

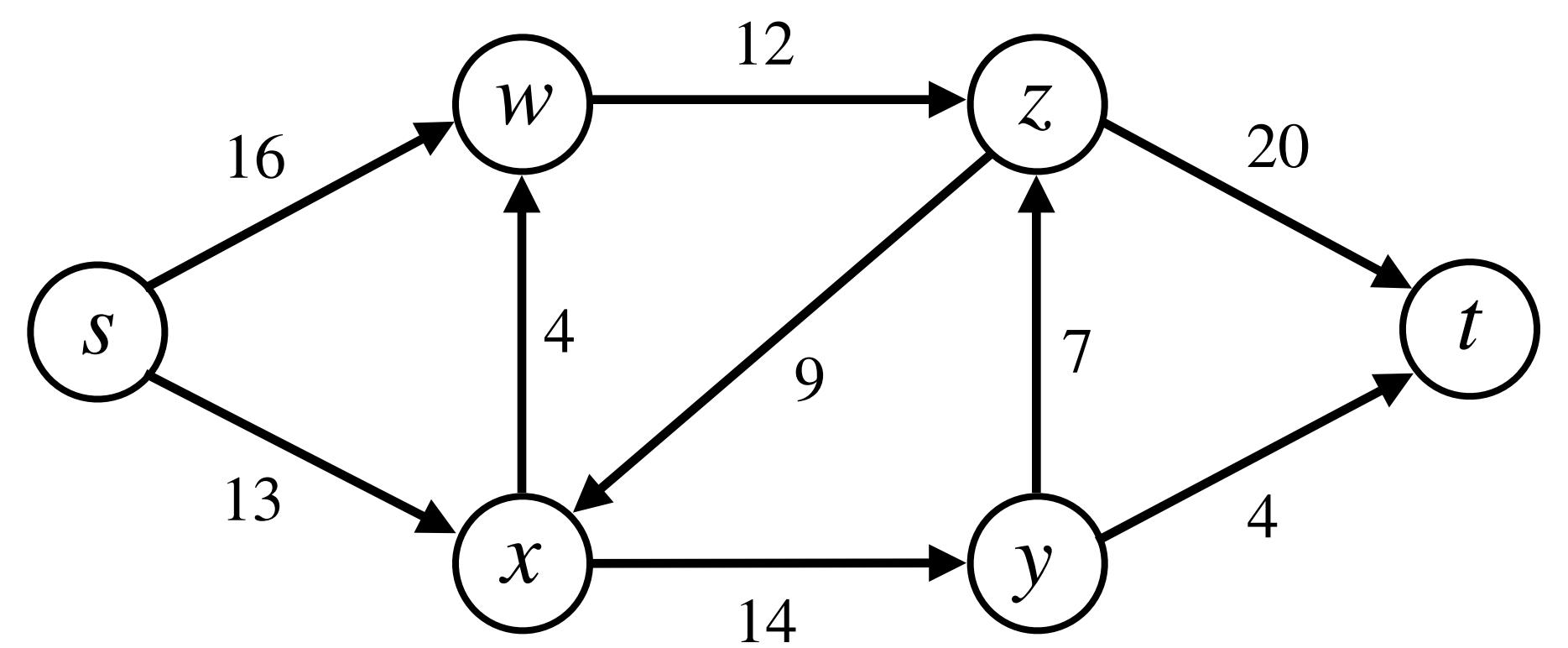
Lecture 17

Flow Networks, Ford-Fulkerson Method

Source: Introduction to Algorithms, CLRS and Kleinberg & Tardos

Flow Networks

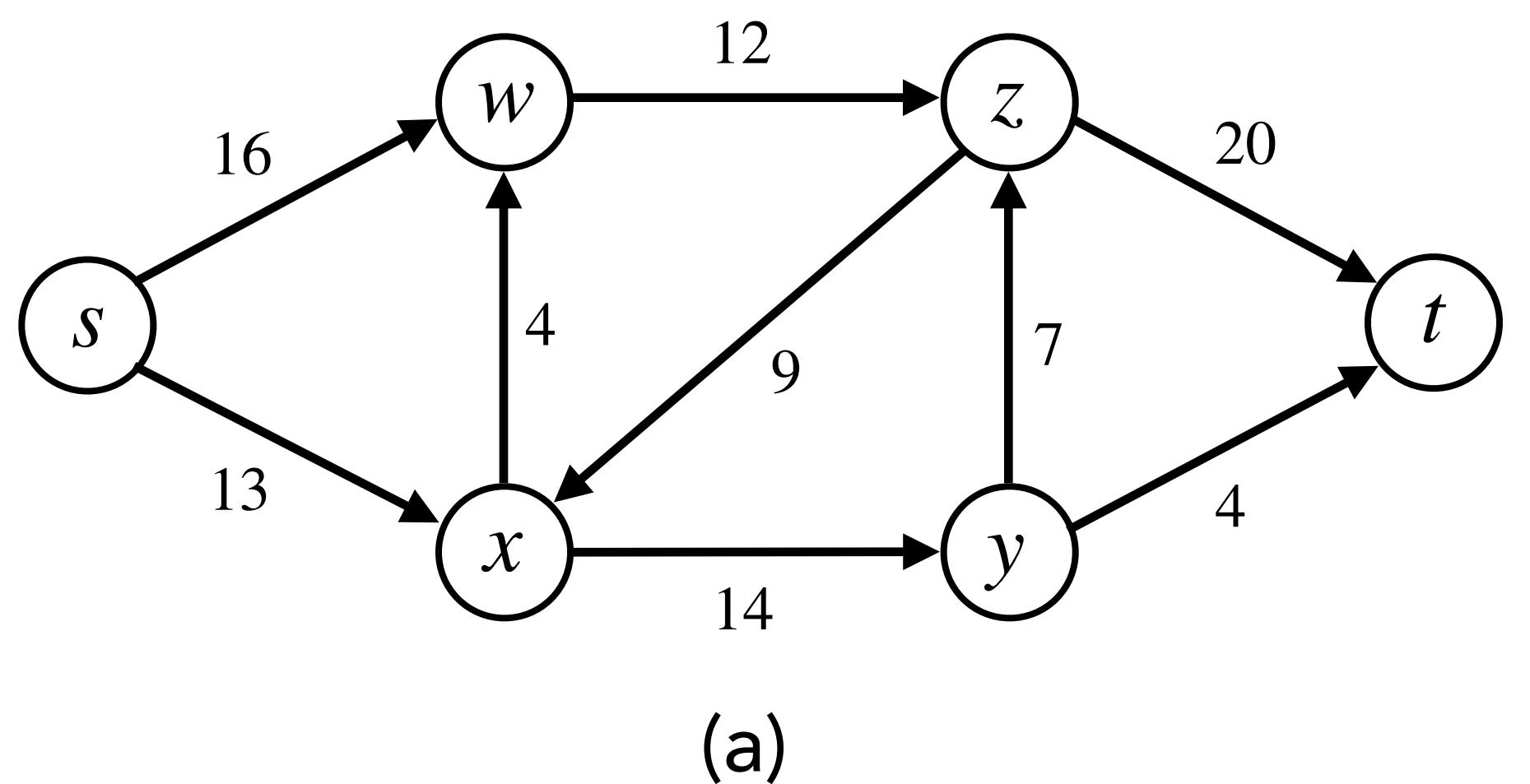
Flow Networks



