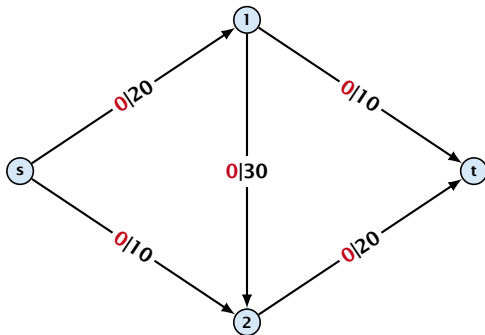


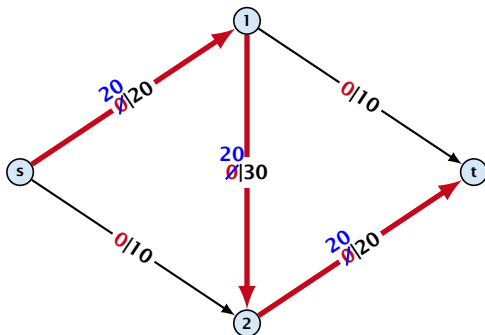
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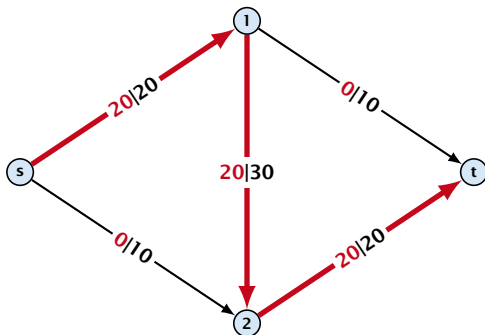
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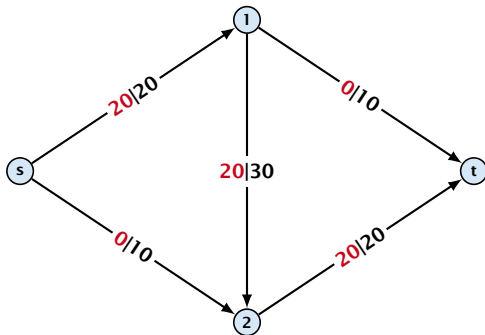
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From the graph  $G = (V, E, c)$  and the current flow  $f$  we construct an auxiliary graph  $G_f = (V, E_f, c_f)$  (the residual graph):

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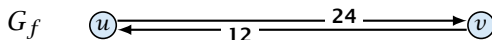
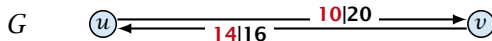
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# Augmenting Path Algorithm

## Definition 1

An **augmenting path** with respect to flow  $f$ , is a path from  $s$  to  $t$  in the auxiliary graph  $G_f$  that contains only edges with non-zero capacity.

Algorithm 1 FordFulkerson( $G = (V, E, c)$ )

