

CS4102 Algorithms

Spring 2021 – Floryan and Horton

Module 4, Day 3: Recorded Lecture

Roadmap: Where We're Going and Why

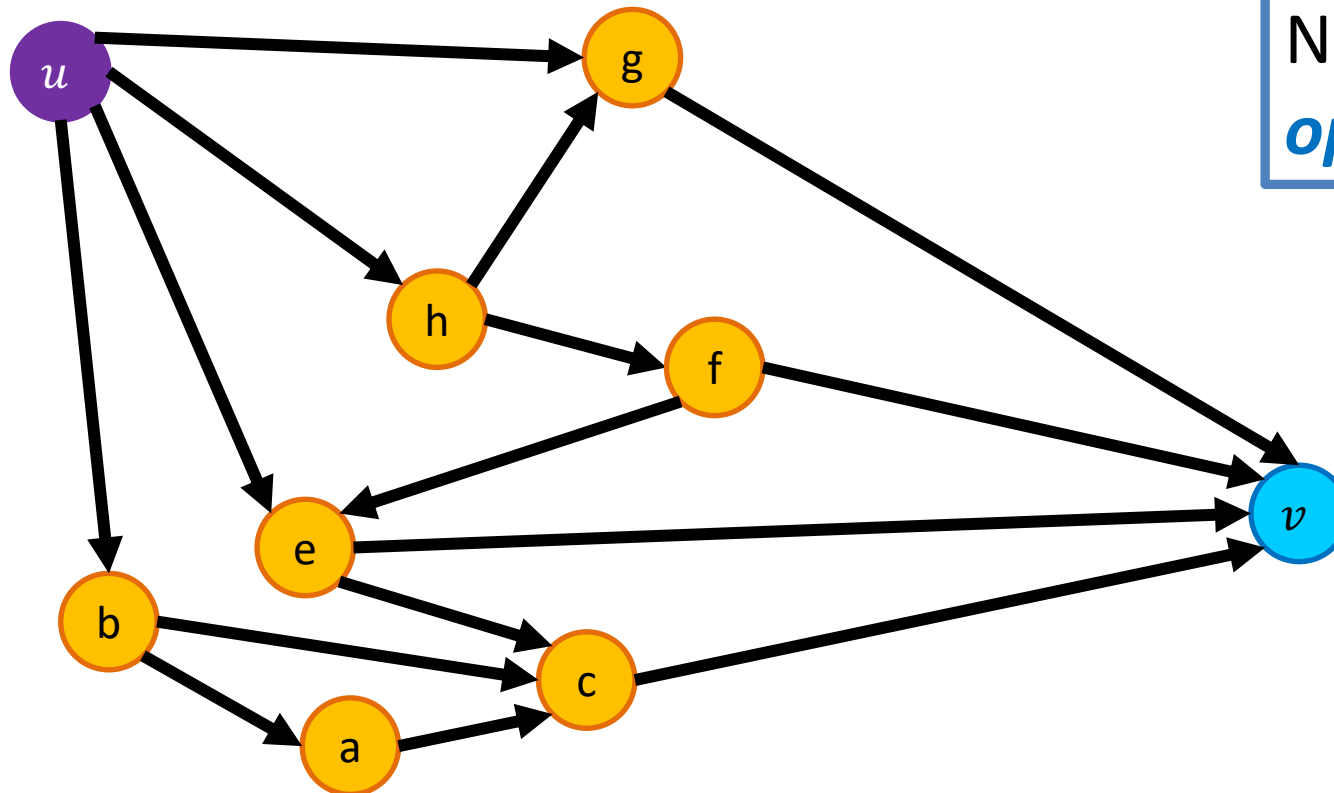
- **Reductions** between problems
 - Why? Can be a practical way of solving a new problem
 - Also: A proof about one problem's complexity can be applied to another
 - Formal definition of a reduction
- Examples
 - Bipartite graphs, matching
 - Vertex cover and independent set

Using One Solution to Solve Something Else

- Sometimes we can solve a “new” problem using a solution to another problem
 - We need to “re-cast” the “new” problem as an *instance* of the other problem
 - We may need to relate how the answer found for the other problem gives the answer for the “new” problem
- Some examples coming in this lecture:
 - We’ll see how to solve *edge-disjoint path* problem.
Use that to solve *vertex-disjoint path* problem.
 - We know how to find *max network flow*.
Use that to solve *bi-partite matching*.