(a)

Flow Networks

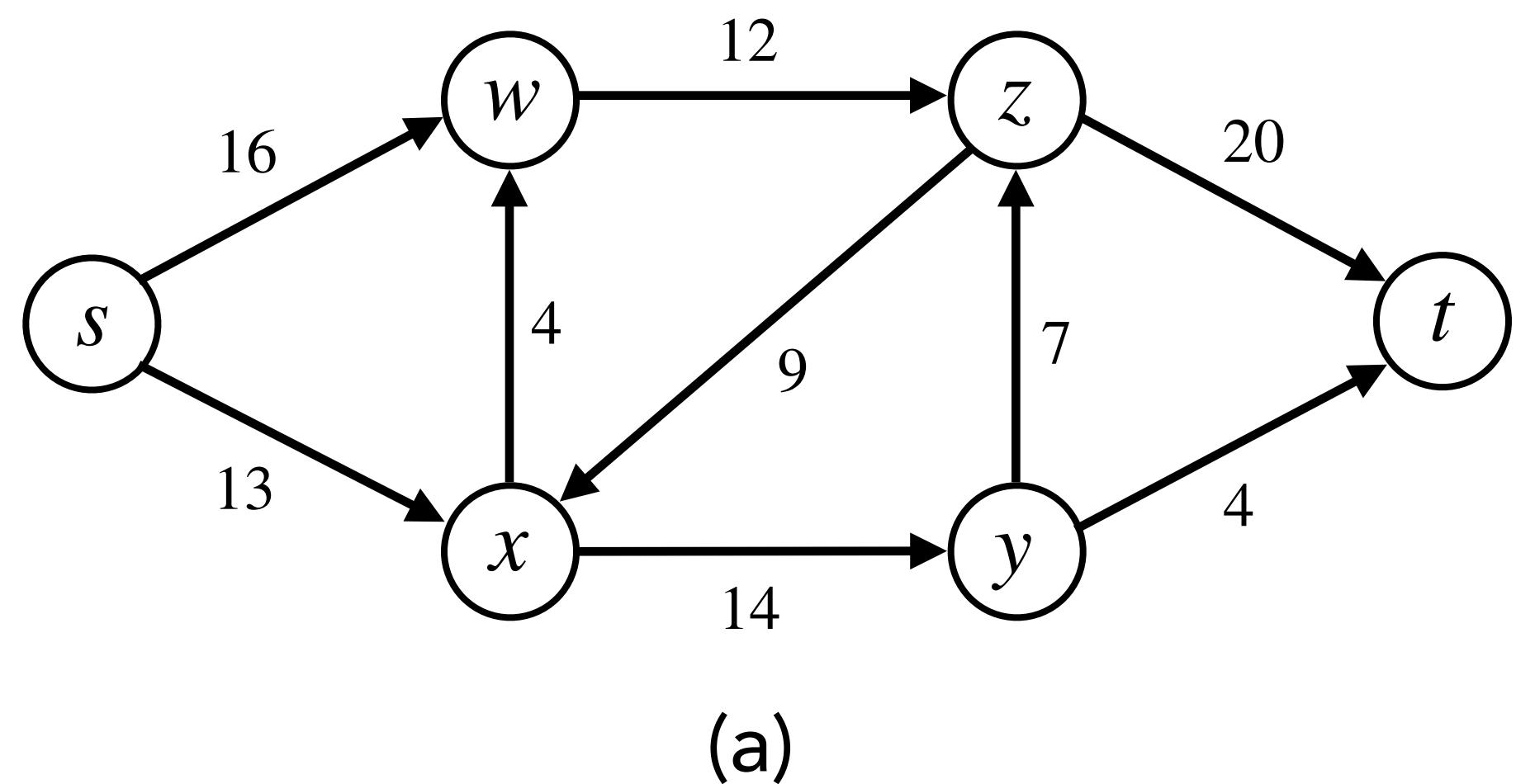
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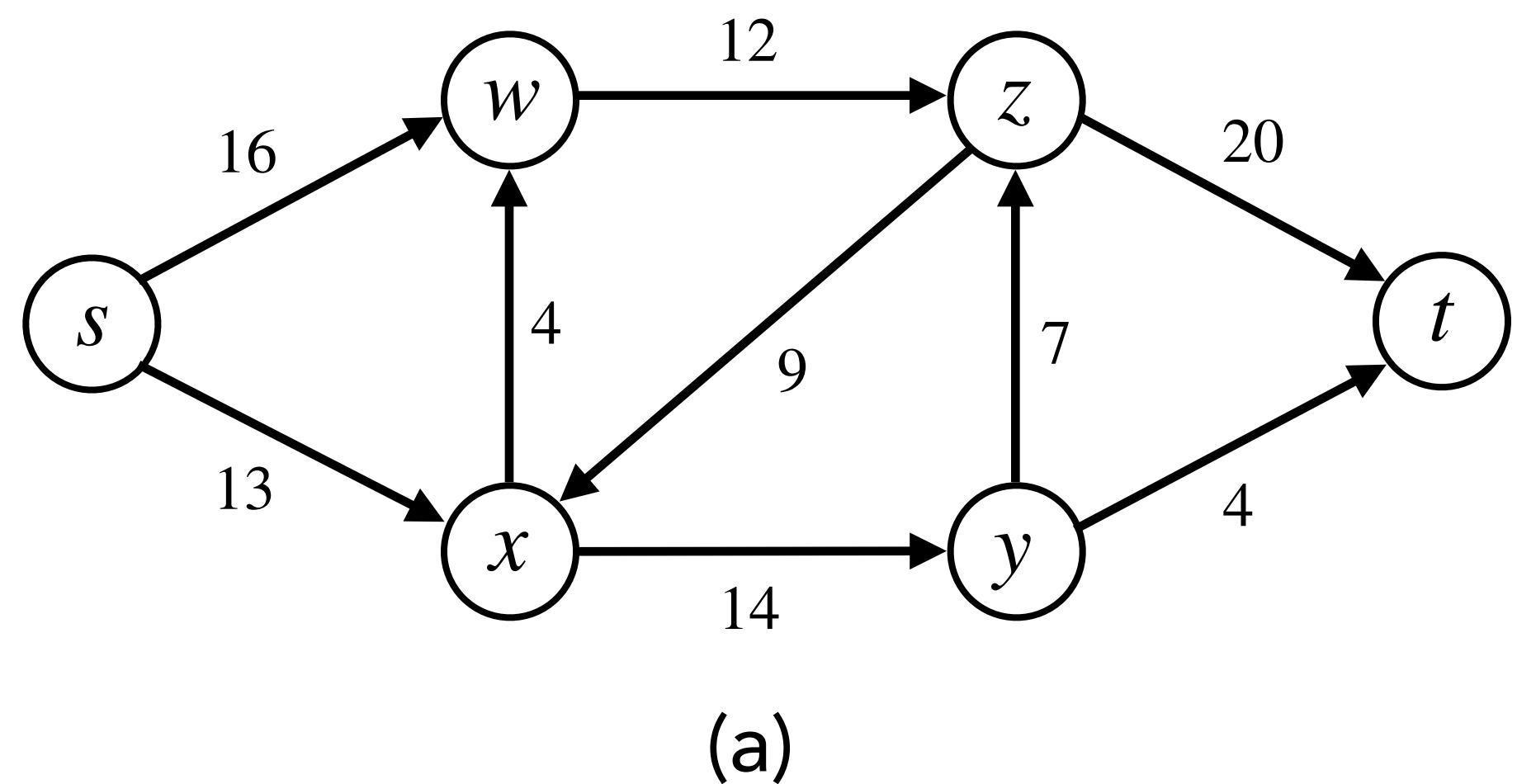
- Vertices represent **cities**. s & t are the source & sink cities.



