
DSC40B:
Theoretical Foundations of Data
Science II

Lecture 13: *Depth-First Search (DFS)*

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One more property about BFS

- ▶ For BFS algorithm, at any moment of the algorithm:
 - ▶ Recall the queue stores all the **pending** nodes (discovered, but not explored)
 - ▶ Recall we are inserting nodes into the queue in non-decreasing order their distance from source
- ▶ It is easy to verify that
 - ▶ The shortest path distance from the source are non-decreasing in the queue
 - ▶ The shortest path distance for nodes in the queue cannot differ more than 1



Prelude

- ▶ Previously, **Search strategy** :
- ▶ How to decide which is the next node to explore?
 - ▶ **BFS (Breadth-first search)**:
 - ▶ choose the ``oldest'' pending node to explore and expand
 - ▶ consequently, it explores as wide as possible before goes any ``deeper'' (in terms of distance to the source)
 - i.e, it visits all nodes at distance k to the source before moving to any node at distance $k + 1$ from the source.
- ▶ Today: **Depth-first search (DFS)**:
 - ▶ Choose the ``newest'' pending node to explore
 - ▶ Consequently, it will go as deep (farther away from source) as possible during exploration



DFS algorithm and time complexity



Depth-first Search (DFS)

▶ DFS(G, s)

- ▶ It will perform depth-first search in G starting from a graph node s called the *source* node.

▶ Idea:

- ▶ All nodes are initialized as **undiscovered**, other than the source node, which is initialized as **pending** (i.e, discovered, and to be processed)
- ▶ At each step:
 - ▶ take the **newest pending** node to explore
 - ▶ *explore all undiscovered nodes reachable from this node*
 - ▶ then mark this node to be **visited**
- ▶ Repeat till there is no more **pending** nodes to explore



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- ▶ To be able to extract the “**newest**” pending node
 - ▶ we need a standard **stack** data structure to provide **FIFO** (first-in-last-out)
 - ▶ In algorithm implementation,
 - ▶ we use **recursive algorithm** to achieve this **FIFO/stack** idea implicitly

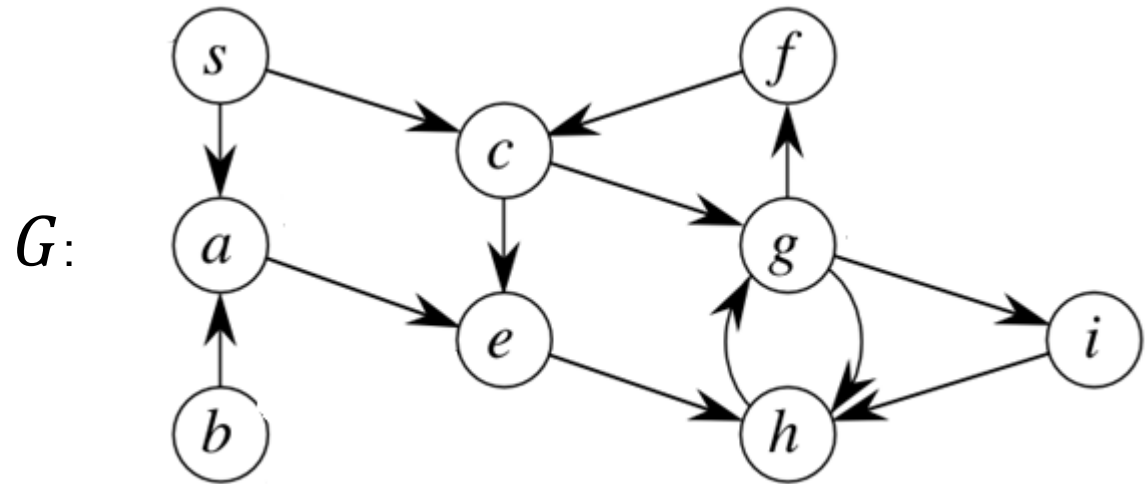


Implementation in Python

```
def dfs(graph, u, status=None):  
    """Start a DFS at `u`."""  
    # initialize status if it was not passed  
    if status is None:  
        status = {node: 'undiscovered' for node in graph.nodes}  
  
    status[u] = 'pending'  
    for v in graph.neighbors(u): # explore edge (u, v)  
        if status[v] == 'undiscovered':  
            dfs(graph, v, status)  
    status[u] = 'visited'
```