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# Augmenting Path Algorithm

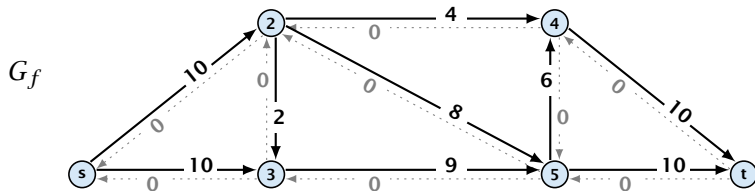
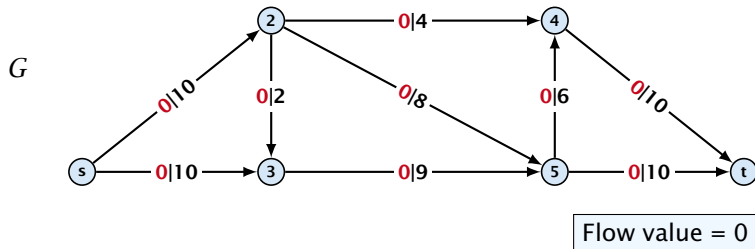
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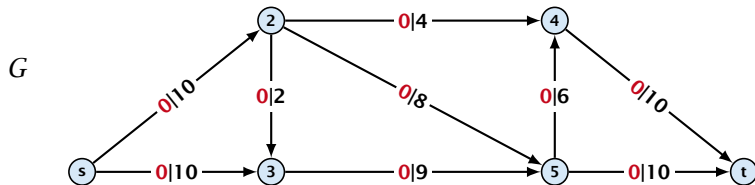
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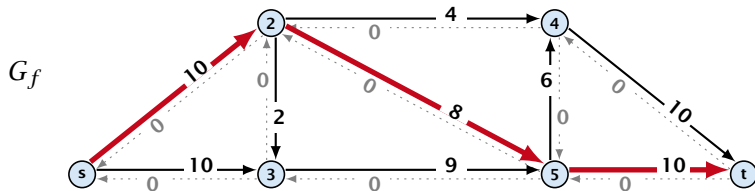
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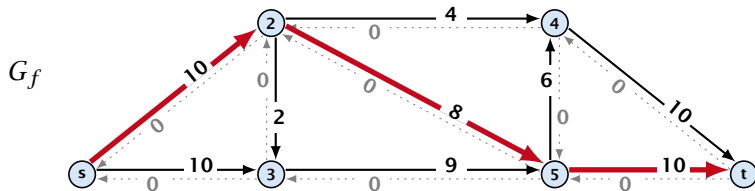
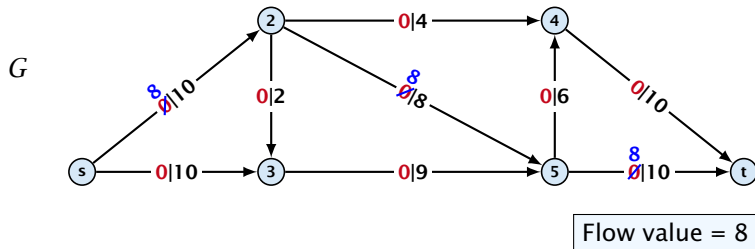
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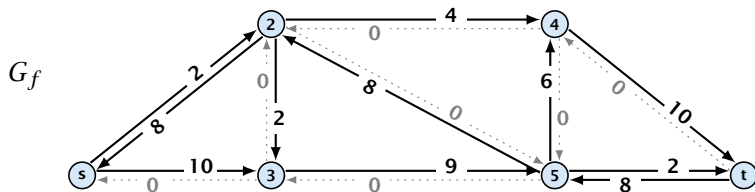
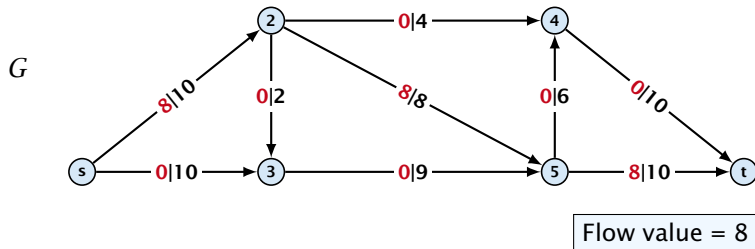
Flow value = 0



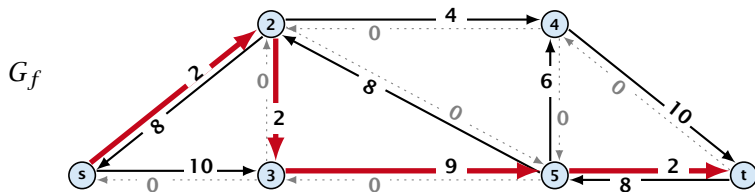
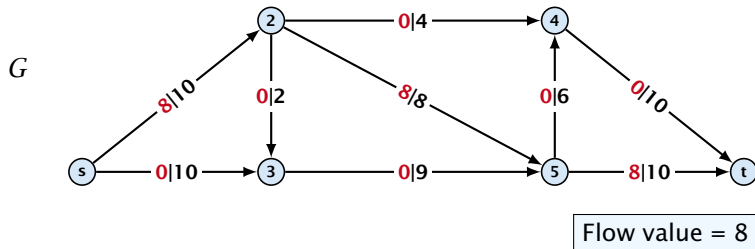
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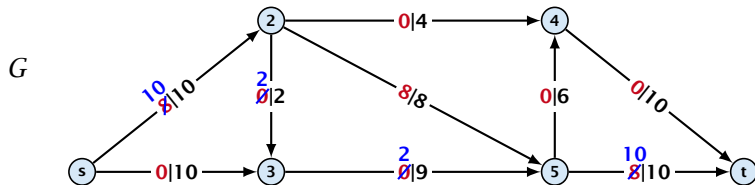
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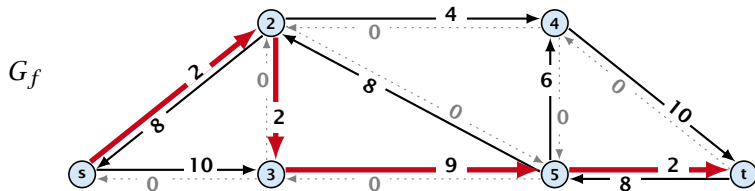
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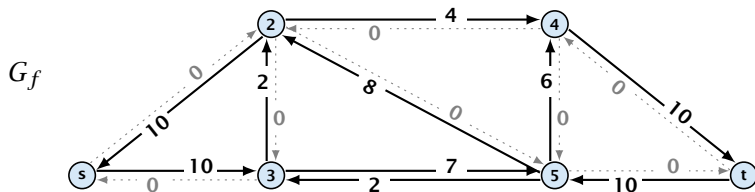
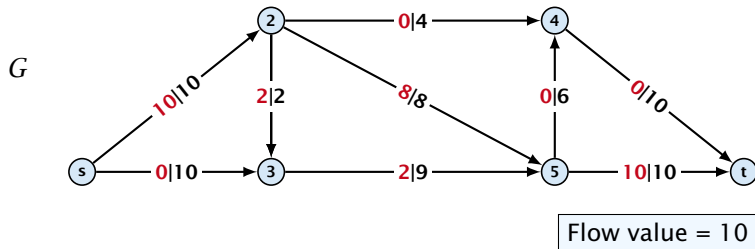