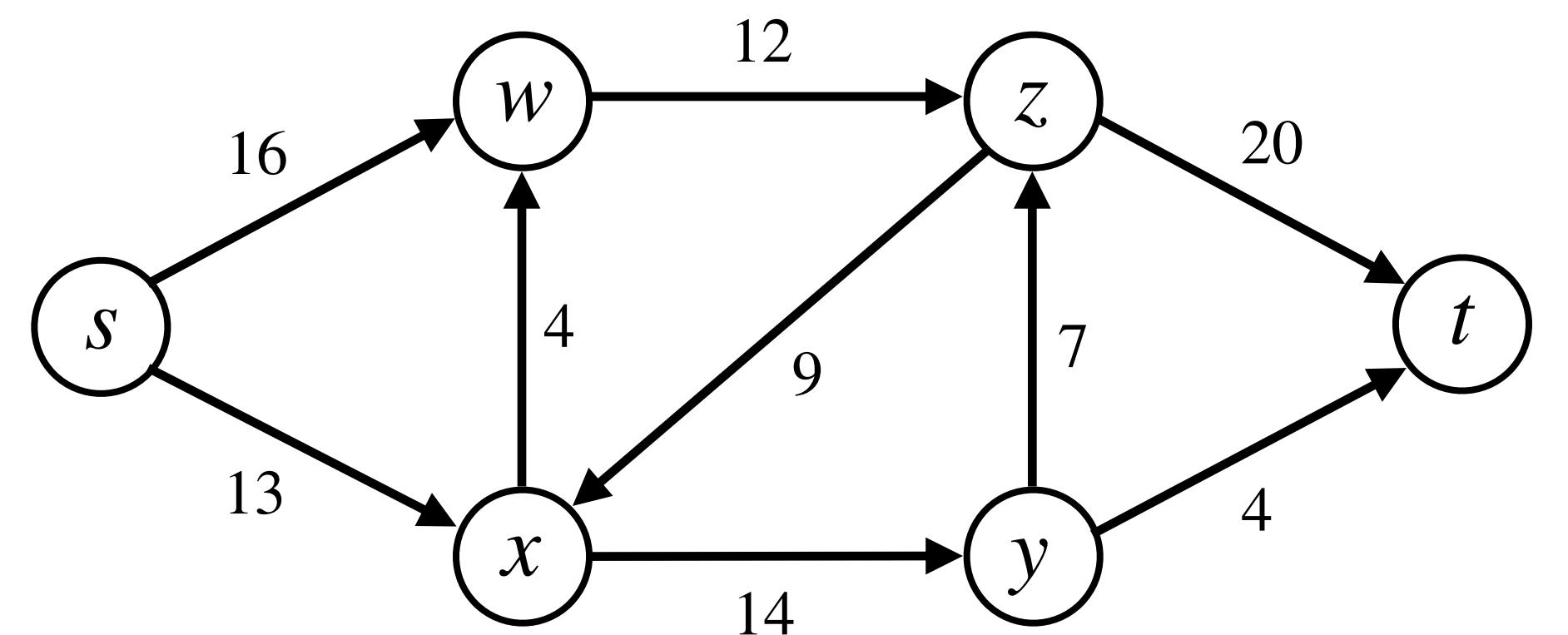
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- Vertices represent **cities**. s & t are the **source** & **sink** cities.
- The number on any (u, v) edge is the **maximum number of packets** that can go from u to v per day.



