DP in DAGs and Strongly Connected Components

Lecture 17

March 27, 2025

Part I

Dynamic Programming and DAGs

Recall LIS

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Consider the *subproblem dependency graph*:

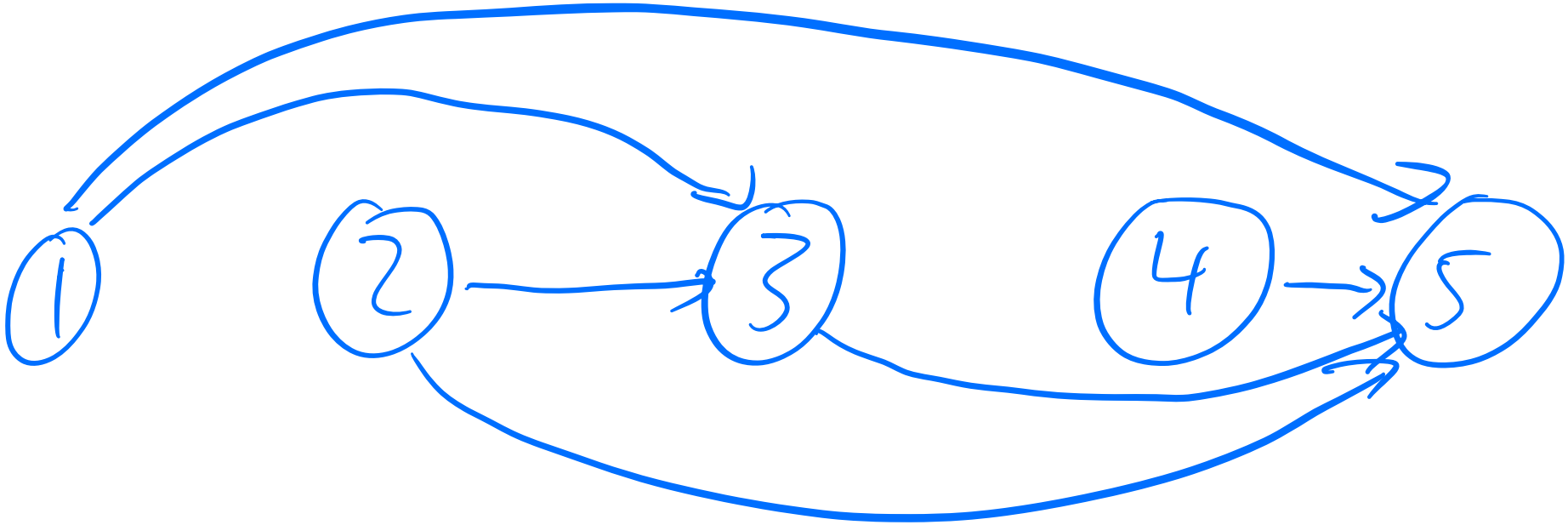
- Vertex i corresponds to subproblem $LIS(i)$
- Include edge (i, j) if computing $LIS(i)$ uses the value of $LIS(j)$

  ("Subproblem i depends on subproblem j")

Subproblem Dependency Graph Example

Suppose $A = [3, 1, 4, 1, 5]$.

(Recall $LIS(i) = 1 + \max\{LIS(j) \mid j > i \text{ and } A[j] > A[i]\}$)



DAG!

DP is DAGs

For *every* DP algorithm, the subproblem dependency graph is a DAG.

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An evaluation order is valid *if and only if* it is a reverse topological order of the dependency graph.

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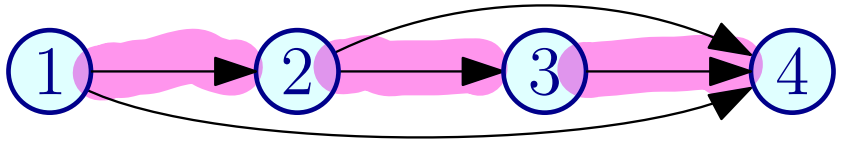
- For each vertex v, define some subproblem corresponding to v.
- Write a recurrence for your subproblems. (Most often v will refer to the subproblems of either its parents or its children.)
- Evaluate the subproblems in (reverse) topological order.

Example of DAG DP

Given an (unweighted) DAG G in topological order, what is the length of the longest path in G?