Flow value = 10

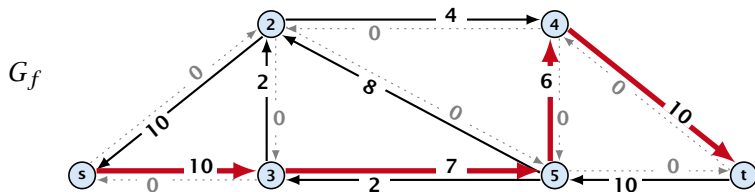
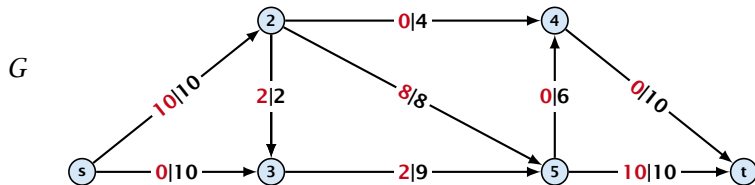




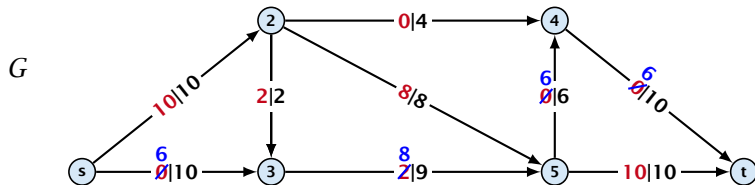
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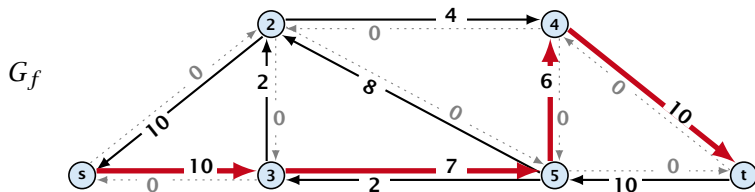
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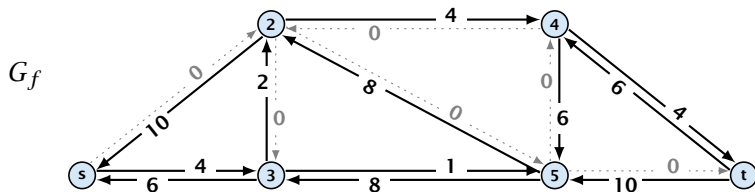
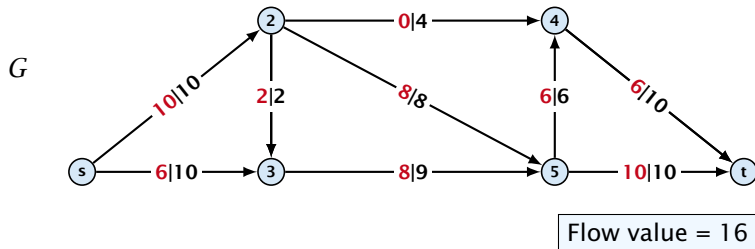
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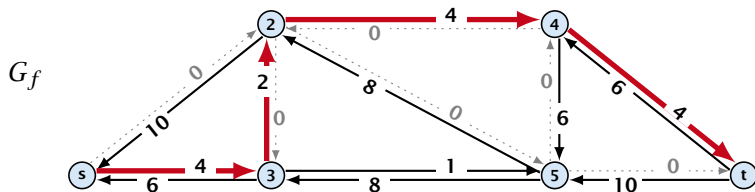
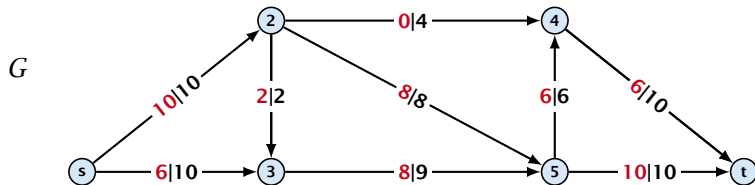
Flow value = 16