Edge-Disjoint Paths

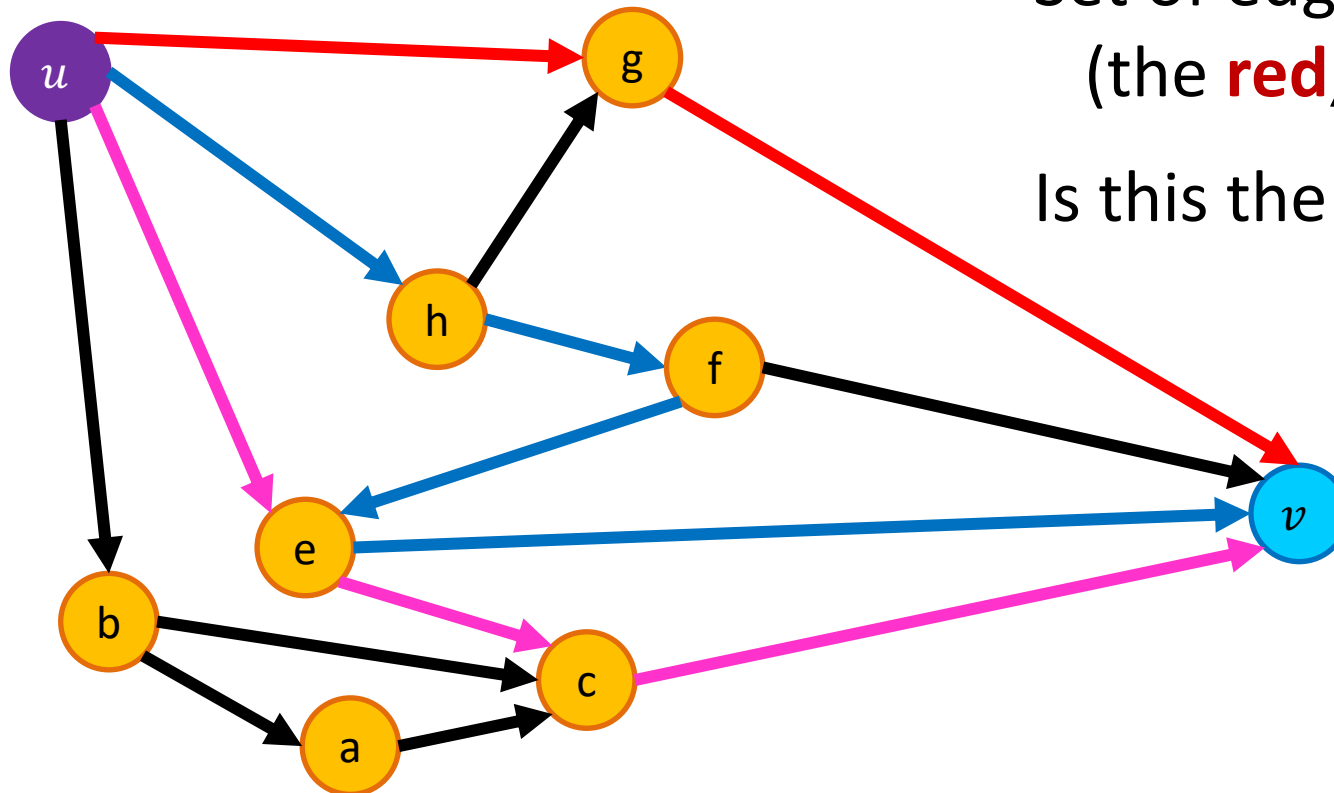
Given a graph $G = (V, E)$, a start node u and a destination node v , give the maximum number of paths from u to v which share no edges



Note this is an
optimization problem.

Edge-Disjoint Paths

Given a graph $G = (V, E)$, a start node u and a destination node v , give the maximum number of paths from u to v which share no edges

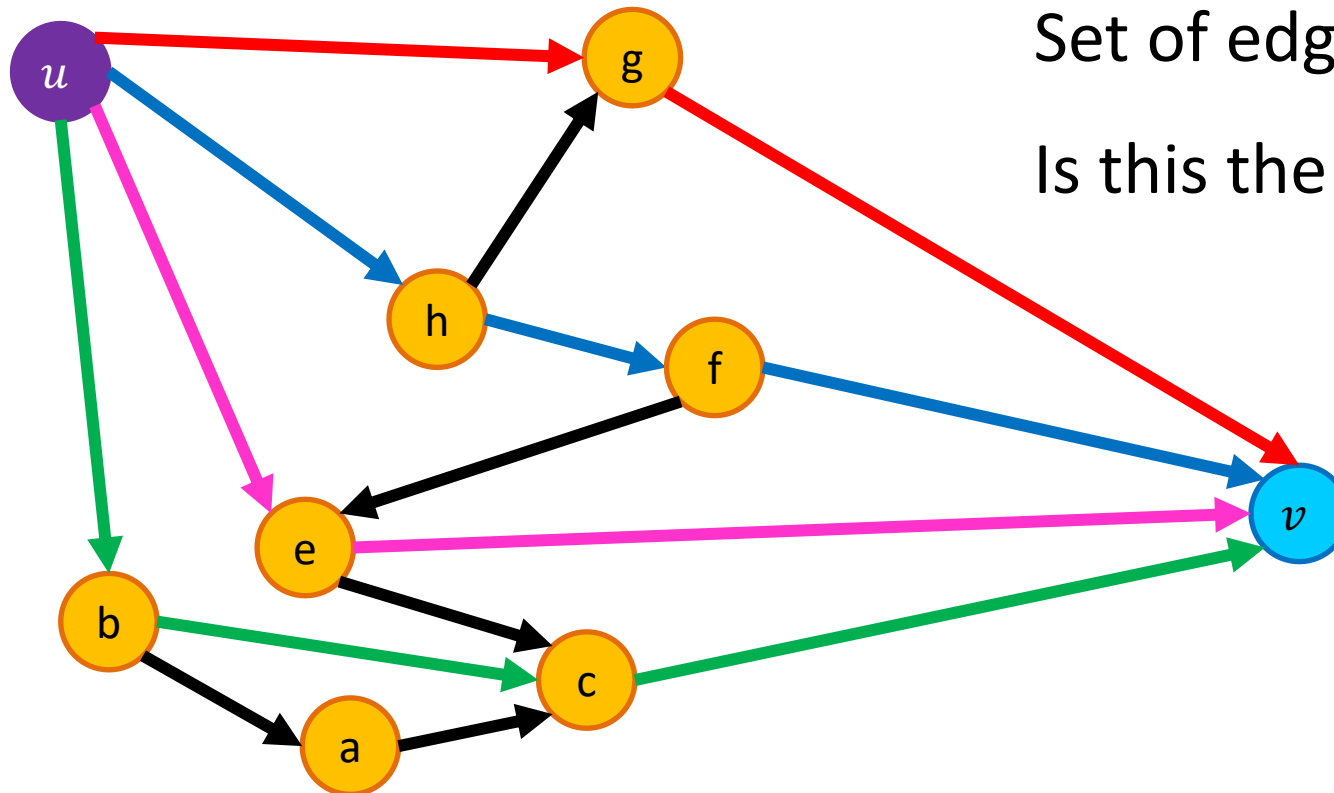


Set of edge-disjoint paths of size 3
(the **red**, **blue**, **magenta** paths)

Is this the max number?