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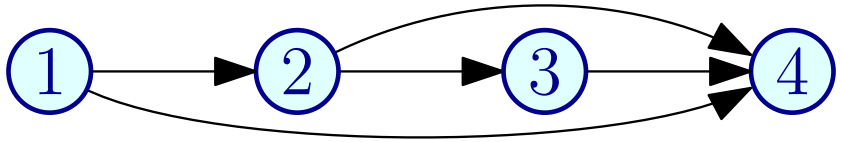

Subproblem definition?

LLP(v) is length of the longest path starting at v

Recurrence?

$$LLP(v) = \begin{cases} 0 & \text{if } v \text{ is a sink} \\ 1 + \max_{(v,u)} LLP(u) & \text{else} \end{cases}$$

$(\max_v LLP(v))$

Evaluation order?

reverse topological order

Longest Path in a DAG

```
DAG-LongestPath(G):
    Initialize array LLP
    for vertices v in reverse topological order:
        if v is a sink:
            LLP[v] = 0
        else
            LLP[v] = 1 + max{LLP(u) | (v, u) ∈ E}
    return max_v LLP(v)
```

repeat V times

$O(1)$

$O(deg(v))$

Efficiency? $\sum_v O(1) + O(deg(v)) = O(V) + O(E)$

$O(V+E)$

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Exercise: generalize this algorithm to

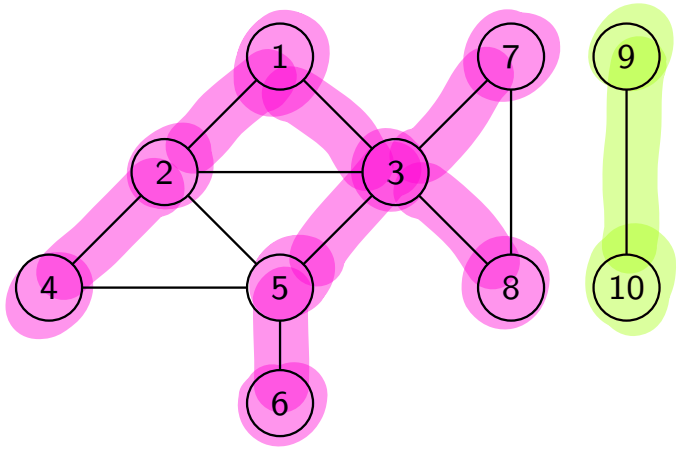
- Weighted edges
- Paths from s to t
- *Shortest* path from s to t
- ...

Part II

Strongly Connected Components

Recall Connected Components

In an *undirected* graph, we defined the connected component containing u as the set of all vertices it can reach.



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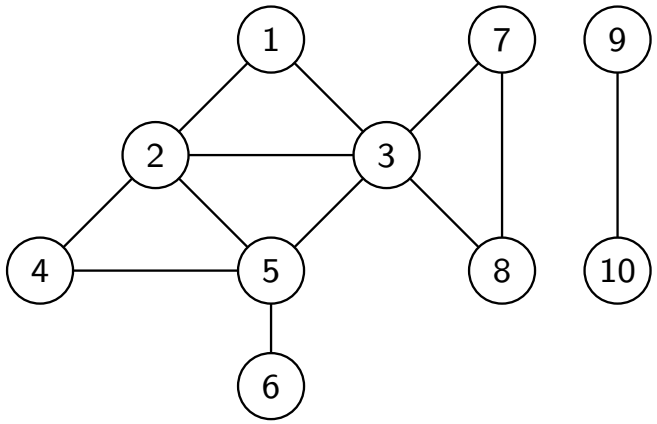
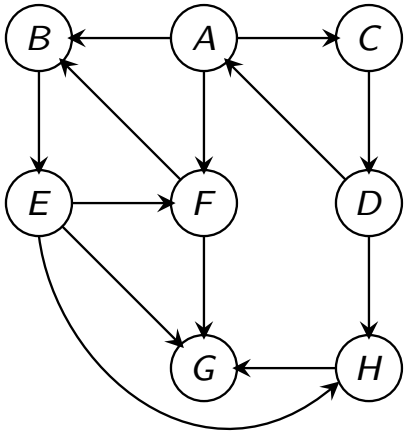


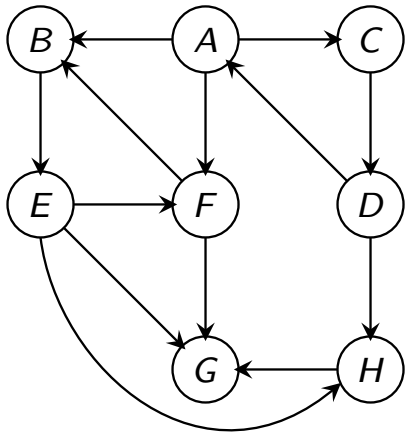


In a *directed* graph this relation isn't symmetric, so we can't talk about the "connected components of ***G***".

