v_k$ such that $(v_i, v_{i+1}) \in E$ for $1 \le i \le k 1$ and $(v_k, v_1) \in E$. By convention, a single node u is not a cycle.

- 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 <u>cycle</u> is a sequence of <u>distinct</u> vertices $v_1, v_2, ..., v_k$ such that $(v_i, v_{i+1}) \in E$ for $1 \le i \le k 1$ and $(v_k, v_1) \in E$. By convention, a single node u is not a cycle.
- A vertex u can <u>reach</u> v if there is a path from u to v.
 Alternatively v can be reached from u

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 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 <u>cycle</u> is a sequence of <u>distinct</u> 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 <u>reach</u> 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.

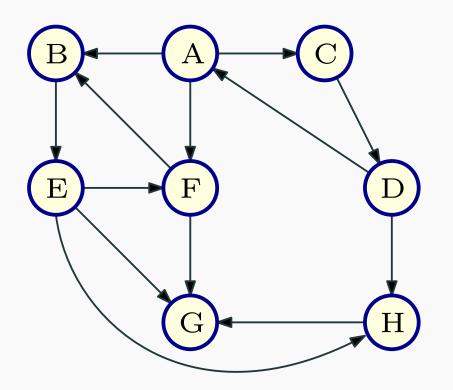
Connectivity contd

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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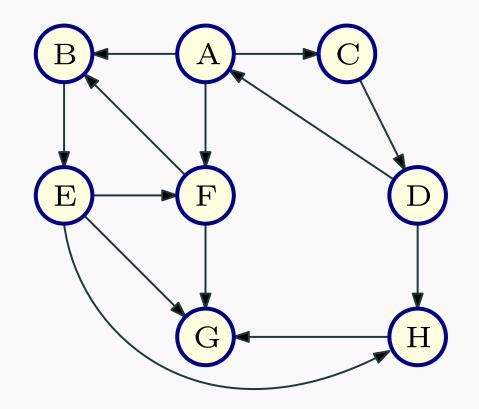
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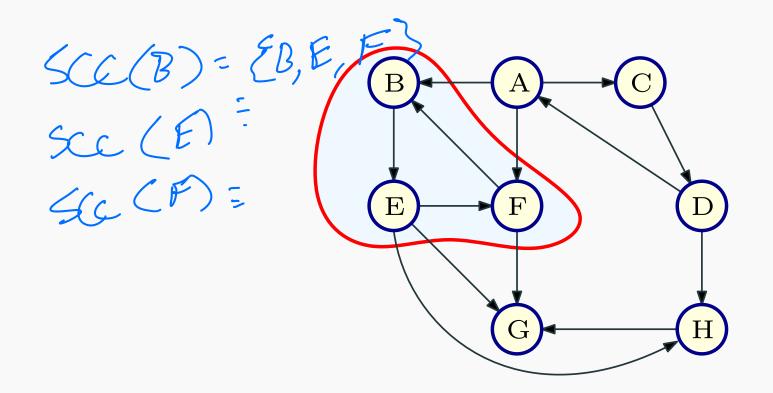
Proposition

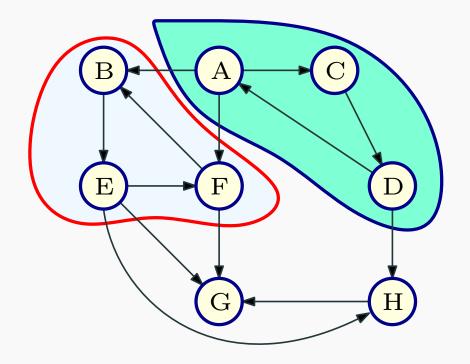
C is an equivalence relation, that is reflexive, symmetric and transitive.

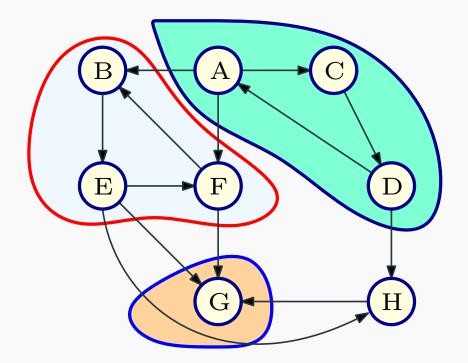
Equivalence classes of *C*: <u>strong connected components</u> of *G*. They <u>partition</u> the vertices of *G*.