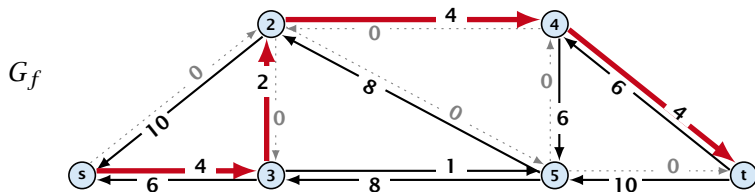
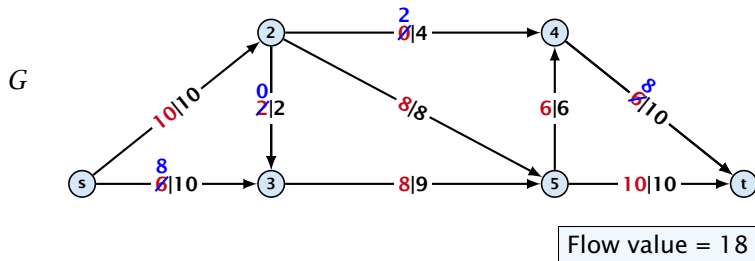
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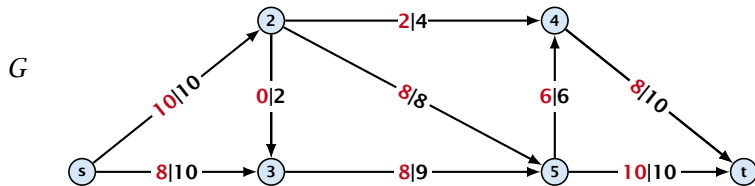
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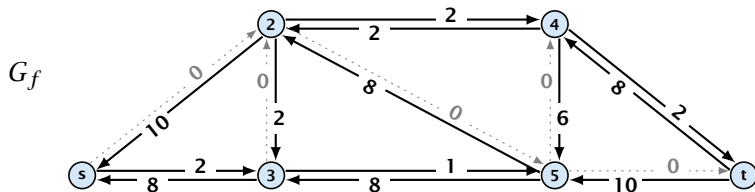
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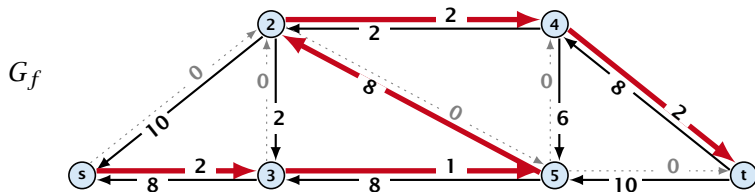
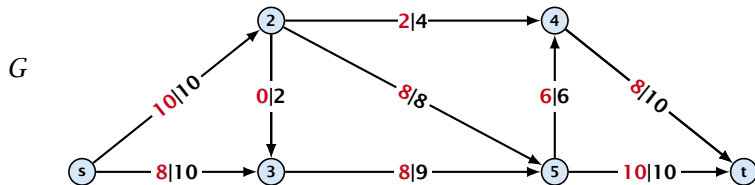
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Flow value = 18

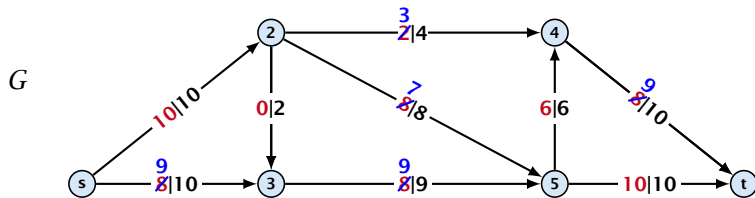


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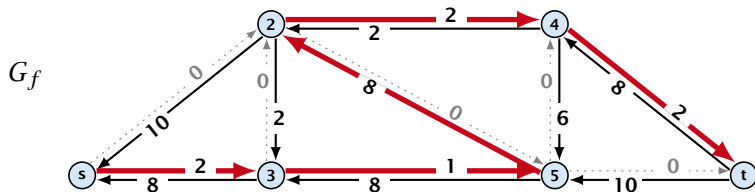