Example



► Call `dfs(G, s)`



Full DFS

- ▶ DFS will visit all nodes reachable from the source node
 - ▶ If the input graph is connected, then it will visit all nodes in the same connected component as the source
- ▶ To visit all nodes in a graph
 - ▶ Needs full-DFS
 - ▶ which requires we restart from undiscovered nodes.

```
def full_dfs(graph):  
    status = {node: 'undiscovered' for node in graph.nodes}  
    for node in graph.nodes:  
        if status[node] == 'undiscovered':  
            dfs(graph, node, status)
```



Complete code

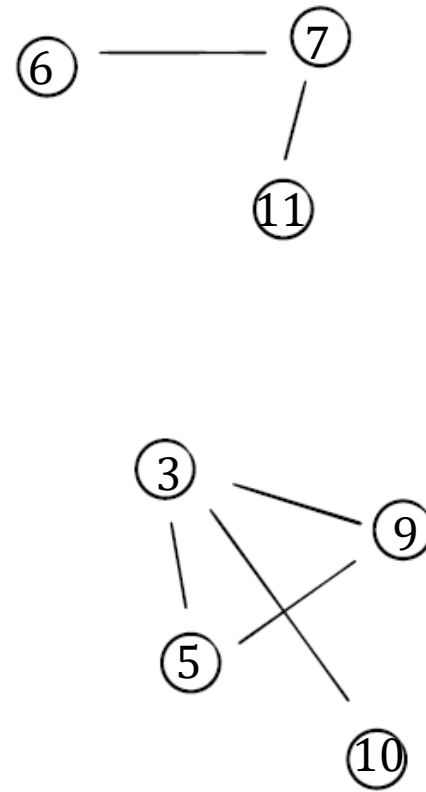
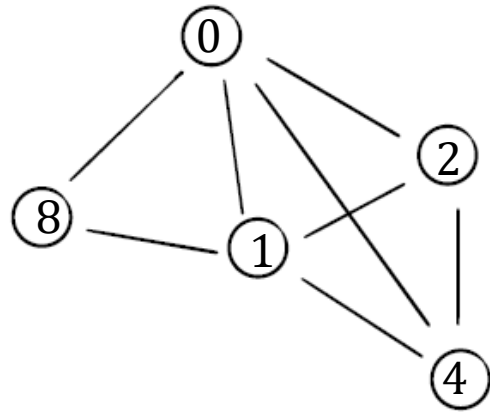
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    for v in graph.neighbors(u): # explore edge (u, v)
        if status[v] == 'undiscovered':
            dfs(graph, v, status)
    status[u] = 'visited'
```



Example



Time complexity analysis

- ▶ (similar to BFS): for full-DFS
 - ▶ Each node will be explored exactly once
 - ▶ Each edge will be traversed (visited) twice for undirected graphs
 - ▶ Each edge will be traversed (visited) once will for directed graphs
 - ▶ Hence total time complexity of full-DFS
 - ▶ $\Theta(|V| + |E|)$



Nesting properties in DFS



full_DFS

```
def full_dfs(graph):  
    status = {node: 'undiscovered' for node in graph.nodes}  
    for node in graph.nodes:  
        if status[node] == 'undiscovered':  
            dfs(graph, node, status)
```