Edge-Disjoint Paths

Given a graph $G = (V, E)$, a start node u and a destination node v , give the maximum number of paths from u to v which share no edges

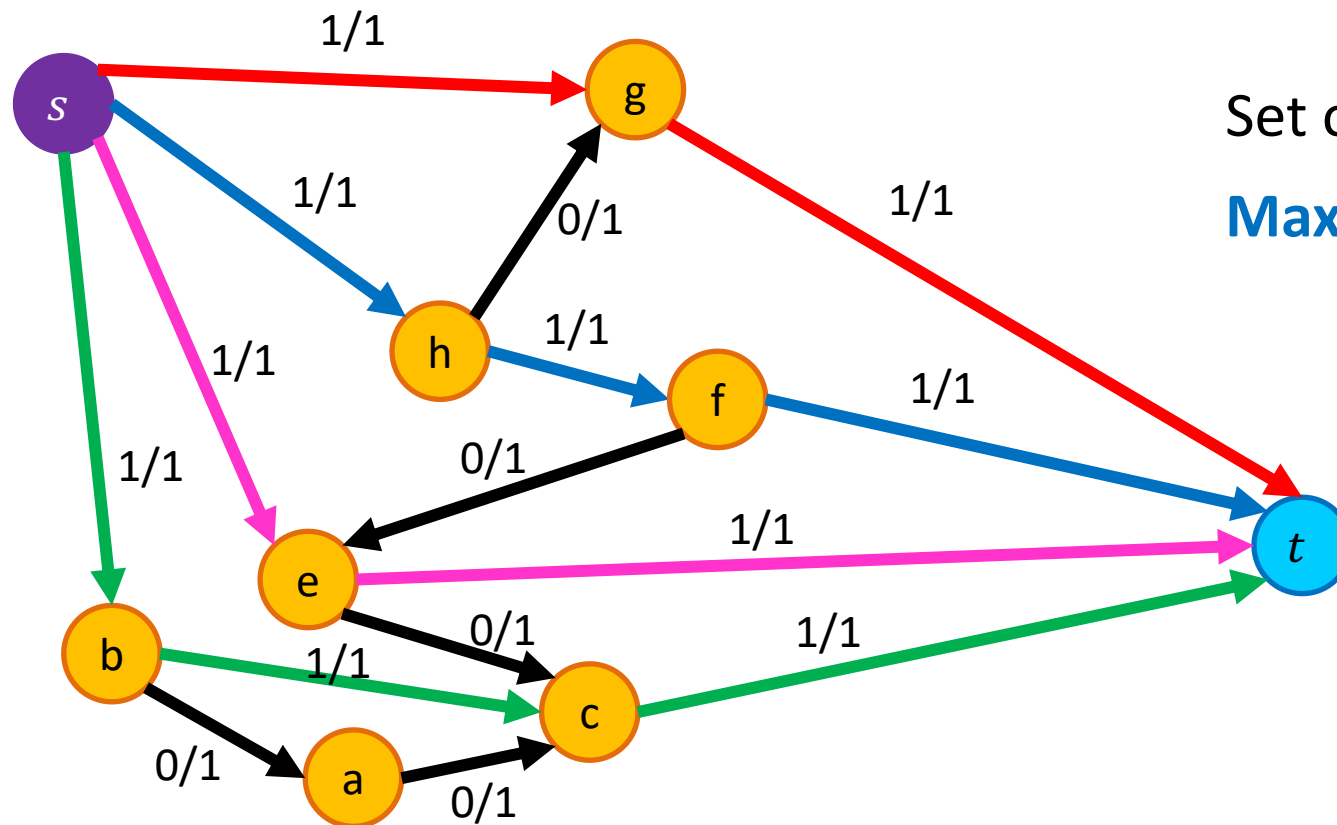


Set of edge-disjoint paths of size 4
Is this the max number?

Edge-Disjoint Paths Algorithm

Use a problem we know how to solve, *max network flow*, to solve this!

Make u and v the source and sink, give each edge capacity 1, find the max flow.



Set of edge-disjoint paths of size 4
Max flow = 4

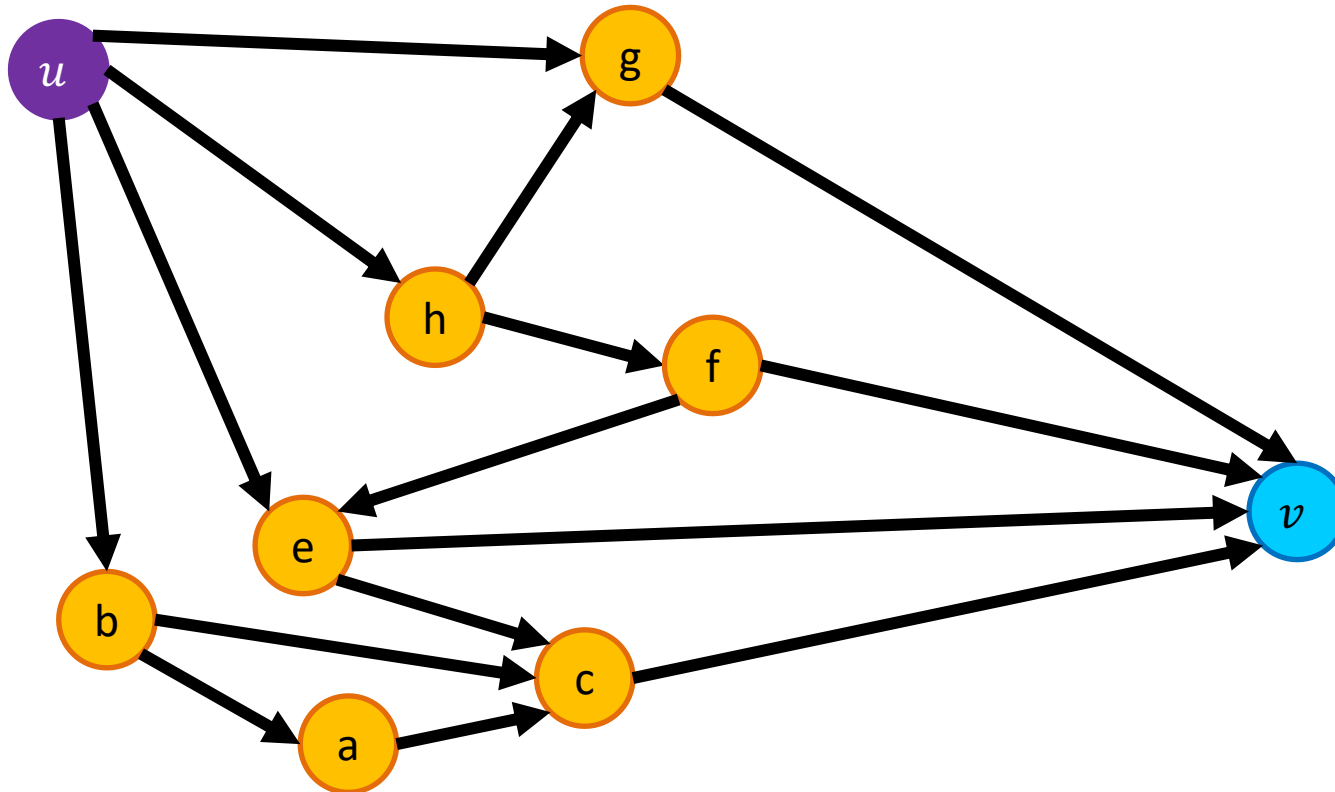
Why does this work?
We need to be able to
make a valid argument
that it always does.