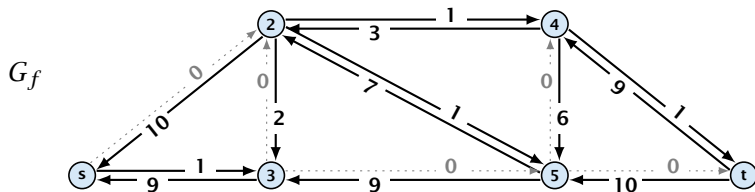
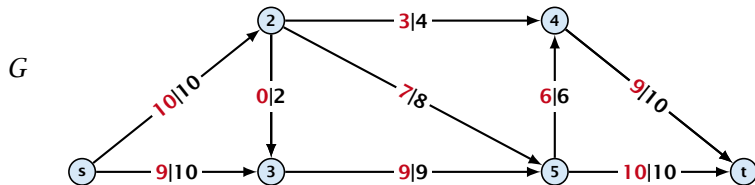
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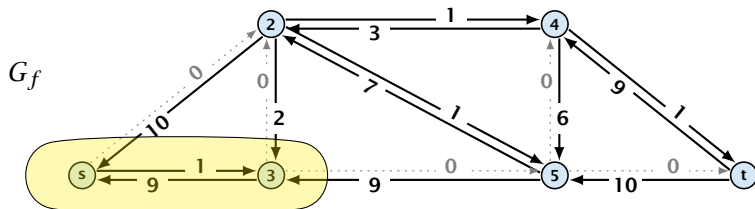
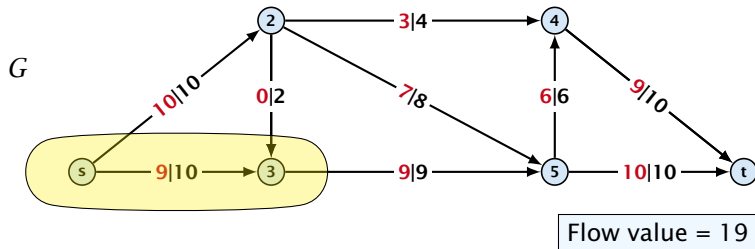
Flow value = 19



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# Augmenting Path Algorithm

## Theorem 2

*A flow  $f$  is a maximum flow iff there are no augmenting paths.*

## Theorem 3

*The value of a maximum flow is equal to the value of a minimum cut.*

## Proof.

Let  $f$  be a flow. The following are equivalent:

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This we already showed.

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If there were an augmenting path, we could improve the flow.  
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This finishes the proof.

Here the first equality uses the flow value lemma, and the second exploits the fact that the flow along incoming edges must be 0 as the residual graph does not have edges leaving  $A$ .

# Analysis

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All capacities are integers between 1 and  $C$ .

Invariant:

Every flow value  $f(e)$  and every residual capacity  $c_f(e)$  remains integral throughout the algorithm.

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The algorithm terminates in at most  $\text{val}(f^*) \leq nC$  iterations, where  $f^*$  denotes the maximum flow. Each iteration can be implemented in time  $\mathcal{O}(m)$ . This gives a total running time of  $\mathcal{O}(nmC)$ .

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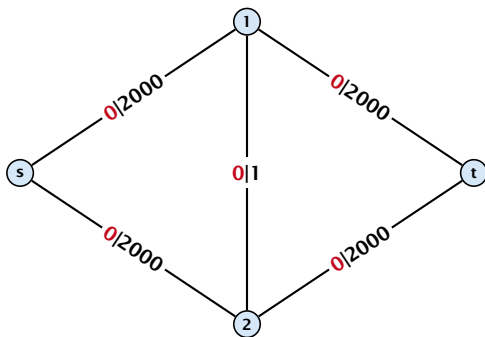
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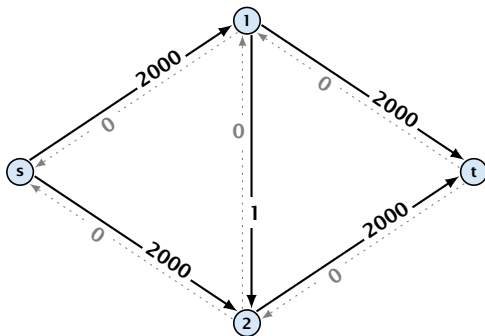
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Problem: The running time may not be polynomial.