Strongly Connected Components

Definition

The ***strongly connected component*** (SCC) of a vertex ***v*** is the set of all vertices ***u*** such that ***v*** can reach ***u*** and ***u*** can reach ***v***.



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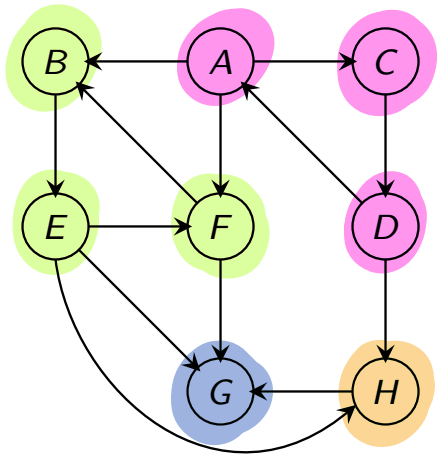

This definition doesn't depend on what vertex we choose within the SCC, so we can talk about the SCCs of ***G***!

The Meta-Graph

Once we have the SCCs of G, we can construct a "meta graph":

- A vertex for each SCC.
- An edge (C_1, C_2) iff there is an edge in G between *some* vertex in C_1 and *some* vertex in C_2.

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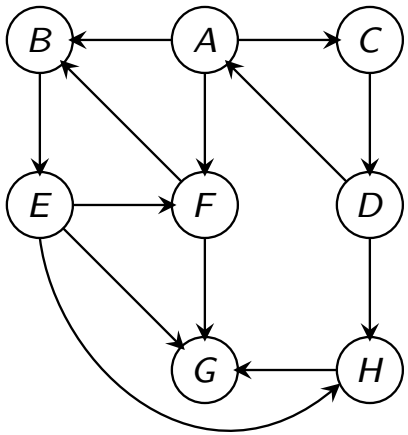