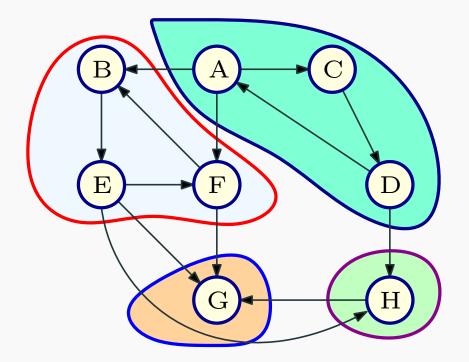
SCC(u): strongly connected component containing u.











Directed Graph Connectivity Problems

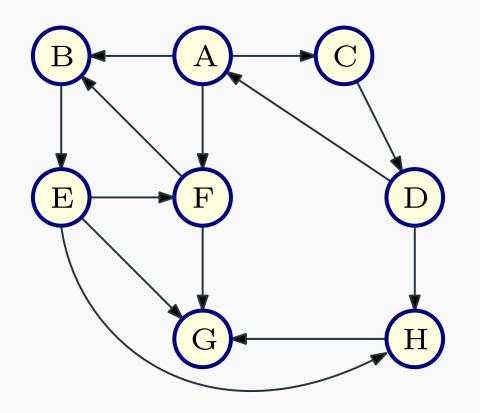
- Given G and nodes u and v, can u reach v?
- Given G and u, compute rch(u).
- Given G and u, compute all v that can reach u, that is all v such that $u \in \operatorname{rch}(v)$.
- Find the strongly connected component containing node u, that is SCC(u).
- Is G strongly connected (a single strong component)?
- Compute <u>all</u> strongly connected components of *G*.

Graph exploration in directed graphs

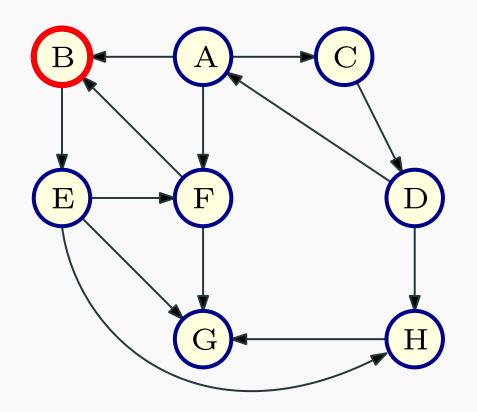
Basic Graph Search in Directed Graphs

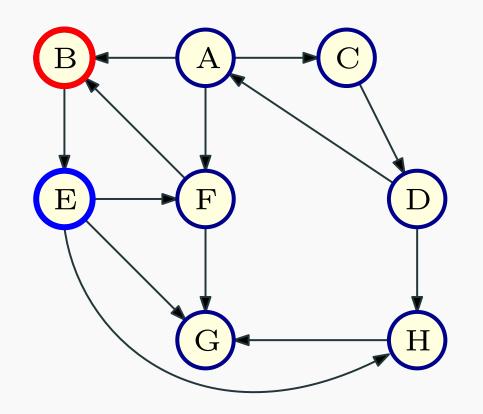
Given G = (V, E) a directed graph and vertex $u \in V$. Let n = |V|.