```
def dfs(graph, u, status=None):  
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    for v in graph.neighbors(u): # explore edge (u, v)  
        if status[v] == 'undiscovered':  
            dfs(graph, v, status)  
    status[u] = 'visited'
```



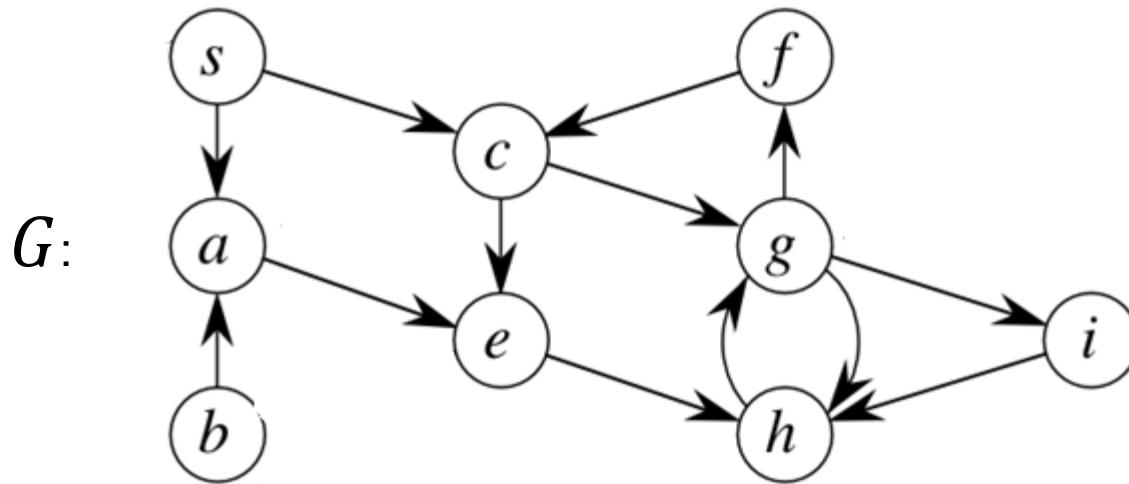
“Parent” (Predecessor) information

- ▶ Similar to BFS, for each node v ,
 - ▶ its (DFS-)predecessor is the node u where through exploring edge (u, v) the node v was first discovered (status changed to pending).
- ▶ Collection of edges of the form $(\text{predecessor}(v), v)$ will give a tree
 - ▶ called DFS-tree.
 - ▶ $\text{Predecessor}(v)$ is the parent of v in this DFS-tree.



Observations

- ▶ Between marking a node as **pending** and **visited**, many other nodes are marked **pending** or **visited**.



Start and Finish times

- ▶ A node status changed
 - ▶ from **undiscovered** to **pending**:
 - ▶ first time this node is discovered
 - ▶ from **pending** to **visited**:
 - ▶ exploration of this node is finished
- ▶ Keep an integer running clock
- ▶ For each node:
 - ▶ **Start time**: status changes from **undiscovered** to **pending**
 - ▶ **Finish time**: status changes from **pending** to **visited**
- ▶ Clock increments by 1 whenever some node is marked **pending/visiting**



full_DFS_times implementation

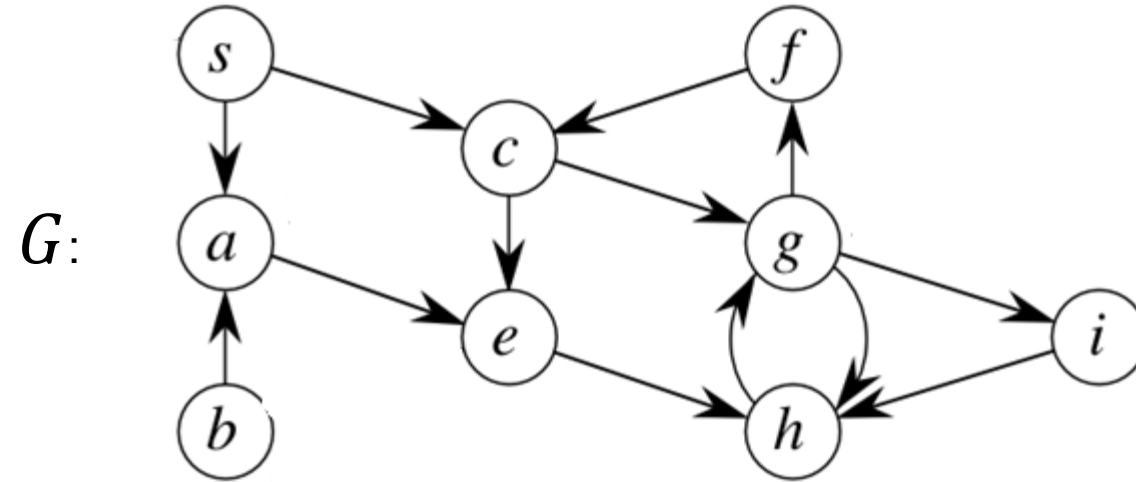
```
from dataclasses import dataclass
@dataclass
class Times:
    clock: int
    start: dict
    finish: dict
```