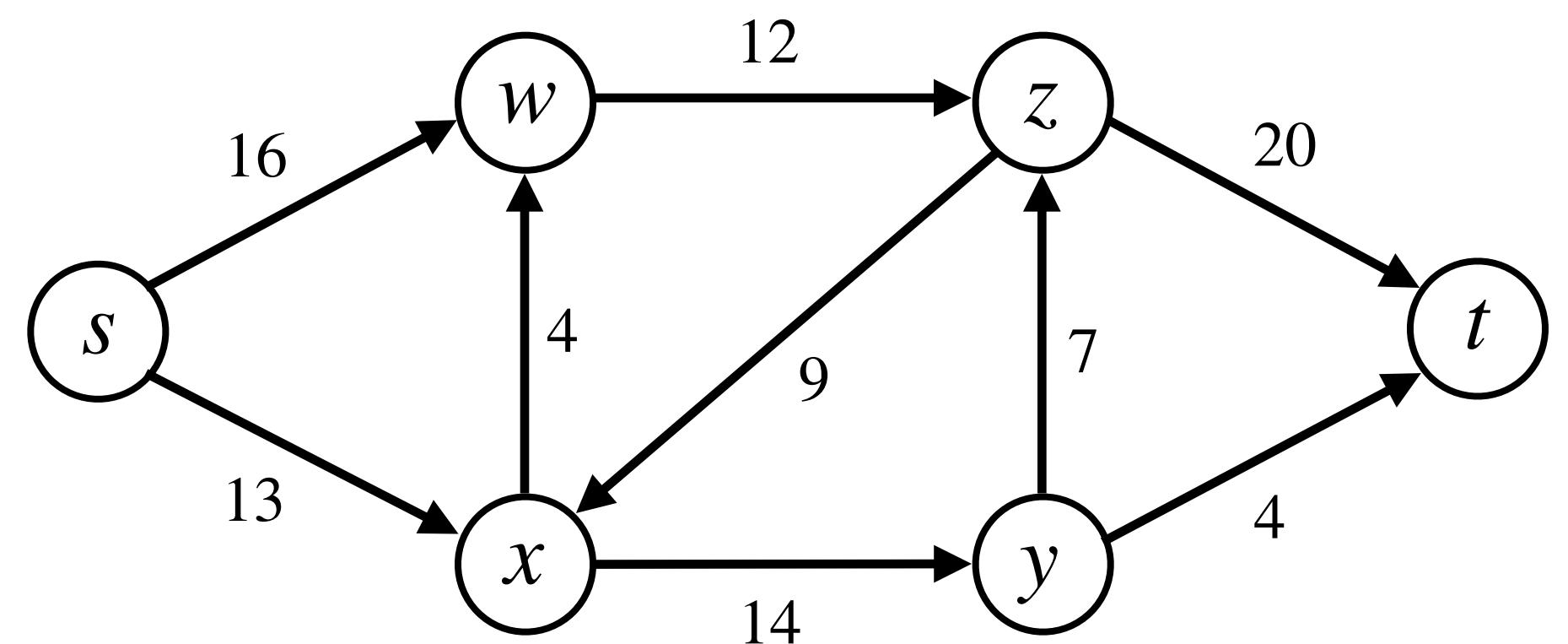
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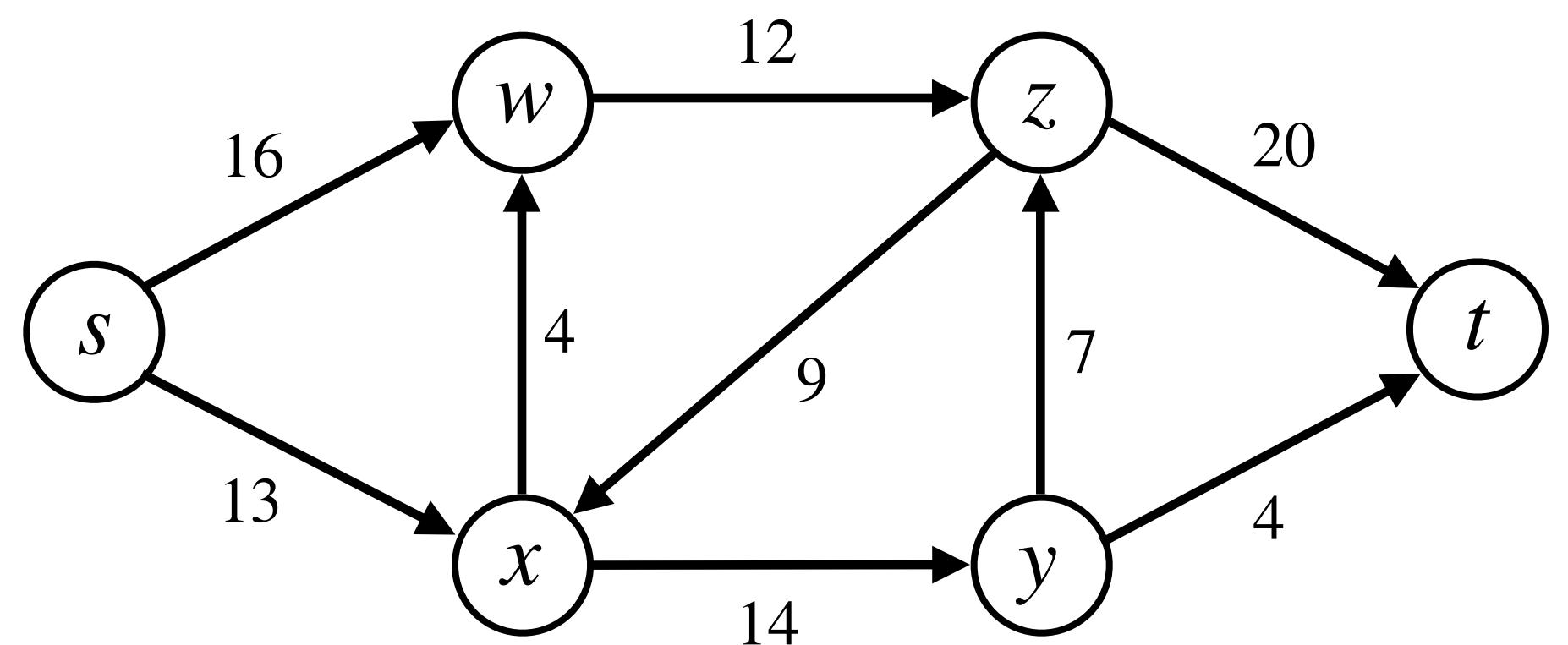
Goal: Find the maximum number of packets that can be shipped from s if the packets received and



