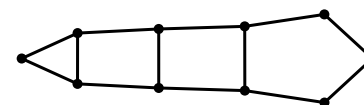
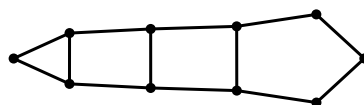
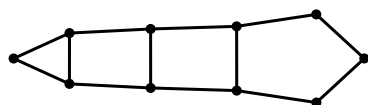


# Matchings in Graphs

*Definition.* A **matching**  $M$  in a graph  $G$  is a subset of edges of  $G$  that share no vertices.

*Definition.* A **maximal matching**  $M$  is a matching such that the inclusion into  $M$  of any edge of  $G \setminus M$  is no longer a matching.

*Definition.* A **maximum matching** is a matching  $M$  that has the most edges possible for the graph  $G$ .



*Definition.* A **perfect matching** is a matching involving **every vertex**.

*Thought Exercise:* What is the result of overlapping two matchings?

## Minimum vs. Minimal

We have just had two definitions related to the concept of “largest” .  
An important concept is the distinction between  
**maximum** and **maximal**.

**Maximum** refers to an element of absolute largest size.  
(of *ALL* elts with *property*, this is largest.)

**Maximal** refers to an element of relative largest size.  
(for *THIS* elt with *property*, no subset has *property*.)

*Example.* maximal vs. maximal path in a graph:

*Example.* maximum vs. maximal clique in a graph:

## Application: Scheduling

Suppose you are working in a group trying to complete all the problems on the homework. Depending on everyone's preferences, you would like to assign each member one problem to do.

Person A likes problems 1, 2, 3, and 5.

Person B likes problems 1, 2, and 4.

Person C likes problems 3, 4, and 5.

Person D likes problems 2 and 3.

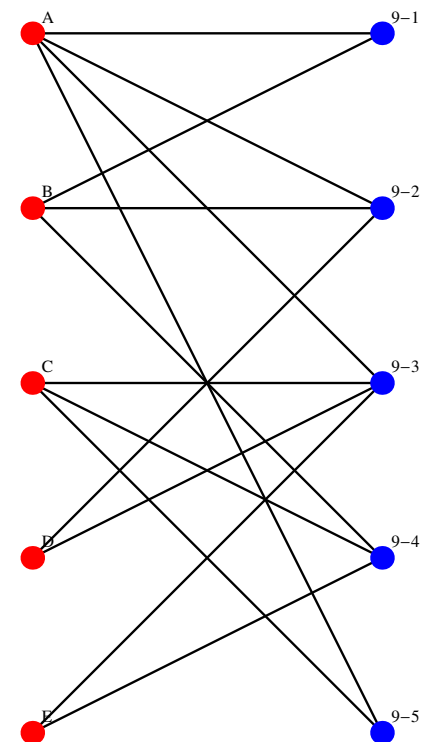
Person E likes problems 3 and 4.

Create a graph that models the situation.

*Question.*

What is a maximum matching for this graph?

We will use an algorithm to answer this question.



# Algorithms

*Definition.* An **algorithm** is a set of rules followed to solve a problem.

In general, an algorithm has the steps: Havel–Hakimi:

1. Organize the input.
2. Repeatedly apply some steps until a termination condition holds
3. Analyze data upon termination

Computers can be used to run the algorithms *once we verify they work*.

To verify the **correctness** of an algorithm:

1. Verify that the algorithm terminates. (often invoking finiteness)
2. Verify that the result satisfies the desired conditions.

# Motivating The Hungarian Algorithm

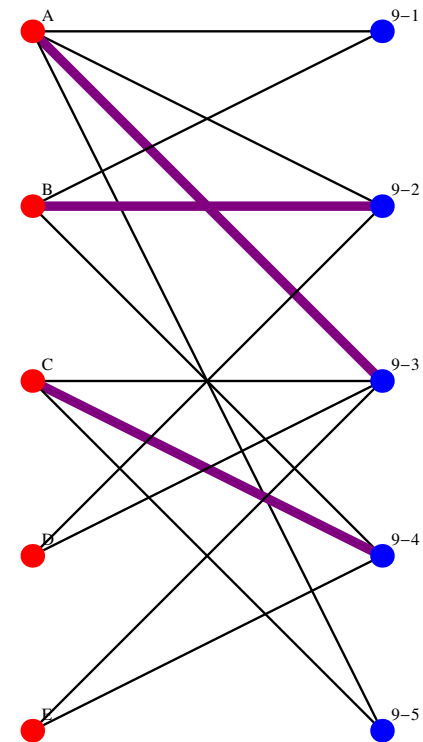
Let us work through the basic idea behind the algorithm. We start with an initial matching; we might as well make it maximal. Why is the pictured matching maximal?

*Definition.* Given a matching  $M$  in a graph  $G$ , an  $M$ -**alternating path** is a path in  $G$  that starts at a vertex not in  $M$ , and whose edges alternate between being in  $M$  and not in  $M$ .