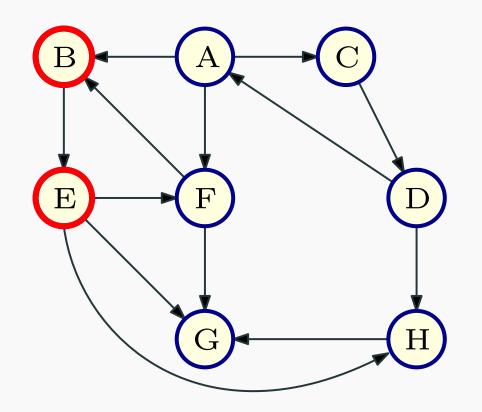
```
Explore(G,u):
  array Visited[1..n]
  Initialize: Set Visited[i] \leftarrow FALSE for 1 \le i \le n
  List: ToExplore, S
  Add u to ToExplore and to S, Visited[u] \leftarrow TRUE
  Make tree T with root as u
  while (ToExplore is non-empty) do
       Remove node x from ToExplore
      for each edge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

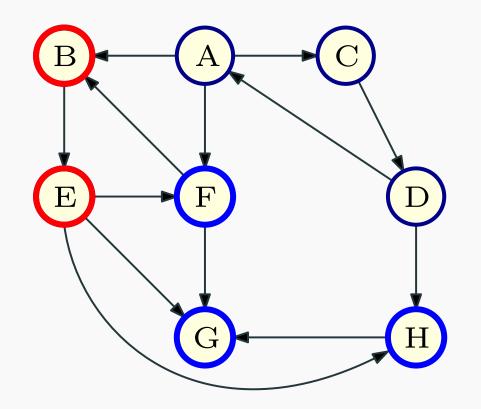


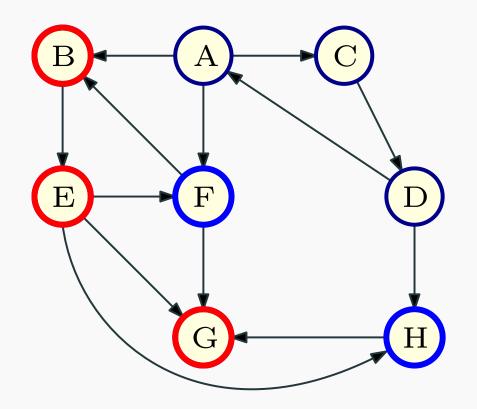
V=B

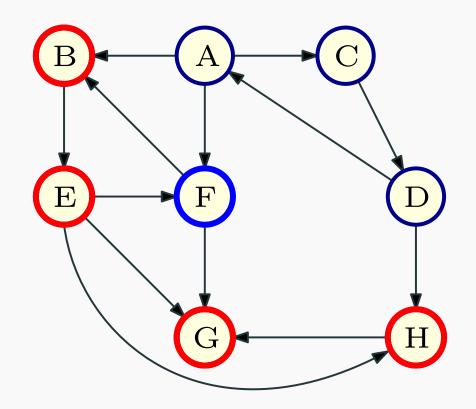


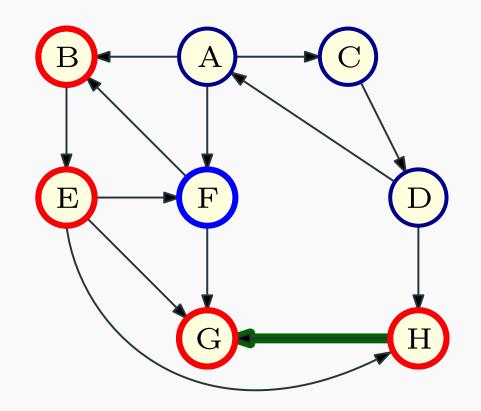


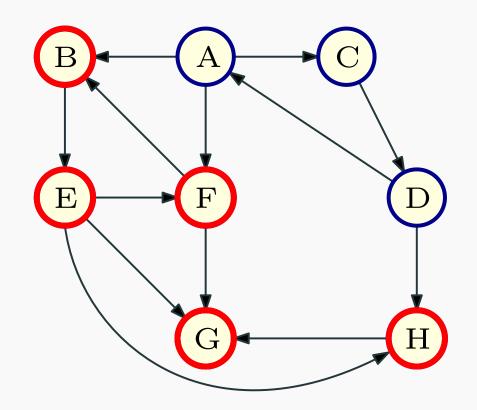


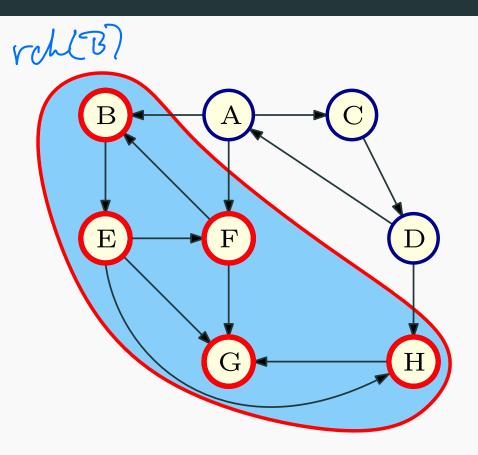












Proposition Explore(G, u) terminates with S = rch(u).

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.

Directed Graph Connectivity Problems

- Given G and nodes u and v, can u reach v? $V \in \operatorname{ToExplore}(G, u)$
- Given G and u, compute rch(u).
- Given G and u, compute all v that can reach u, that is all v such that $u \in \operatorname{rch}(v)$. $v \cdot \operatorname{So}_{\operatorname{Explore}}(v) = v \circ \operatorname{O}(v + v)$
- Find the strongly connected component containing node *u*, that is *SCC(u)*.
- Is G strongly connected (a single strong component)?
- Compute <u>all</u> strongly connected components of *G*.

Directed Graph Connectivity Problems

- Given G and nodes u and v, can u reach v?
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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

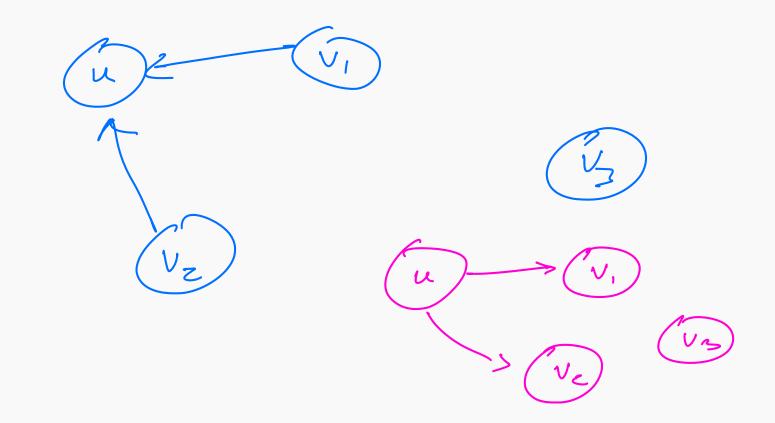
• Given G and u, compute all v that can reach u, that is all v such that $u \in \operatorname{rch}(v)$.

To Explore (G, u) -> veh (w)

Algorithms via Basic Search - II

• Given G and u, compute all v that can reach u, that is all v such that $u \in \operatorname{rch}(v)$. Naive: O(n(n + m)) • Given G and u, compute all v that can reach u, that is all v such that $u \in \operatorname{rch}(v)$. 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 $E' = \{(y, x) \mid (x, y) \in E\}$