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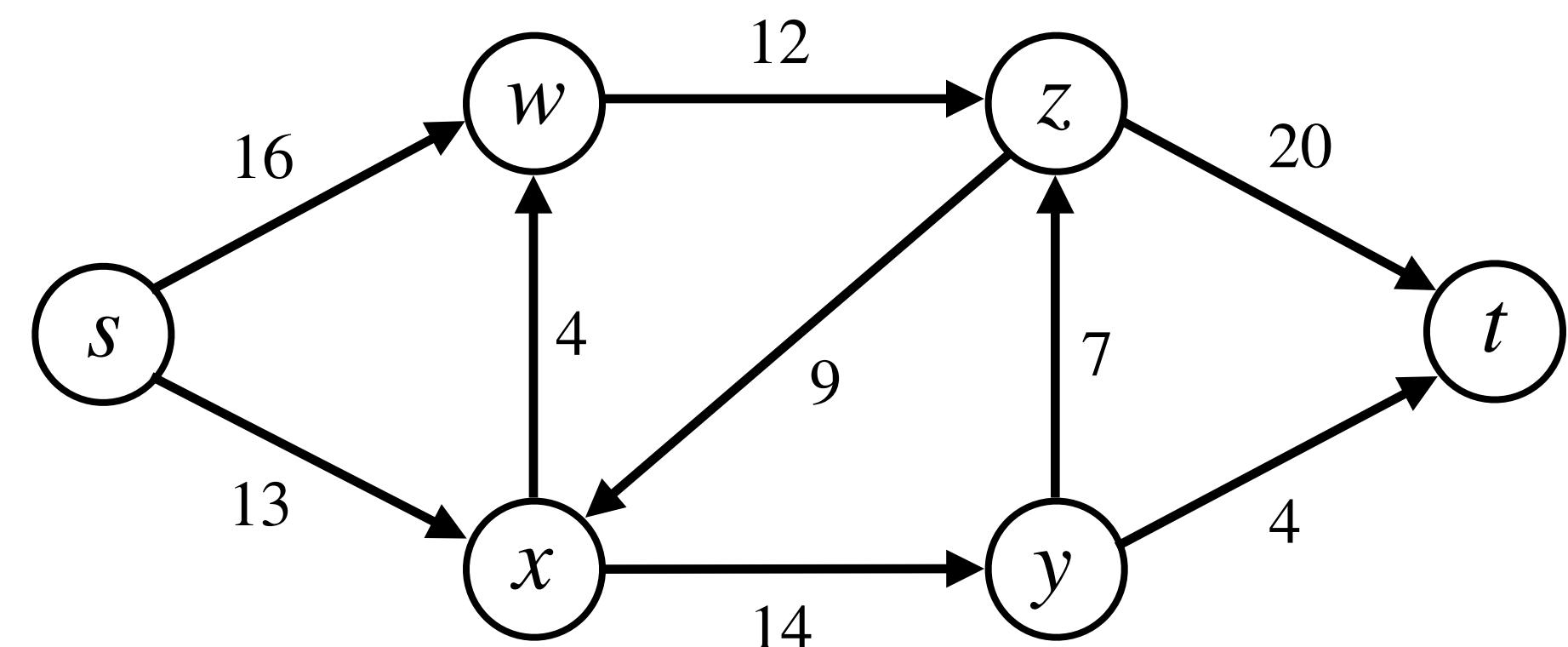
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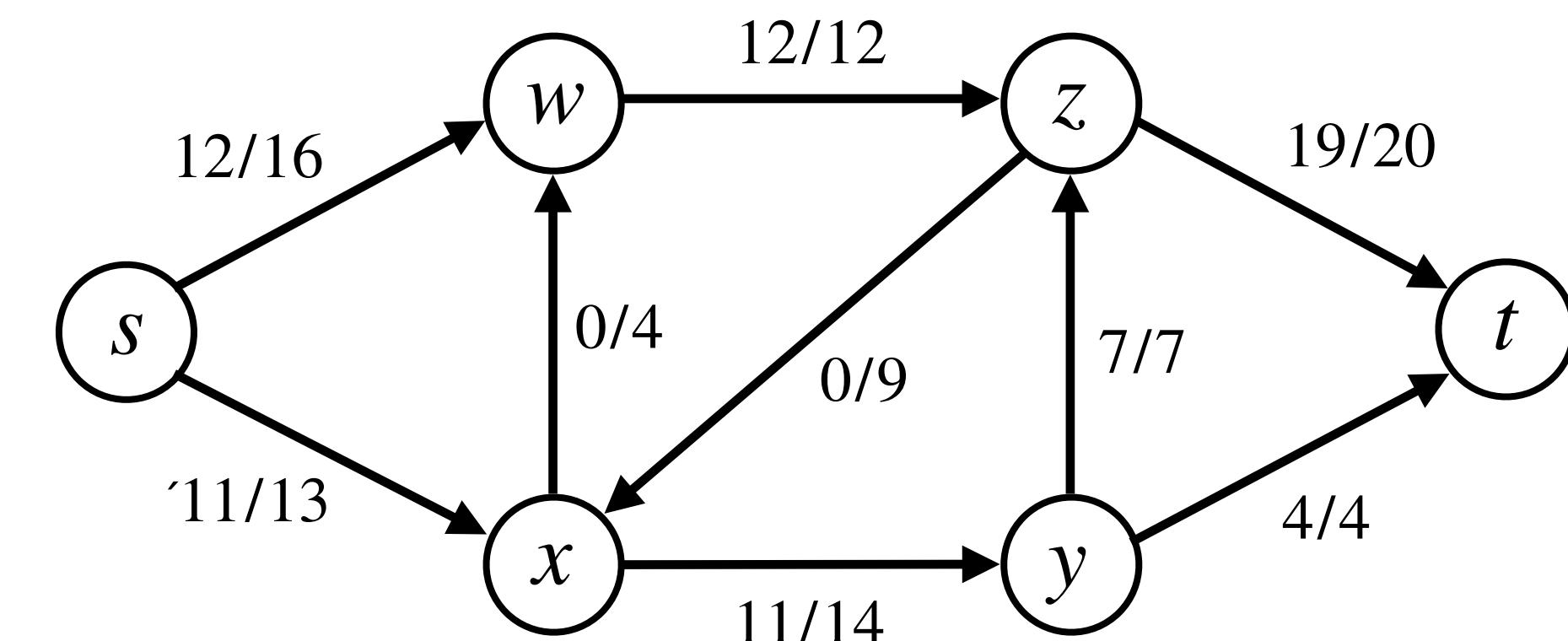
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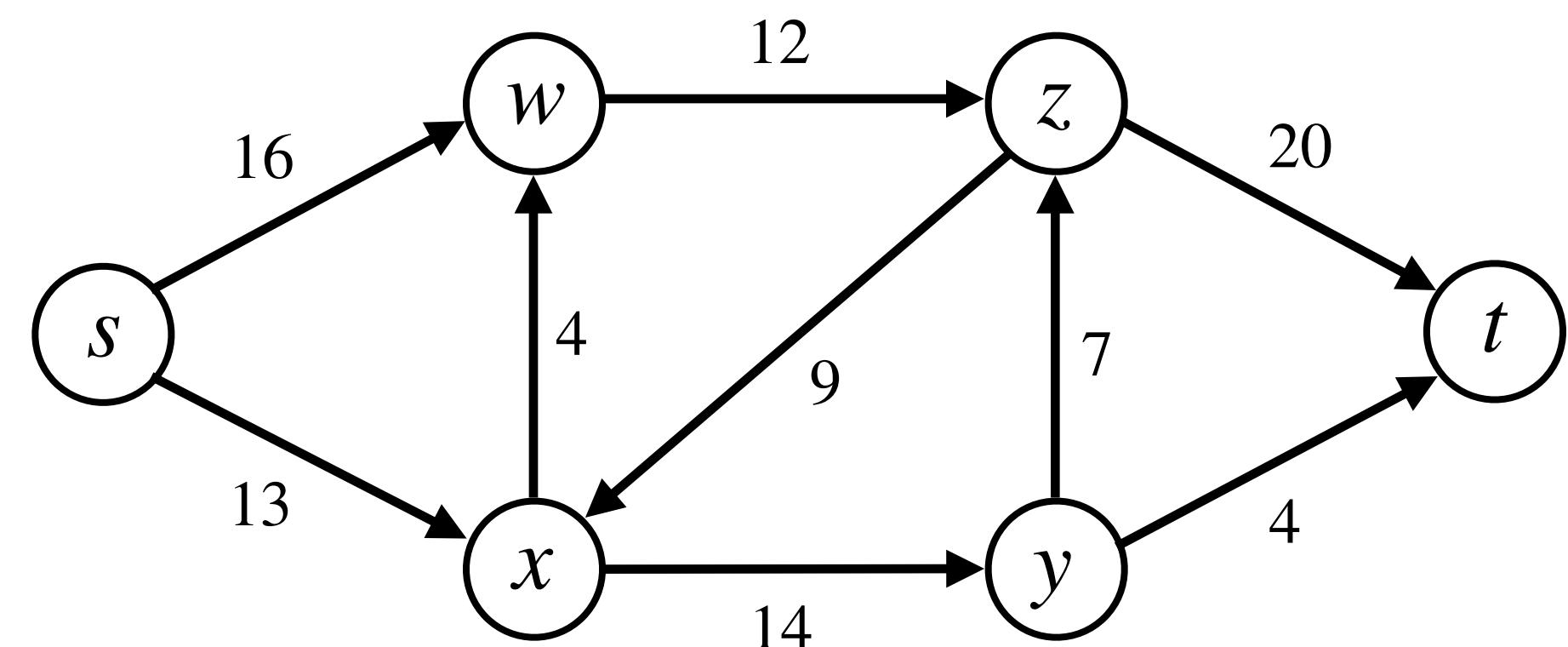