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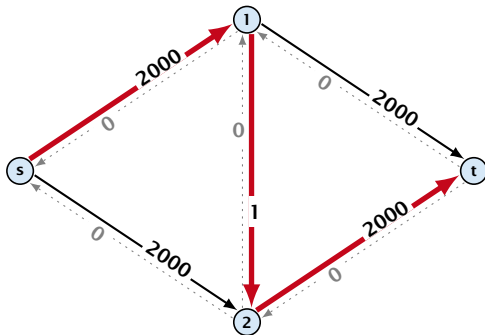


Question:

Can we tweak the algorithm so that the running time is polynomial in the input length?

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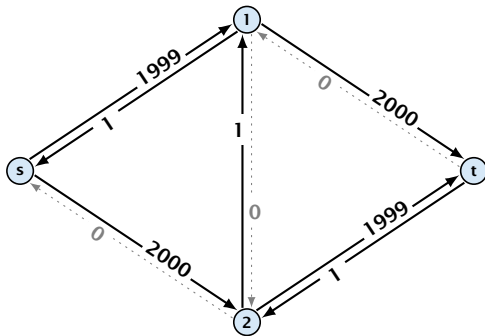


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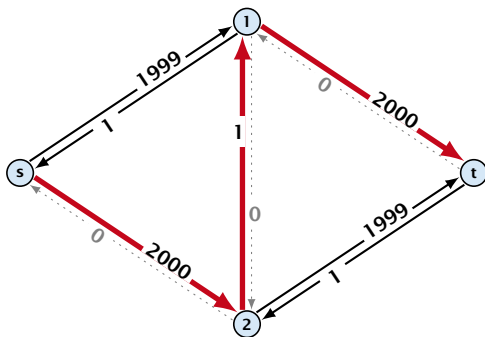


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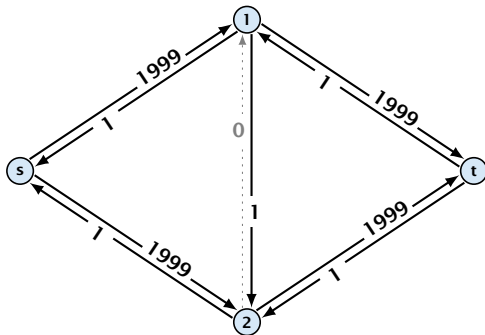


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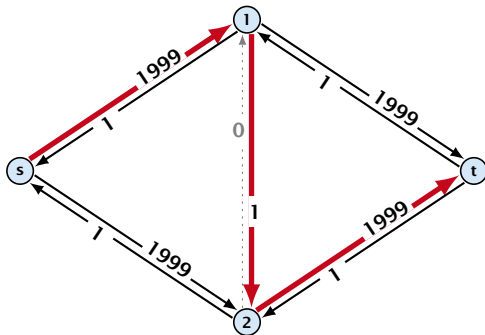


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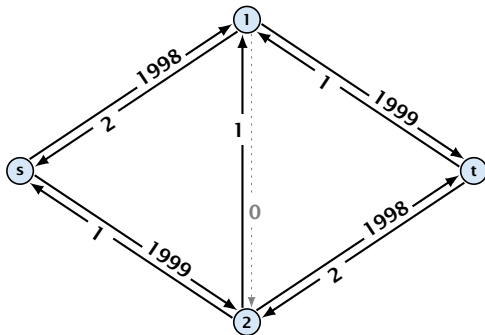


Question:

Can we tweak the algorithm so that the running time is polynomial in the input length?

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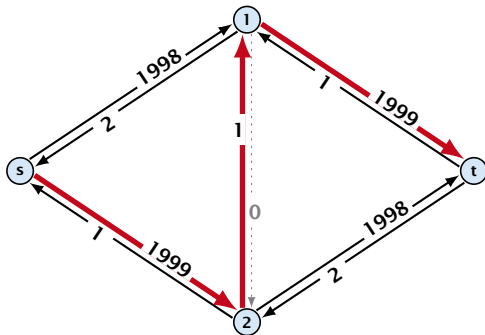
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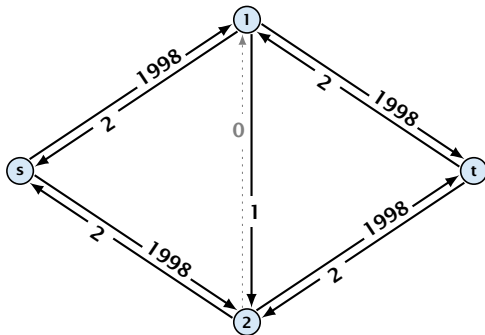