What's the situation?

- Given an input I_1 for the *max network flow problem* (graph G with edge capacities), we can find the max flow for that input
- Given an input I_2 for *edge-disjoint path problem*, we can:
 - Convert that input I_2 to make a valid input I_1 for *network flow problem*, by using same graph G but adding capacity=1 for each edge
 - Solve *max network flow problem* for I_1 and get result R_1
 - Use R_1 to give the solution R_2 for *edge-disjoint path* for input I_2
 - In this case, $|f|$ = the number of paths
- Next, let's solve another problem using our new *edge-disjoint path* solution

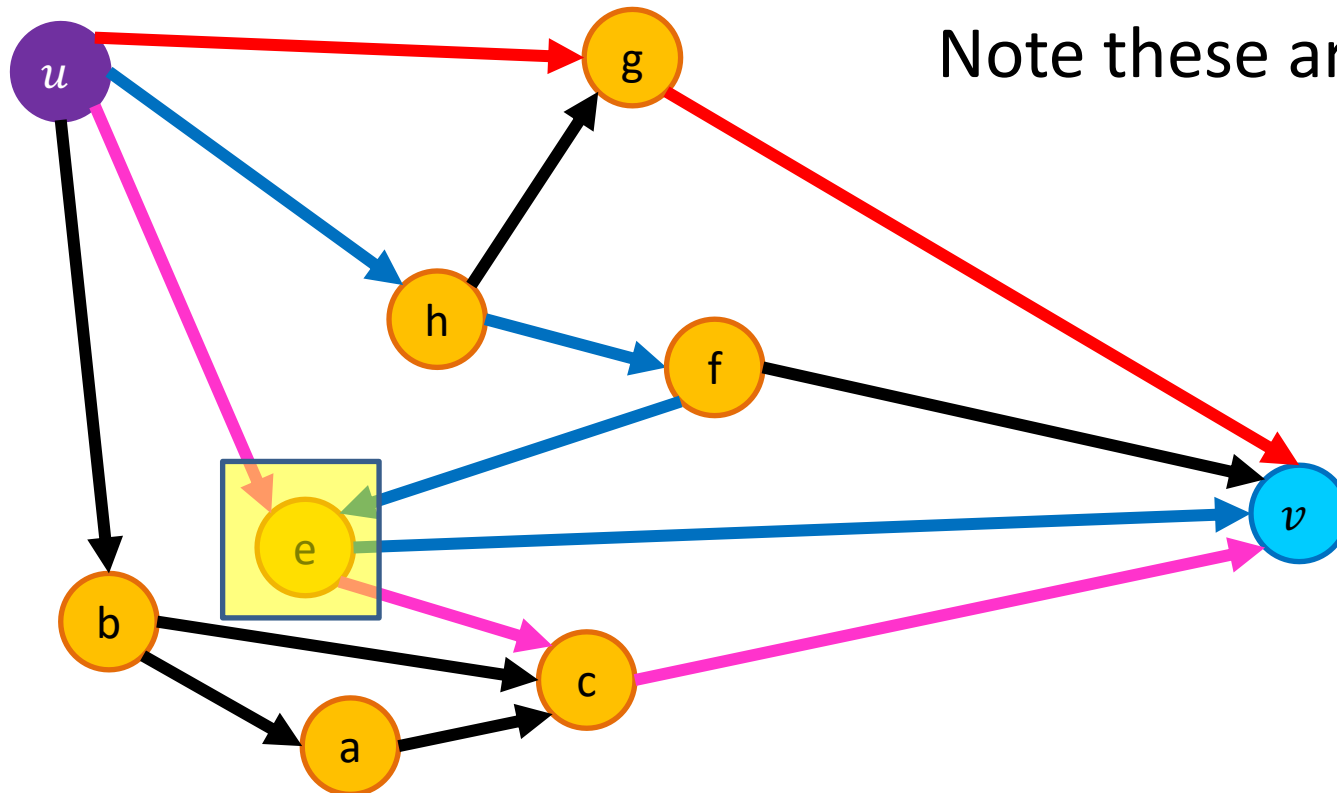
Vertex-Disjoint Paths

Given a graph $G = (V, E)$, a start node u and a destination node v , give the maximum number of paths from u to v which share no vertices



Vertex-Disjoint Paths

Given a graph $G = (V, E)$, a start node u and a destination node v , give the maximum number of paths from u to v which share no vertices