```
def full_dfs_times(graph):
    status = {node: 'undiscovered' for node in graph.nodes}
    predecessor = {node: None for node in graph.nodes}
    times = Times(clock=0, start={}, finish={})
    for u in graph.nodes:
        if status[u] == 'undiscovered':
            dfs_times(graph, u, status, times)
    return times, predecessor
```

```
def dfs_times(graph, u, status, predecessor, times):
    times.clock += 1
    times.start[u] = times.clock
    status[u] = 'pending'
    for v in graph.neighbors(u): # explore edge (u, v)
        if status[v] == 'undiscovered':
            predecessor[v] = u
            dfs_times(graph, v, status, times)
    status[u] = 'visited'
    times.clock += 1
    times.finish[u] = times.clock
```



Example



Nesting Property of DFS

▶ Claim A:

- ▶ Take any two nodes u and v . Assume $\text{start}[u] \leq \text{start}[v]$
- ▶ Exactly one of the following two is true:
 - ▶ $\text{start}[u] \leq \text{start}[v] \leq \text{finish}[v] \leq \text{finish}[u]$
 - ▶ $\text{start}[u] \leq \text{finish}[u] \leq \text{start}[v] \leq \text{finish}[v]$



Nesting Property of DFS

- ▶ Take any two nodes u and v where v is reachable from u
 - ▶ If $\text{start}[u] \leq \text{start}[v]$, meaning we visit u first.
 - ▶ then $\text{start}[u] \leq \text{start}[v] \leq \text{finish}[v] \leq \text{finish}[u]$
 - ▶ Intuitively, when exploring a node u , we will finish the exploration of all reachable nodes from u (if they are not yet discovered) before we finish exploring u