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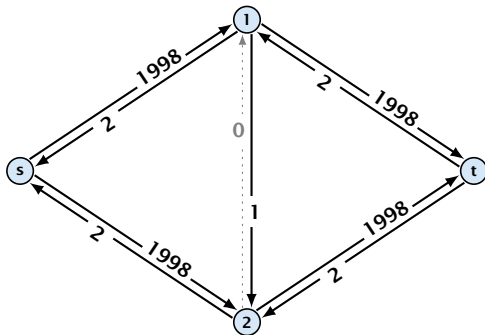


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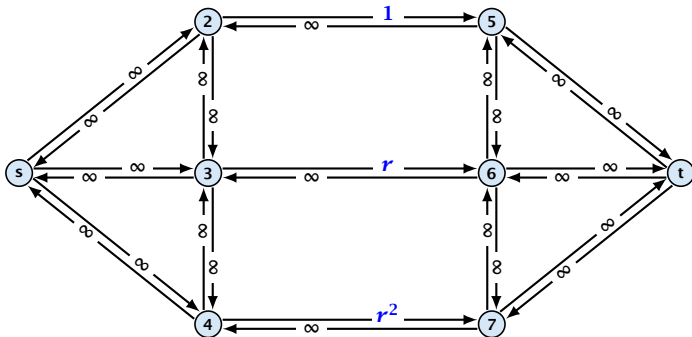