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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 $E' = \{(y, x) \mid (x, y) \in E\}$

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*).

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

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

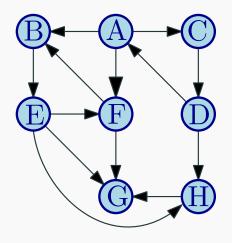
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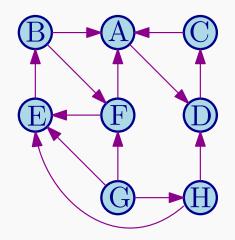
Hence, SCC(G, u) can be computed with Explore(G, u) and $Explore(G^{rev}, u)$. Total O(n + m) time.

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

Graph G and its reverse graph G^{rev}

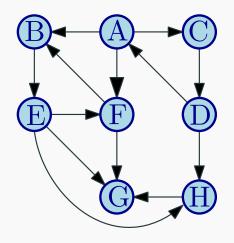


Graph G

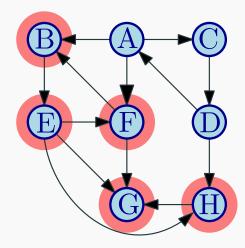


Reverse graph G^{rev}

Graph G a vertex F and its reachable set rch(G, F)

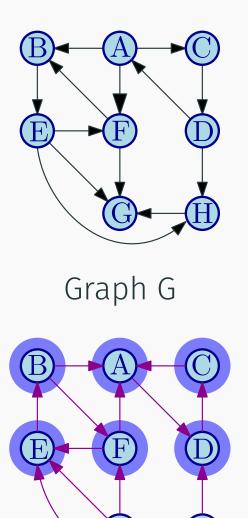


Graph G



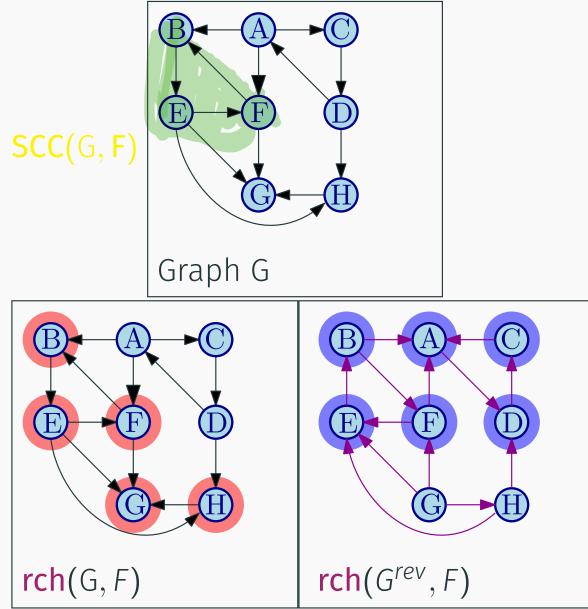
Reachable set of vertices from *F*

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





Graph G a vertex F and its strong connected component in G:



Algorithms via Basic Search - IV

• Is G strongly connected?

Algorithms via Basic Search - IV

• Is G strongly connected?

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

Algorithms via Basic Search - V

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

Algorithms via Basic Search - V

• 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



Algorithms via Basic Search - V

• 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?

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

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While G is not empty do
Pick arbitrary node u
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```
Running time: O(n(n + m)).
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```
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```

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

Find out next time.....