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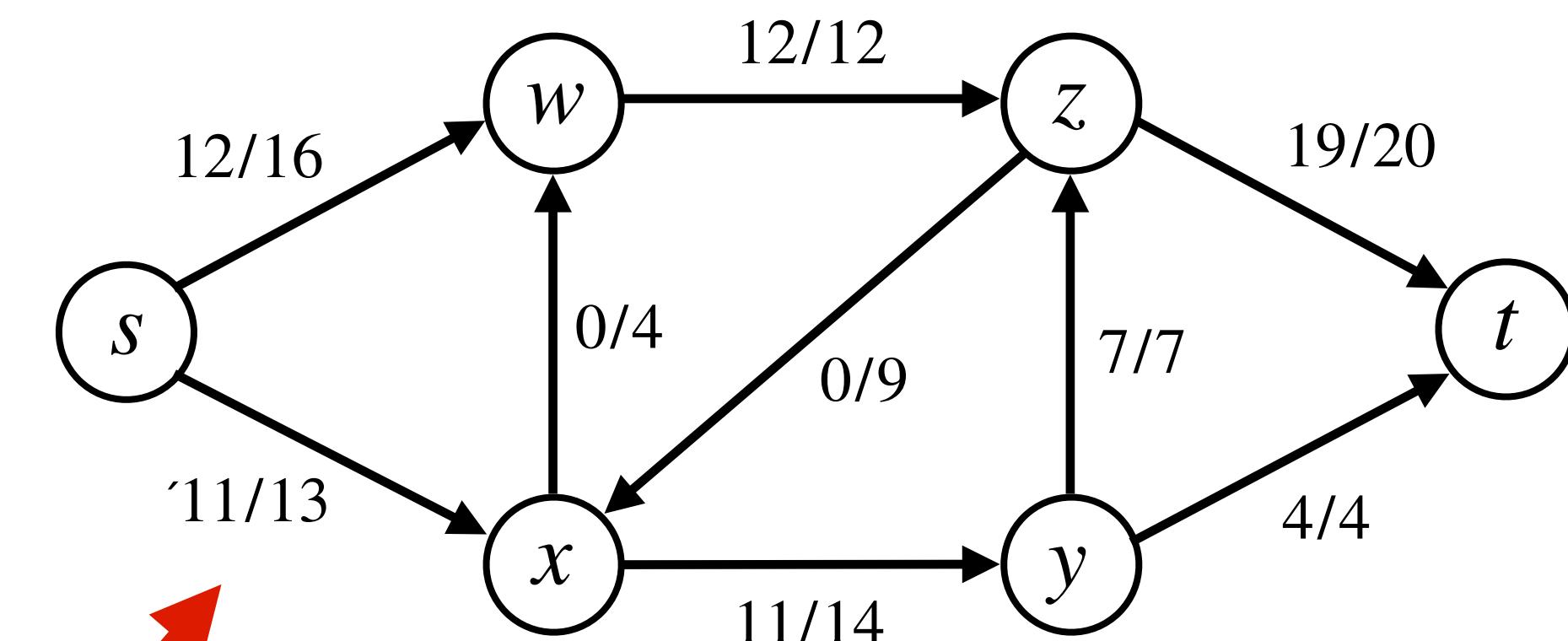
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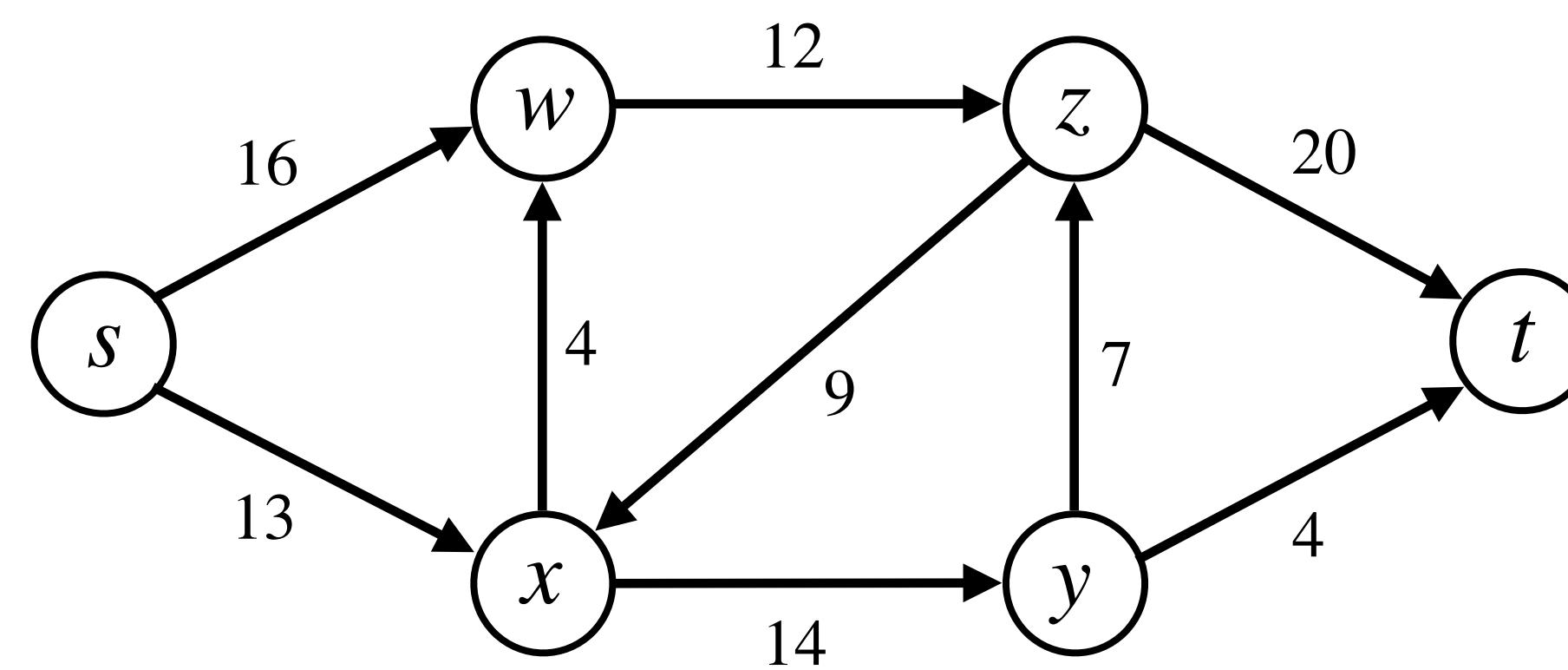
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