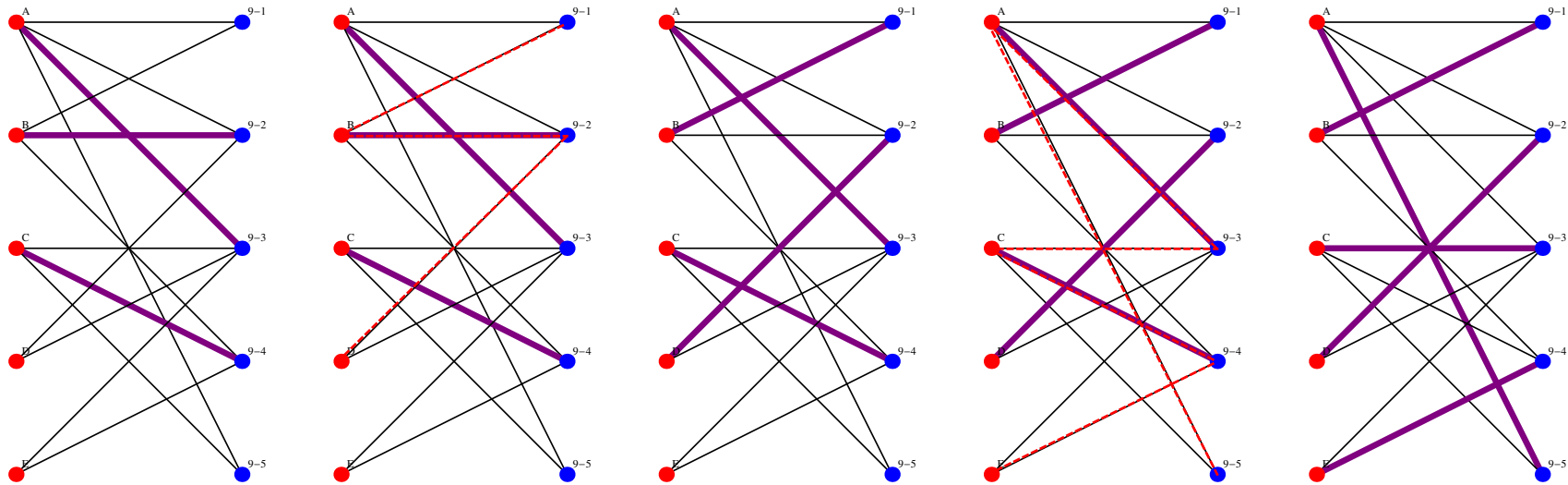
*Example.*  $D \rightarrow 2 \rightarrow B \rightarrow 4 \rightarrow C$  is an  $M$ -alternating path.

*Definition.* An  $M$ -**augmenting path** is an  $M$ -alternating path that begins AND ends at unmatched vertices.

It is **augmenting** because we can improve  $M$  by toggling the edges between those in  $M$  and those not in  $M$ .



# Motivating The Hungarian Algorithm



Given  $M$ ,  $P = D \rightarrow 2 \rightarrow B \rightarrow 1$  is an  $M$ -augmenting path.  
Toggling the edges in  $P$  gives a new matching  $M'$ .

Given  $M'$ ,  $P' = E \rightarrow 4 \rightarrow C \rightarrow 3 \rightarrow A \rightarrow 5$  is an  $M'$ -augmenting path. Toggling the edges in  $P'$  gives a new matching  $M''$ .

The matching  $M''$  is maximal. (Why?)

# The Hungarian Algorithm

*The Hungarian Algorithm* (Kuhn, König, Egeváry) [*Finds a maximum matching in a bipartite graph (w/red and blue vertices)*]

1. Start with a bipartite graph  $G$  and any matching  $M$ .  
Label all red vertices *eligible* (for augmentation).
2. If all red, eligible vertices are matched, stop. Otherwise, there exists a red, unmatched, eligible vertex to use in the next step.
3. Let  $v$  be an unmatched, eligible, red vertex. Start growing all possible  $M$ -alternating paths from  $v$ . That is, follow every edge not in  $M$  to a blue vertex. From a matched blue vertex, follow the edge of  $M$  back to a red vertex, and repeat as far as possible.  
 $\left\{ \begin{array}{l} \text{If there is an } M\text{-augmenting path, toggle edges to augment } M. \\ \text{If there is no } M\text{-augmenting path, mark } a \text{ ineligible.} \end{array} \right.$

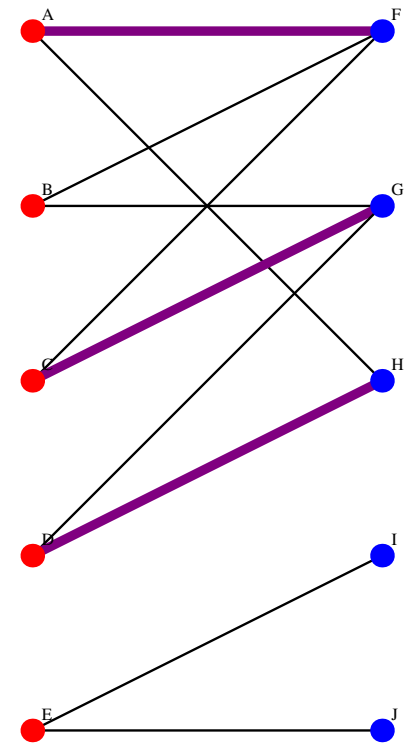
Return to Step 2.

# Applying the Hungarian Algorithm

Here is something that might happen during an application of the Hungarian algorithm:

*Example.* There is no  $M$ -augmenting path starting at  $B$  in the graph to the right.

We would mark  $B$  ineligible and move on to the next eligible, unmatched red vertex in the graph ( $E$ ).





## Proof of Correctness

*Claim.* The Hungarian Algorithm gives a maximum matching.

*Proof.* We must show that the algorithm always stops, and that when it stops, the output is indeed a maximum matching.

**The algorithm terminates.** Each time Step 3 is run, one red vertex either becomes matched or becomes ineligible. Also, no red vertex that starts matched becomes unmatched. Since there are a finite number of red vertices, the algorithm must terminate.

**The output is a maximum matching.** The output  $M$  is a matching inducing no  $M$ -augmenting paths in the graph. Suppose that there were another matching  $M^*$  that used more edges than  $M$ .

When we overlap  $M$  and  $M^*$ , the result is a union of cycles and paths. At least one path must have more edges from  $M^*$  than  $M$ .

This path is an  $M$ -augmenting path, contradicting the definition of  $M$ .