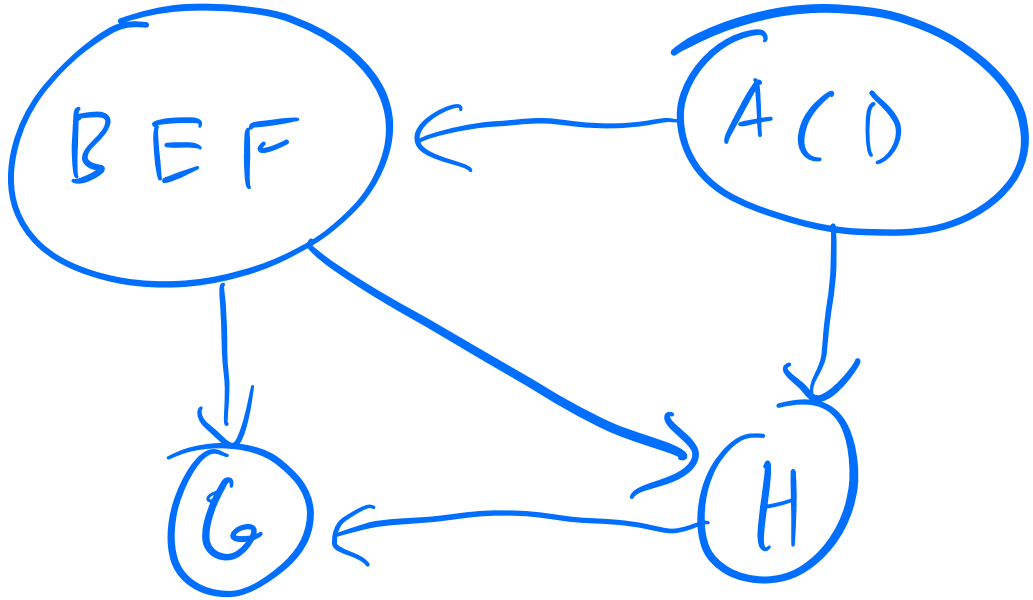

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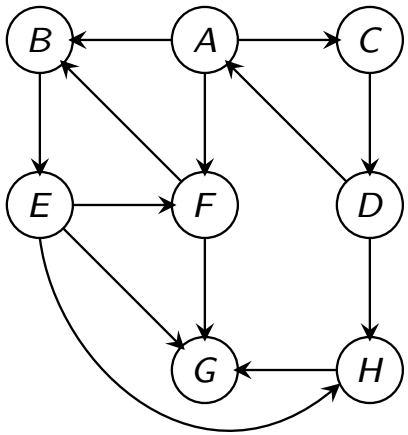
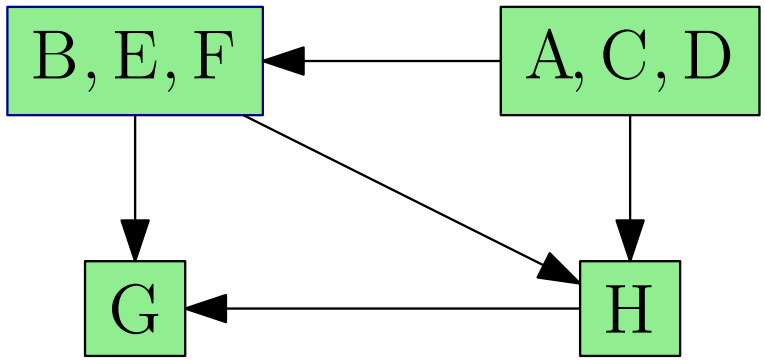
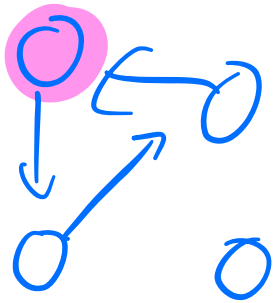


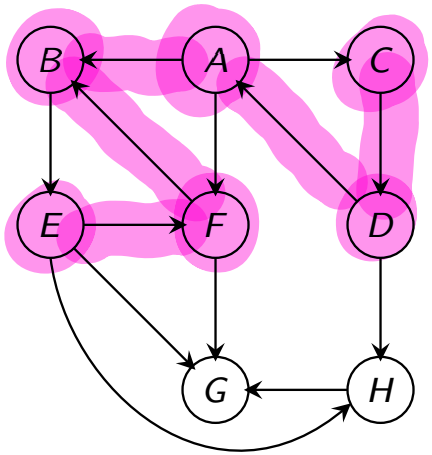


Meta graph will always be a DAG! (Also called the "DAG of SCCs".)

Warm-Up: Compute One SCC

How would we compute the SCC of a particular vertex? (Say ***B***)