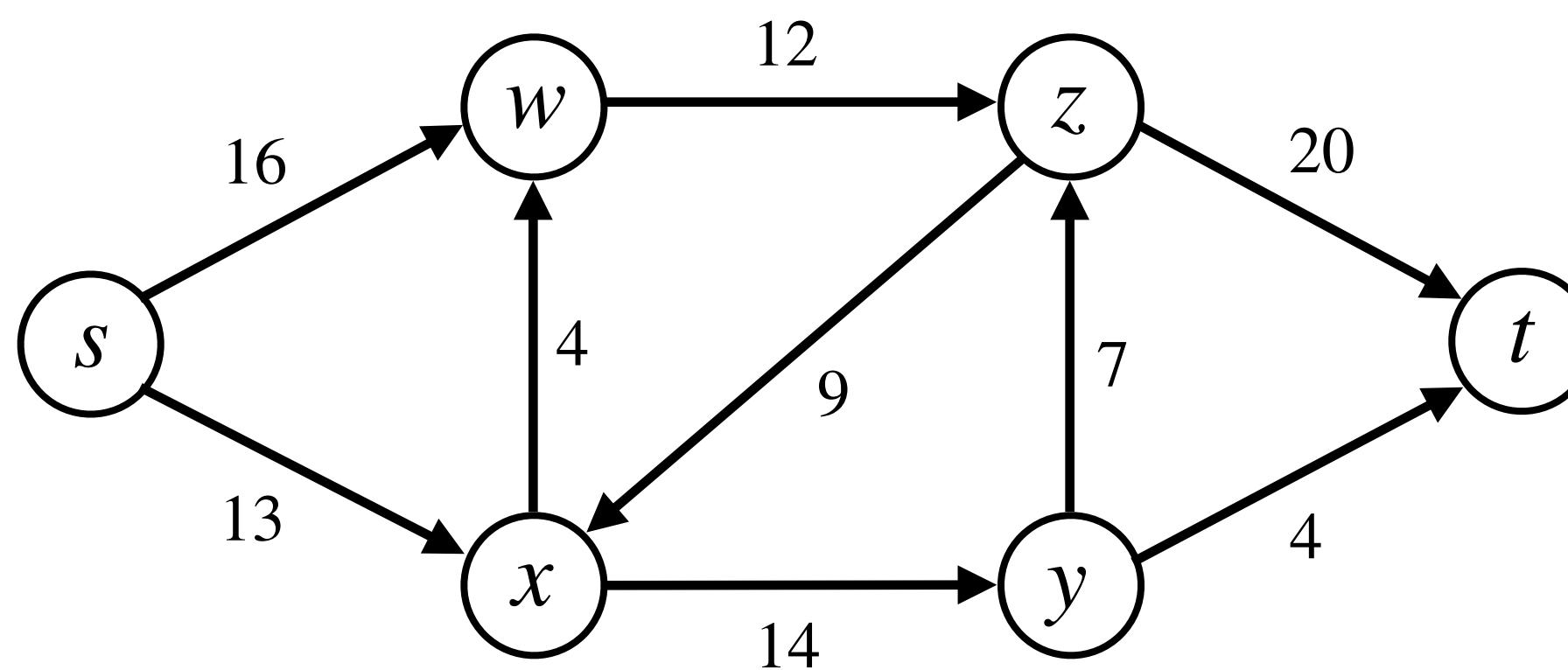
max packets = 23

Flow Networks



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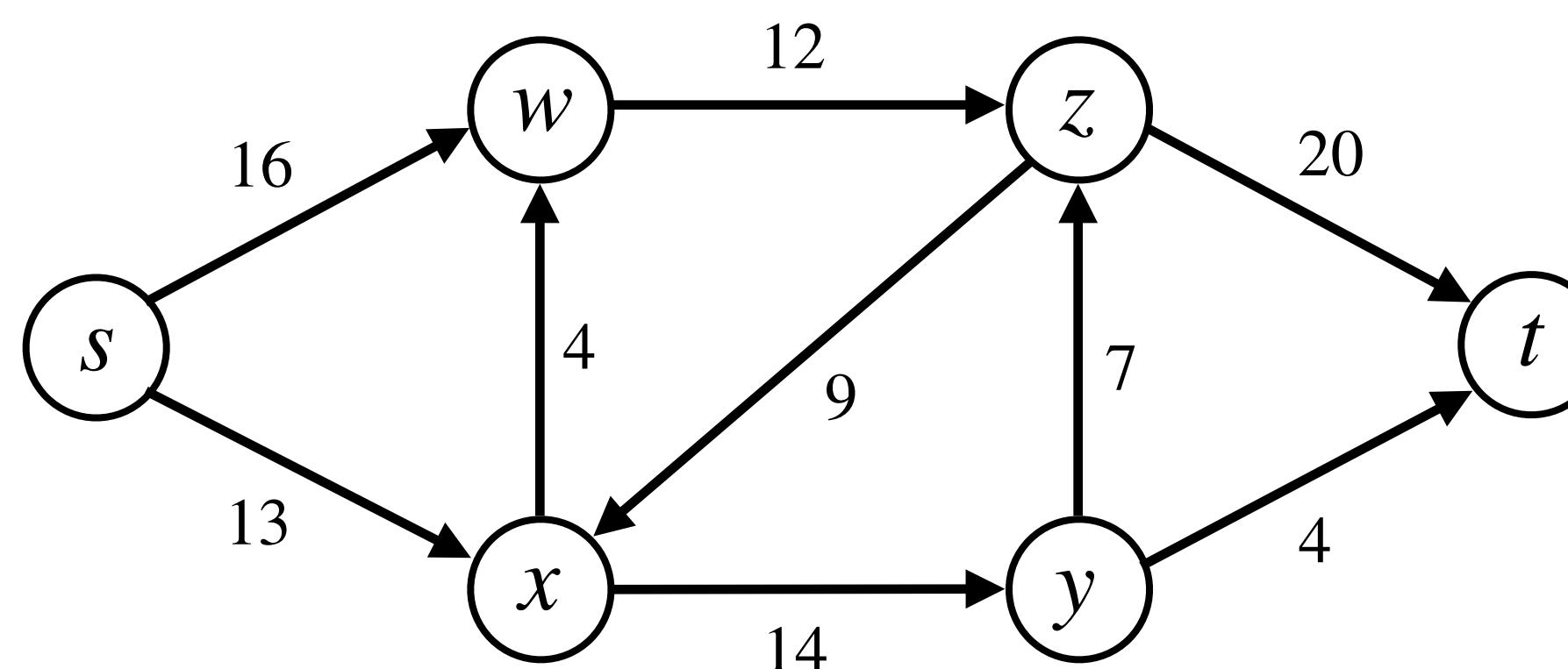
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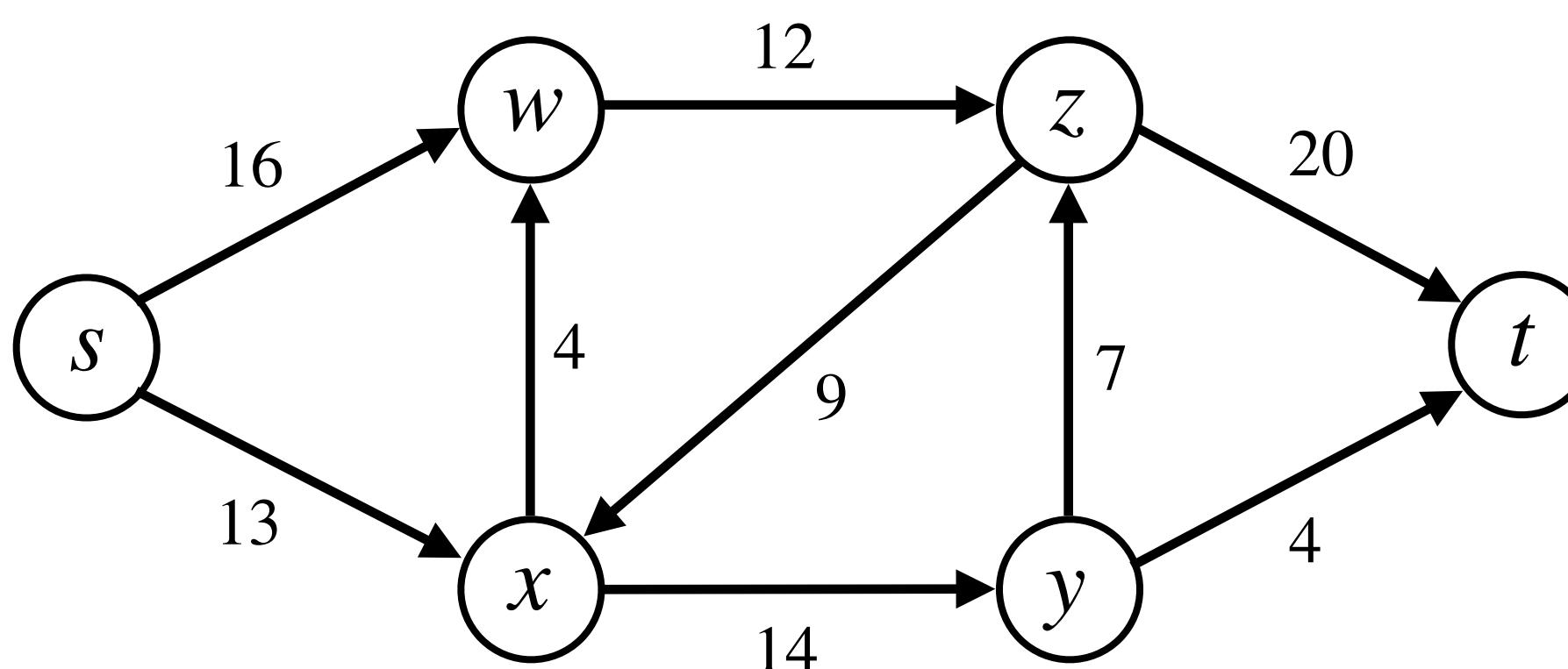