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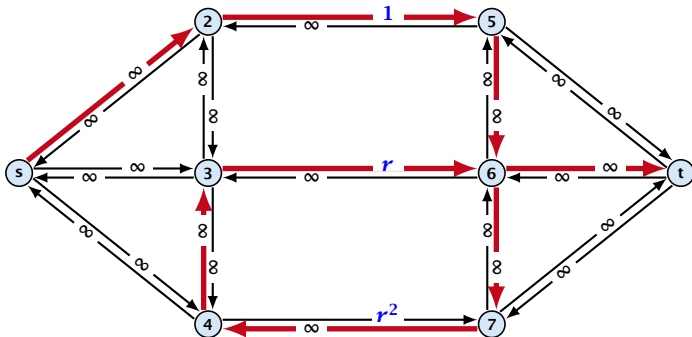
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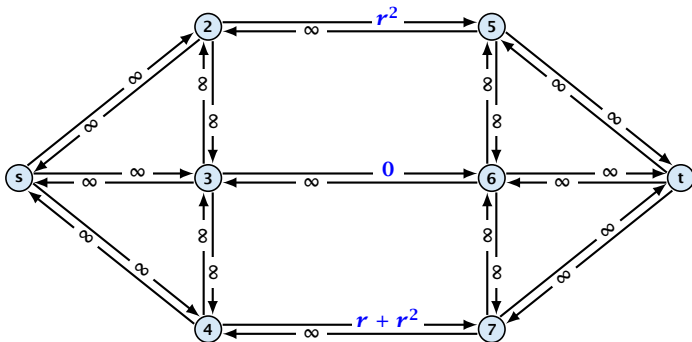
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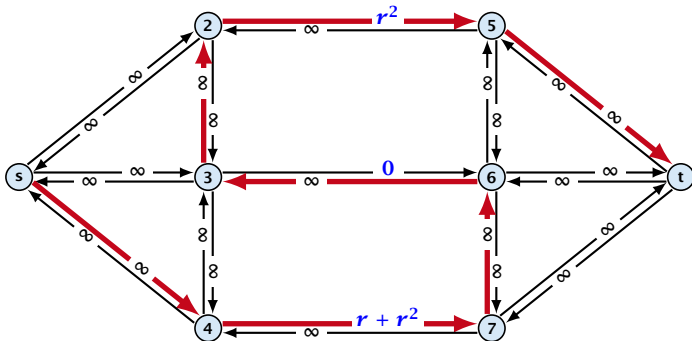
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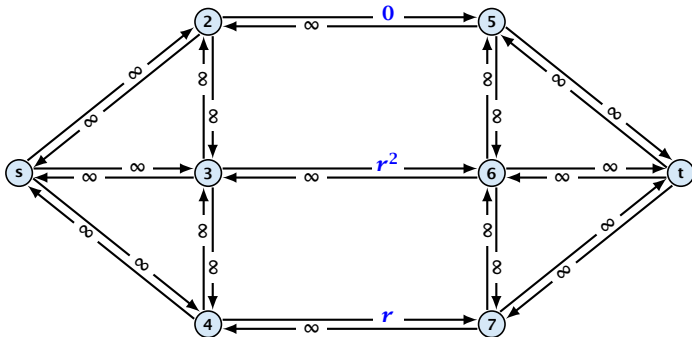
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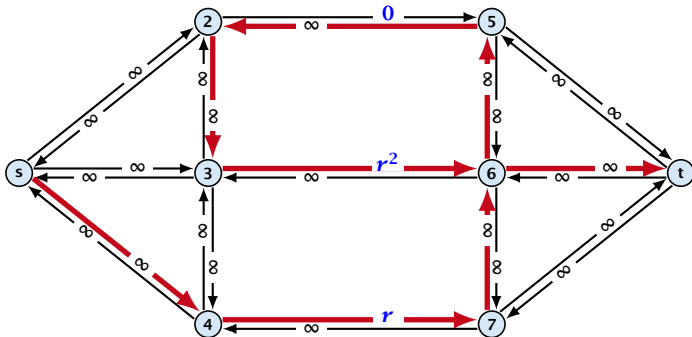
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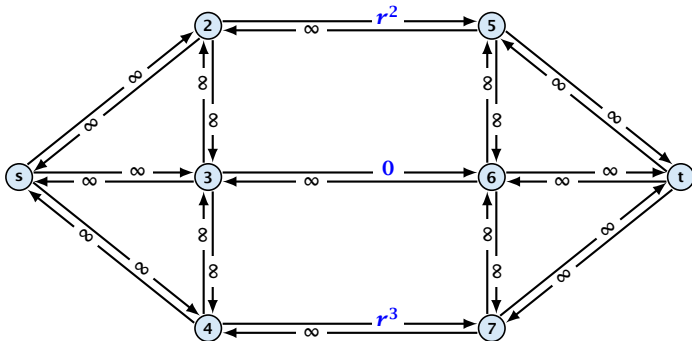
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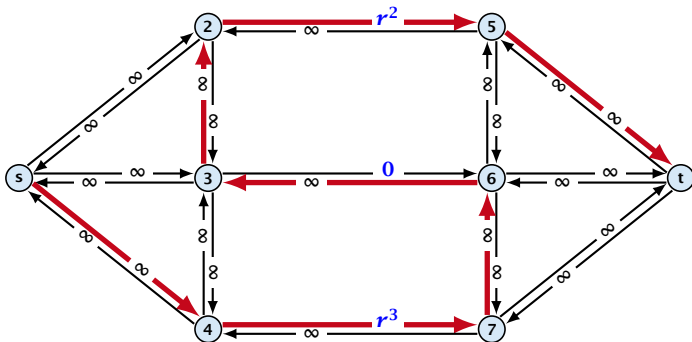
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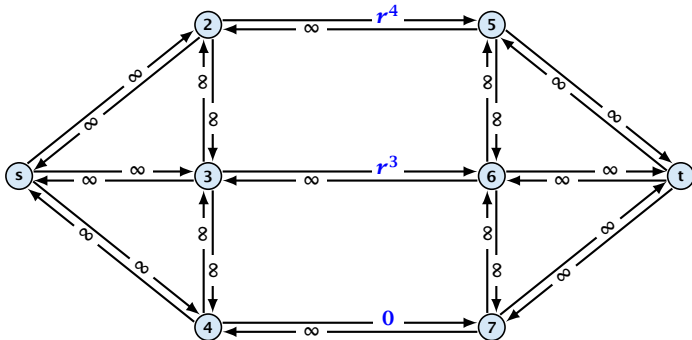
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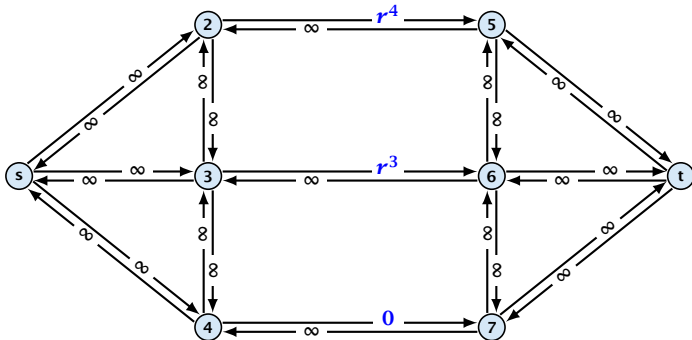
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Running time may be infinite!!!



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- ▶ Choose the shortest augmenting path.