what vertices can B reach? B, E, F, G, H

what vertices can reach B? A, B, C, D, E, F

$SCC(B) = \{B, E, F\}$

Naïve SCC Algorithm

```
FindSCC(G, v):
    Construct ``reverse graph'' G^R (reverse all edges)
    Compute reach(v) = WFS(G, v)
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→ repeat V times

→ O(V+E)

Efficiency?

Find SCC takes time O(V+E)

Naive SCC takes time O(V·(V+E))

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Efficiency? $O(V + E)$ for FindSCC, $O(V \cdot (V + E))$ for NaïveSCC

Can we do better?

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Efficiency?

WFS(G,v) will take time $O(V' + E')$, where $G' = (V', E')$ as the SCC of V.

total time $\sum_{SCC} O(V' + E') = O(V + E)$

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How do we find a vertex in a sink SCC though?

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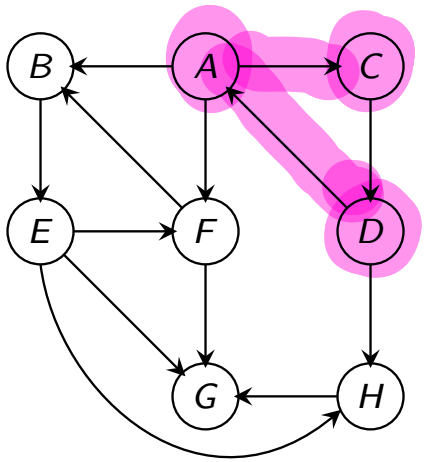

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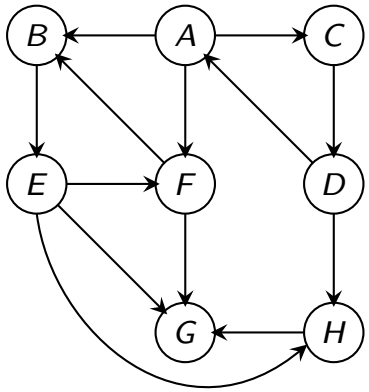

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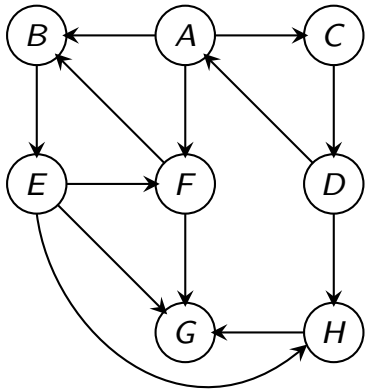

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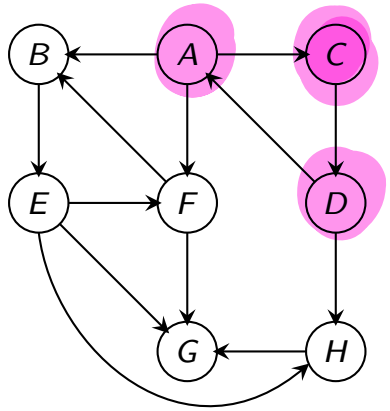