- ▶ Claim B: If node v is reachable from u , but u is not-reachable from v
 - ▶ then $\text{finish}[v] \leq \text{finish}[u]$



DAGs and topological sort

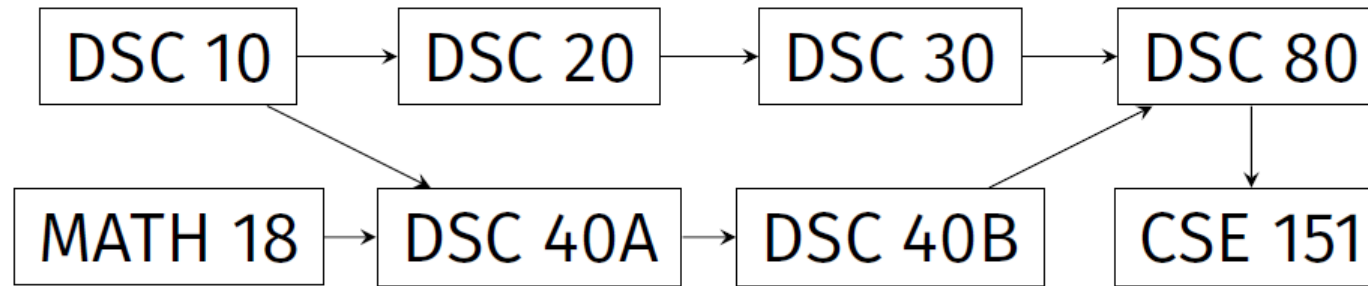


Applications of DFS

- ▶ Is node v reachable from u ?
- ▶ Is a undirected graph connected?
- ▶ How many connected components are there in a undirected graph?
- ▶ Is the input graph a tree?
- ▶ Find the shortest path to a source node u ?
 - ▶ **NO !**
 - ▶ Unlike BFS.



Prerequisite graphs

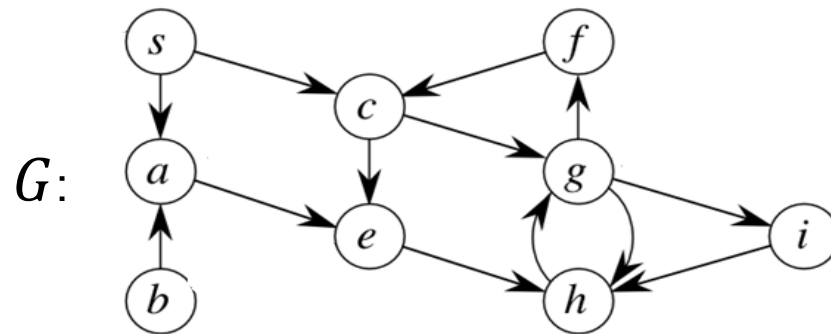


- ▶ Note that this graph is a directed graph
 - ▶ edge (u, v) means that course u is prerequisite for v
- ▶ Goal:
 - ▶ Find an ordering so as to take these classes satisfying all prerequisite requirements

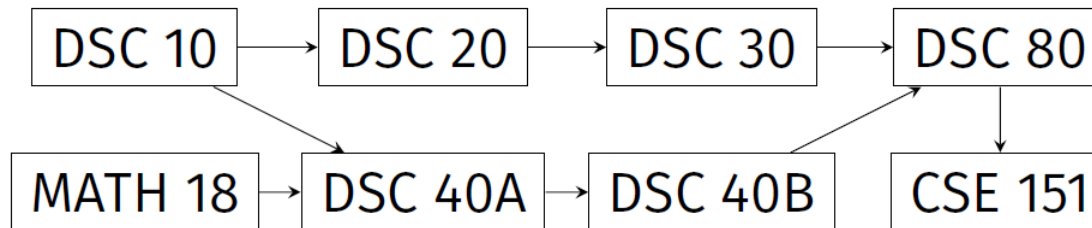


Directed Acyclic Graphs (DAGs)

- ▶ A **directed cycle** is a (directed) path from a node to itself.
- ▶ A **directed acyclic graph (DAG)** is a directed graph that does not contain any directed cycles



Not a DAG!

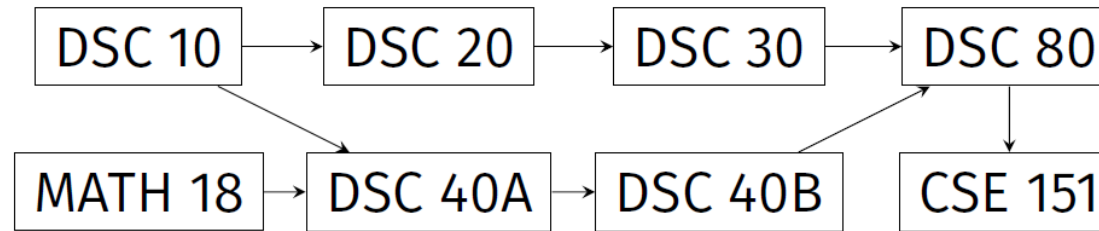


A DAG!



Topological Sorts

- ▶ Given a DAG $G = (V, E)$, a **topological sort** of G is an ordering of V such that
 - ▶ for any edge $(u, v) \in E$, then u comes before v in this ordering



A topological sort:

DSC10, Math18, DSC20, DSC40A, DSC30, DSC40B, DSC80, DSC151

Another topological sort:

Math18, DSC10, DSC20, DSC30, DSC40A, DSC40B, DSC80, DSC151