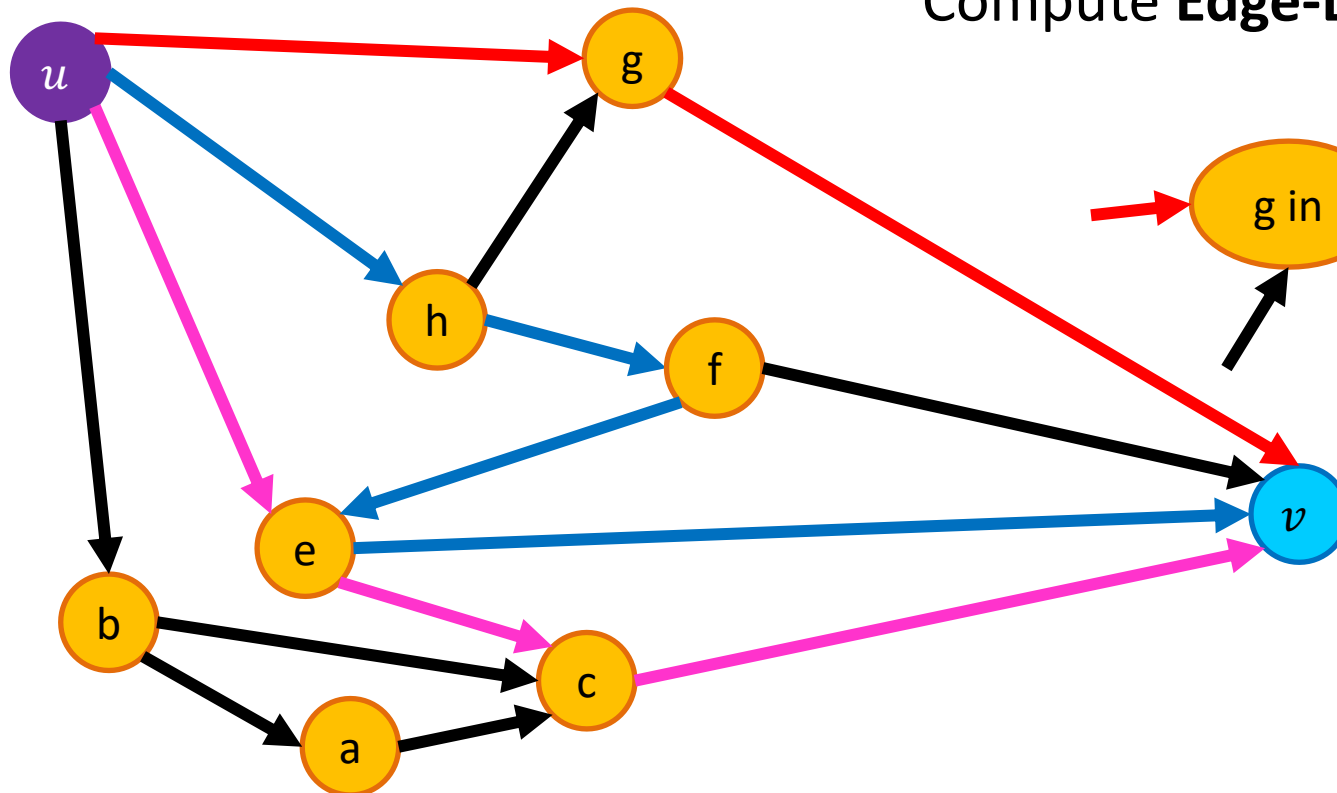


This shows 3 edge-disjoint paths.
Note these aren't vertex-disjoint paths!

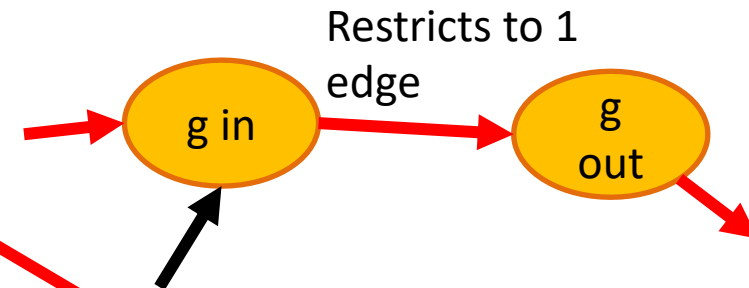
Vertex-Disjoint Paths Algorithm

Idea: Convert an instance of the vertex-disjoint paths problem into an instance of edge-disjoint paths

Make two copies of each node, one connected to incoming edges, the other to outgoing edges



Compute **Edge-Disjoint Paths** on new graph



Why does this work?
We need to be able to
make a valid argument
that it always does.

What's the situation now?

- Given an input I_1 for the **max network flow problem** (graph G with edge capacities), we can find max flow for that input
- Given an input I_2 for **edge-disjoint path problem**, we can:
 - Convert that input I_2 to make a valid input I_1 for **network flow problem**, and solve that to find **number of edge-disjoint paths**
- Given an input I_3 for **vertex-disjoint path problem**, we can:
 - Convert that input I_3 to make a valid input I_2 for **edge-disjoint path problem**
 - See above! Convert I_2 to I_1 and solve **max network flow problem**
- This chain of “problem conversions” finds lets us solve **vertex-disjoint path problem**
 - **Time complexity?** Cost of solving max network flow plus two conversions

Reductions

(We're about to get interested in problems that seem to require exponential time...)