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Vertices ***A***, ***C***, and ***D*** aren't reachable from anywhere else, so `DFSAll` must (at some point) start a DFS run from one of them.

But *everything* is reachable from ***A***, ***C***, and ***D***, so that DFS run finds all remaining vertices, making it the last one.

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So the *last* run of DFS must start from a source SCC!

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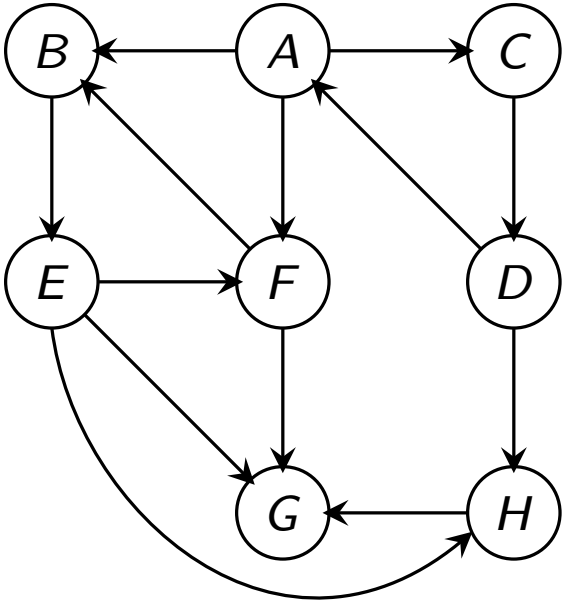
Efficiency? $O(V+E)$

Kosaraju-Sharir Example



Graph G

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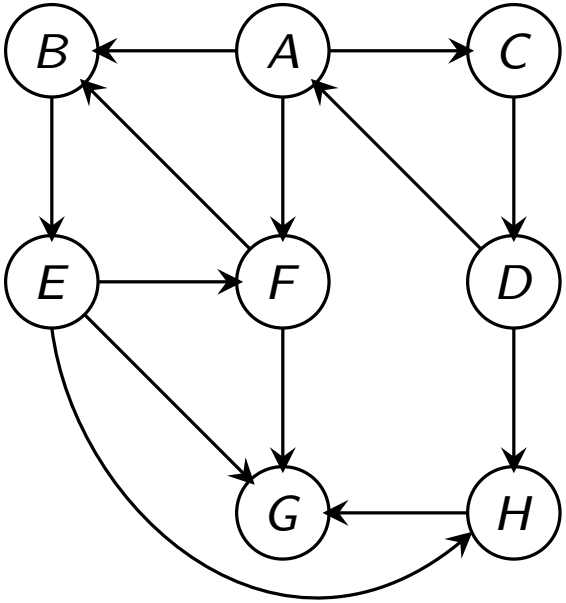

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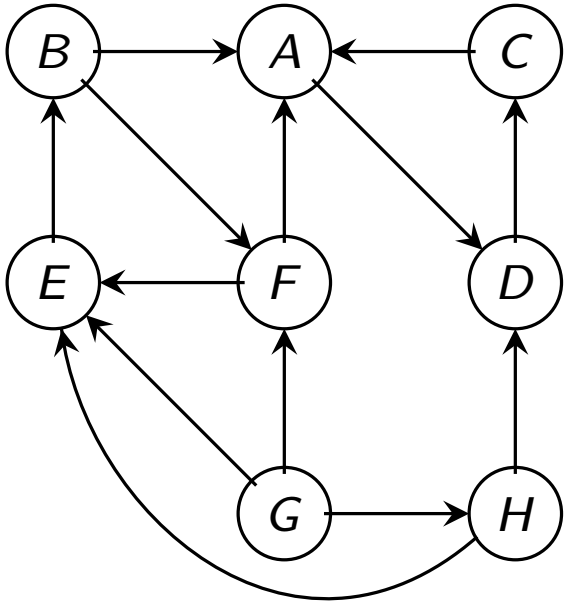

Graph G^R

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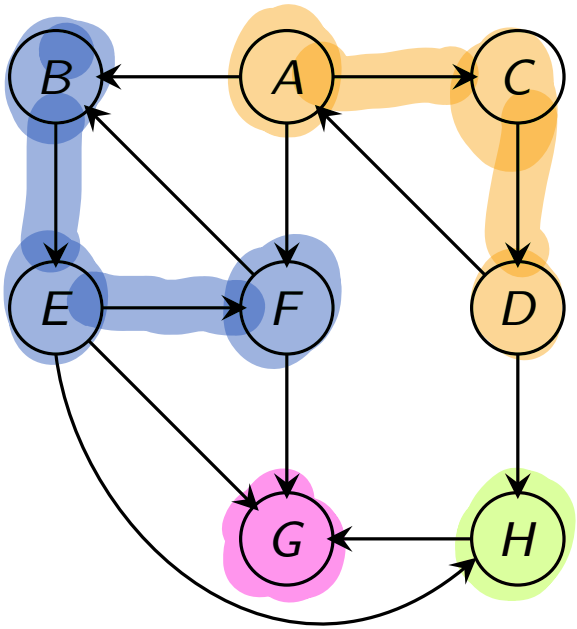

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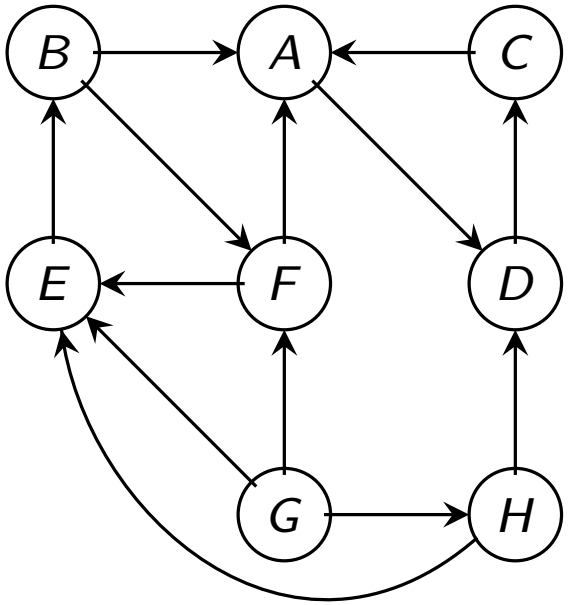

Graph G^R

Vertex	A	B	C	D	E	F	G	H
pre	1	7	3	2	9	8	13	14
post	6	12	4	5	10	11	16	15

Problems and Algorithms From Today

- Longest path in a DAG (weighted or unweighted)
  - Use DP, $O(V + E)$ time
  - Can also find shortest $s - t$ path
- Find the SCC of a single vertex
  - WFS on G and G^R, $O(V + E)$
- Find *all* SCCs of G
  - Kosaraju-Sharir, $O(V + E)$
  - Can also output the DAG of SCCs