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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.

Aside: Maximum vs. Maximal

There is an important distinction between maxim*um* and maxim*al*.

Maximum	refers to an element of <u>absolute</u> largest size. (of *ALL* elts with *property*, this is largest.)
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Example. maximum vs. maximal path in a graph:

Example. maximum vs. maximal matching: (next page)

Matchings in Graphs

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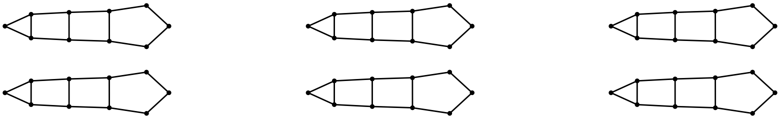

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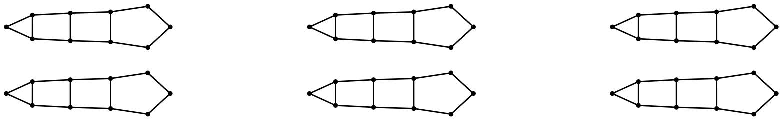

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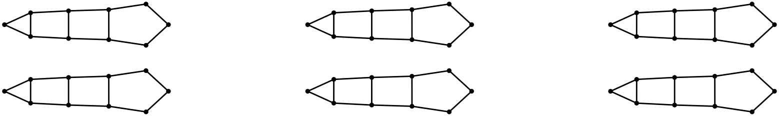

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Recall. A **perfect matching** is a matching involving every vertex of G.

$\star$ We will discuss matchings in a bipartite graph $\star$

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.
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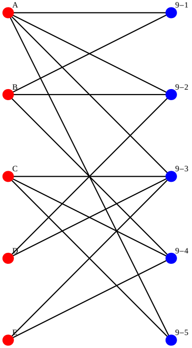

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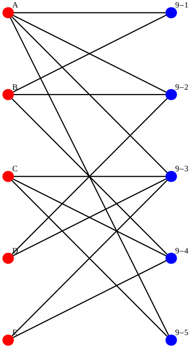


Create a graph that models the situation.

Question:
What is a maximum matching for this graph?

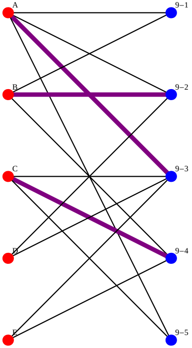
We will use an algorithm to answer this question.

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?



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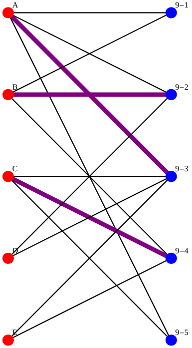

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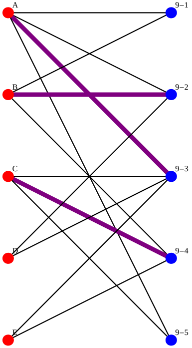

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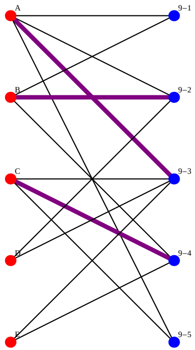

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Definition: An M-**augmenting path** is an M-alternating path that begins AND ends at unmatched vertices.

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

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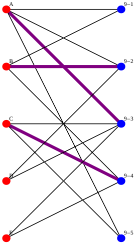

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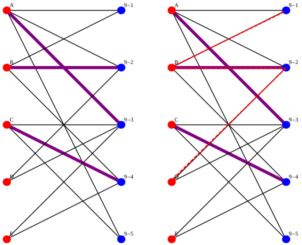

Given M, $P = D \to 2 \to B \to 1$ is an M-augmenting path.
Toggling the edges in P gives a new matching M'.

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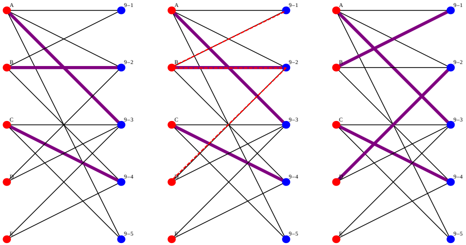

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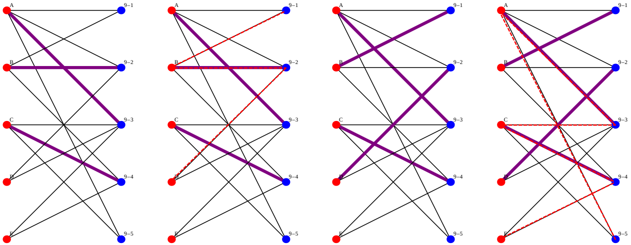

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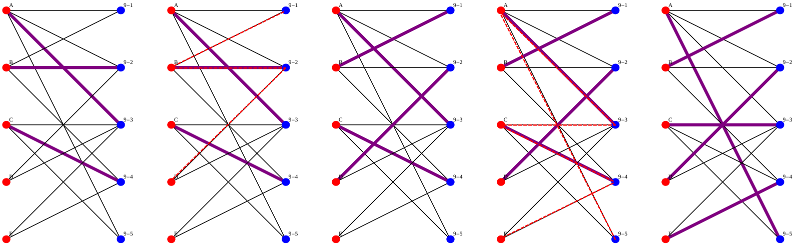

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The matching M'' is maximal. (Why?)

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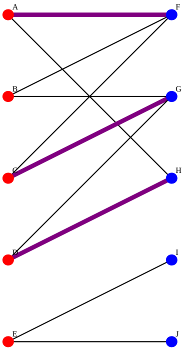
Return to Step 2.

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).

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