Max-flow vs. min-cut

- These two problems are “equivalent”
 - Remember? *max-flow min-cut theorem*
 - Here we’re saying: if you can solve one, you can solve the other
- Alternatively, we can say that one problem *reduces* to the other
 - The problem of finding min-cut *reduces to* the problem of finding max-flow
 - Maybe this *reduction* requires some work to “convert”
 - Could be nothing or minimal
 - For these problems, the cost of the conversion is *polynomial*

Reduction

- A **reduction** is a transformation of one problem into another problem
 - Min-cut is reducible to max-flow because we can use max-flow to solve min-cut
 - Formally, problem A is **reducible** to problem B if we can use a solution to B to solve A
- We're particularly interested in reductions that happen in *polynomial time*
- If A is **polynomial-time reducible** to B, we denote this as:
$$A \leq_p B$$

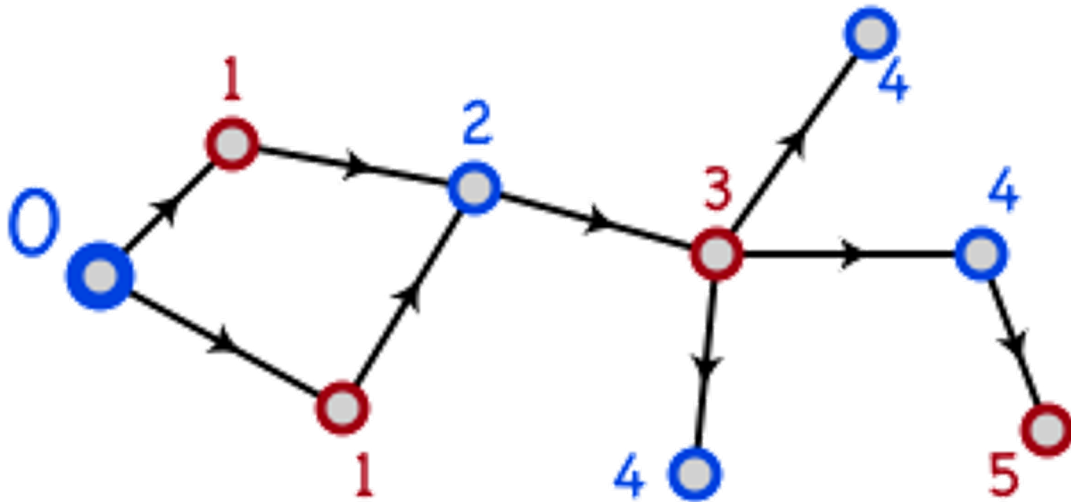
Reducing both ways

- It's easy to see that:
 - $\text{Min-cut} \leq_p \text{max-flow}$
 - $\text{Max-flow} \leq_p \text{min-cut}$
- Because they reduce both ways, they are *polynomial-time equivalent*
 - If we find a polynomial solution for one, the other is also polynomial
 - What if we prove an exponential lower-bound for one?
Is it possible that the other one could have a polynomial solution?

Bipartite Matching

Bipartite Graphs

- A graph is *bipartite* if node-set V can be split into sets X and Y such that every edge has one end in X and one end in Y
 - X and Y could be colored red and blue
 - Or Boolean true/false



How to determine if G is bipartite?

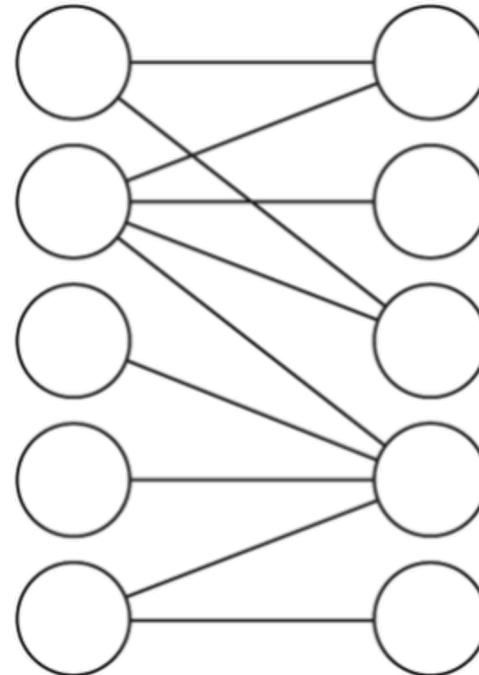
The numbers and arrows on edges may give you a clue....

BFS or DFS, and label nodes by levels in tree.

Non-tree edge to node with same label means NOT bipartite.

Notes and assumptions

- We assume the graph is connected
 - Otherwise we will only look at each connected component individually
- A triangle cannot be bipartite
 - In fact, any graph with an odd length cycle cannot be bipartite