- ▶ Topological sorts of the same DAG are not unique.

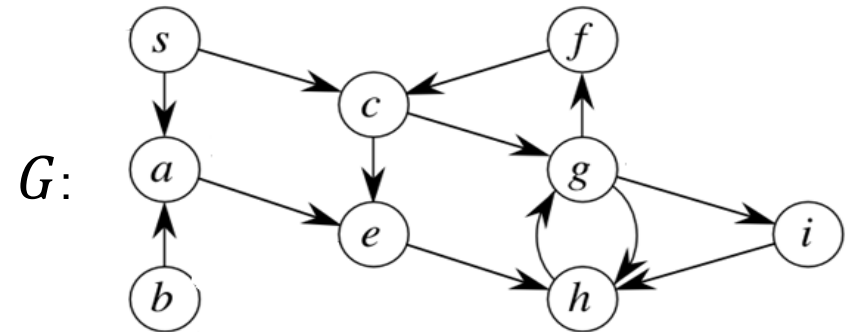
Why DAG?

► Claim:

- A directed graph $G = (V, E)$ can admit a topological sort **if and only if** G is a DAG!

► Why?

- If there is a cycle, then there is no valid ordering for nodes in that cycle!
- If it is a DAG, then we will give an algorithm to show we can compute topological sort.



An algorithm to compute topo-sort

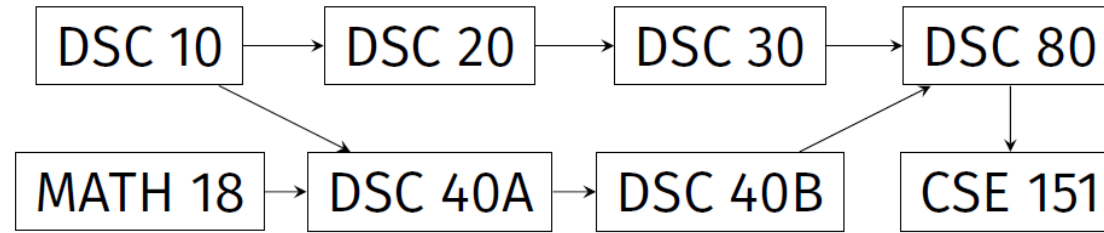
- ▶ Recall Claim B: If node v is reachable from u , but u is not-reachable from v
 - ▶ then $\text{finish}[v] \leq \text{finish}[u]$
- ▶ If v is reachable from u , then u should come before v in any topological sort.
- ▶ So nodes with later **finish-time** should come first

- ▶ Topo-sort Algorithm:
 - ▶ First perform DFS on input graph $G = (V, E)$
 - ▶ Output the order in decreasing order of **finish-time**.

- ▶ Time complexity: $\Theta(|V| + |E|)$



Example



Remark:

- ▶ There are many other ways to compute a topological sort for a DAG in the same time complexity, without using DFS.



Summary

▶ DFS:

- ▶ Yet another graph search strategy
- ▶ Similarly to BFS, as a graph search strategy, can help solve many problems, such as checking for connectivity, reachability, finding a path and so on.
- ▶ But has some different properties as BFS
 - ▶ BFS: useful for finding shortest path to the source
 - ▶ DFS: has nesting properties in terms of start/finish times.
 - An application: Topological sort in DAG



FIN