Maximum Bipartite Matching

Dog Lovers

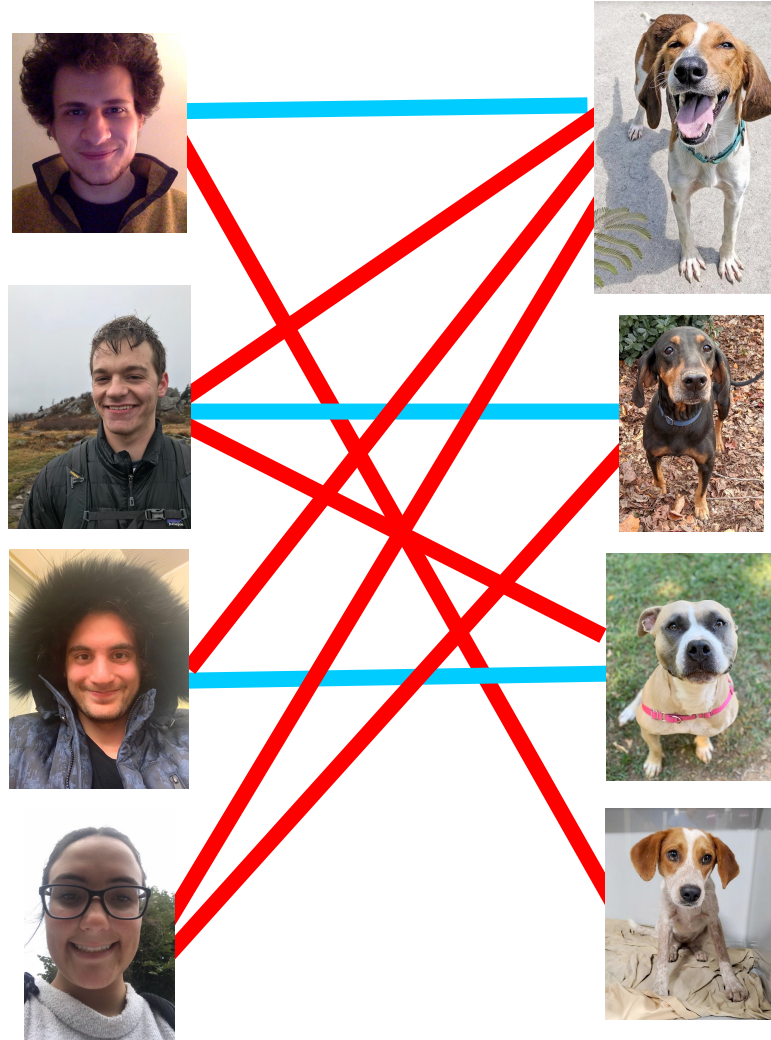
Adoptable Dogs



Maximum Bipartite Matching

Dog Lovers

Adoptable Dogs

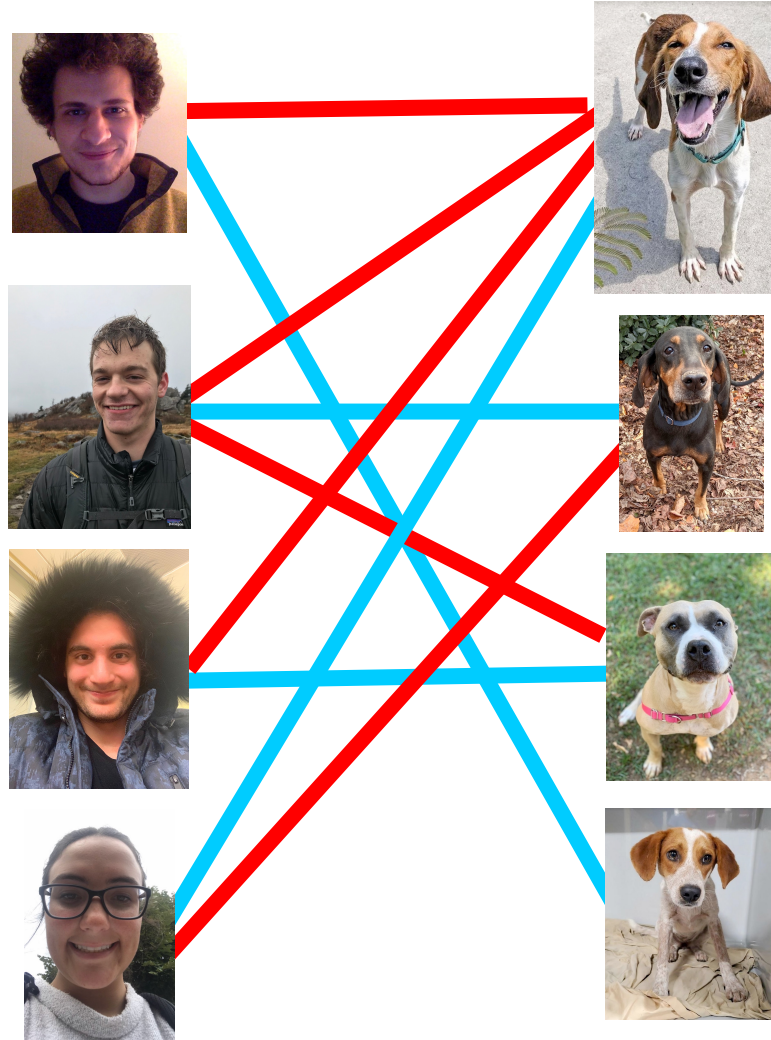


Is this the best possible?
The largest possible set
of edges?

Maximum Bipartite Matching

Dog Lovers

Adoptable Dogs



Better! In fact, the maximum possible!
How can we tell?

A *perfect bipartite match*:
Equal-sized left and right subsets, and all nodes have a matching edge

Maximum Bipartite Matching

Given a graph $G = (L, R, E)$

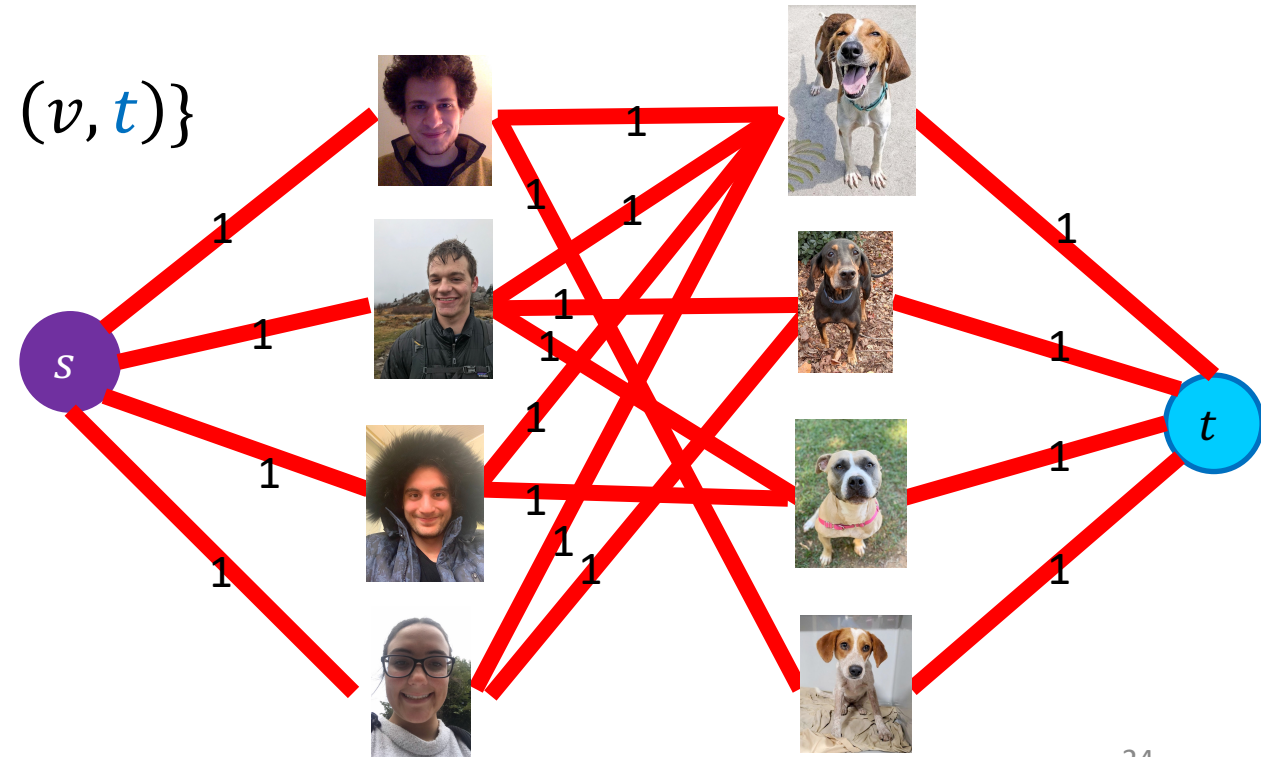
a set of left nodes, right nodes, and edges between left and right

Find the largest set of edges $M \subseteq E$ such that each node $u \in L$ or $v \in R$ is incident to at most one edge.

Maximum Bipartite Matching Using Max Flow

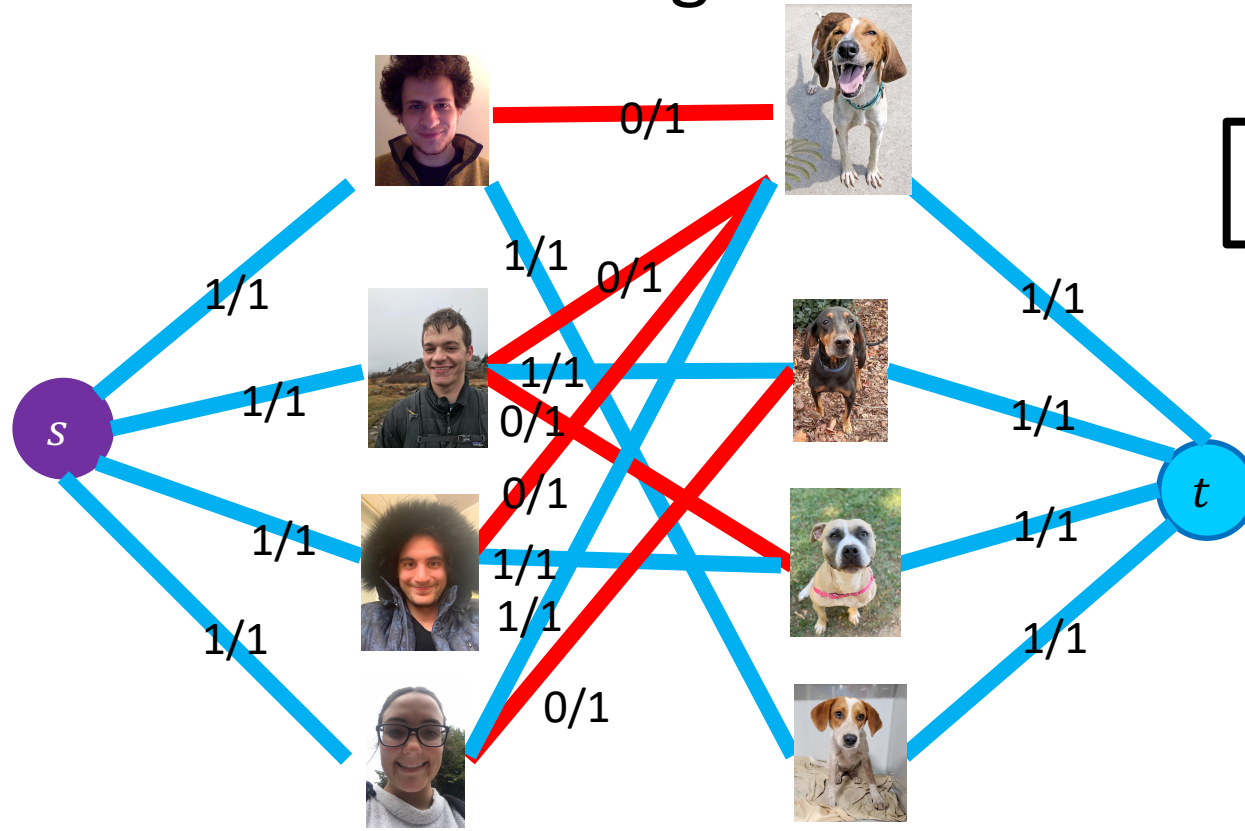
Make $G = (L, R, E)$ a flow network $G' = (V', E')$ by:

- Adding in a **source** and **sink** to the set of nodes:
 - $V' = L \cup R \cup \{s, t\}$
- Adding an edge from **source** to L and from R to **sink**:
 - $E' = E \cup \{u \in L \mid (s, u)\} \cup \{v \in r \mid (v, t)\}$
- Make each edge capacity 1:
 - $\forall e \in E', c(e) = 1$



Maximum Bipartite Matching Using Max Flow

1. Make G into G' $\Theta(L + R)$
2. Compute Max Flow on G' $\Theta(E \cdot V) \quad |f| \leq L$
3. Return M as all “middle” edges with flow 1 $\Theta(L + R)$



Overall: $\Theta(E \cdot V)$

Roadmap: Where We've Been and Why

- **Reductions** between problems
 - Why? Can be a practical way of solving a new problem
 - **Coming soon:** A proof about one problem's complexity can be applied to another
 - Formal definition of a reduction
- Examples
 - Bipartite graphs, matching
- **Next:** example problems: vertex cover and independent set
 - Then, classes of problems: P, NP, NP-Hard, NP-complete