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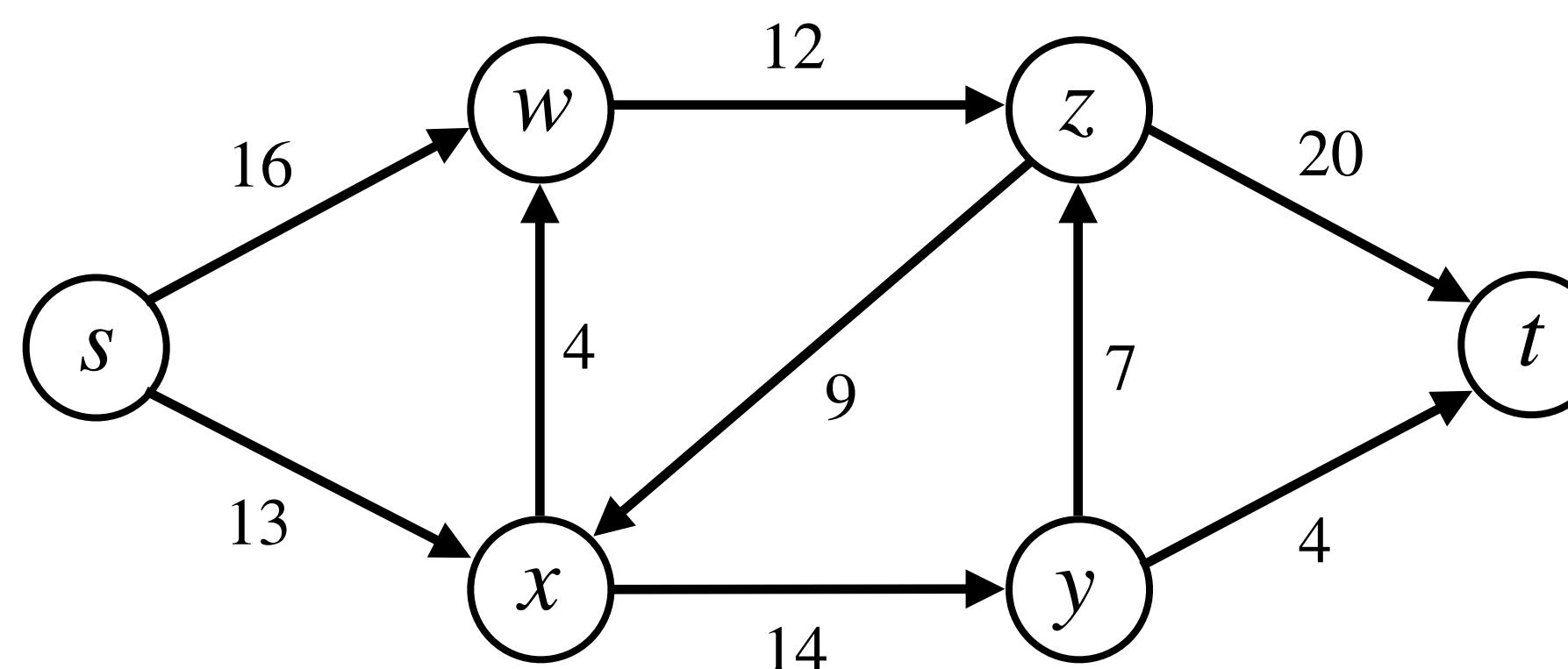
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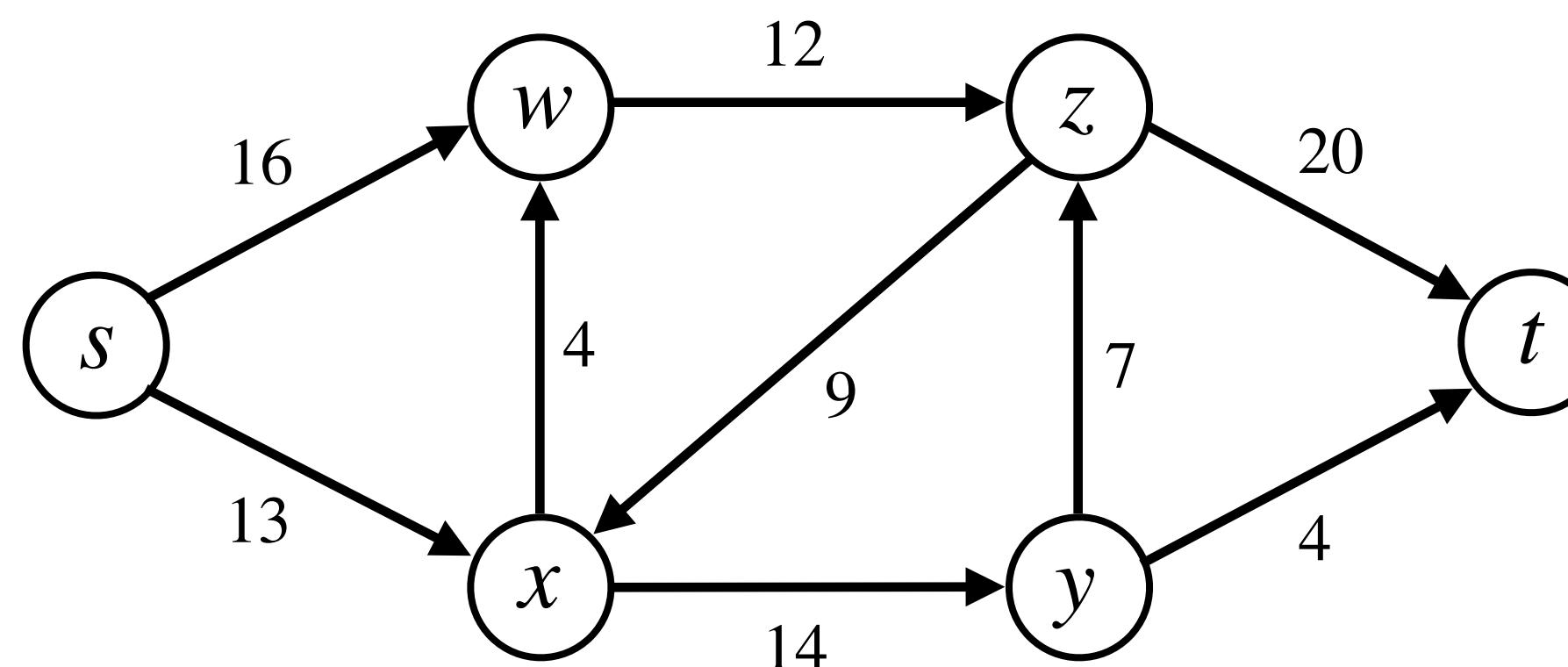
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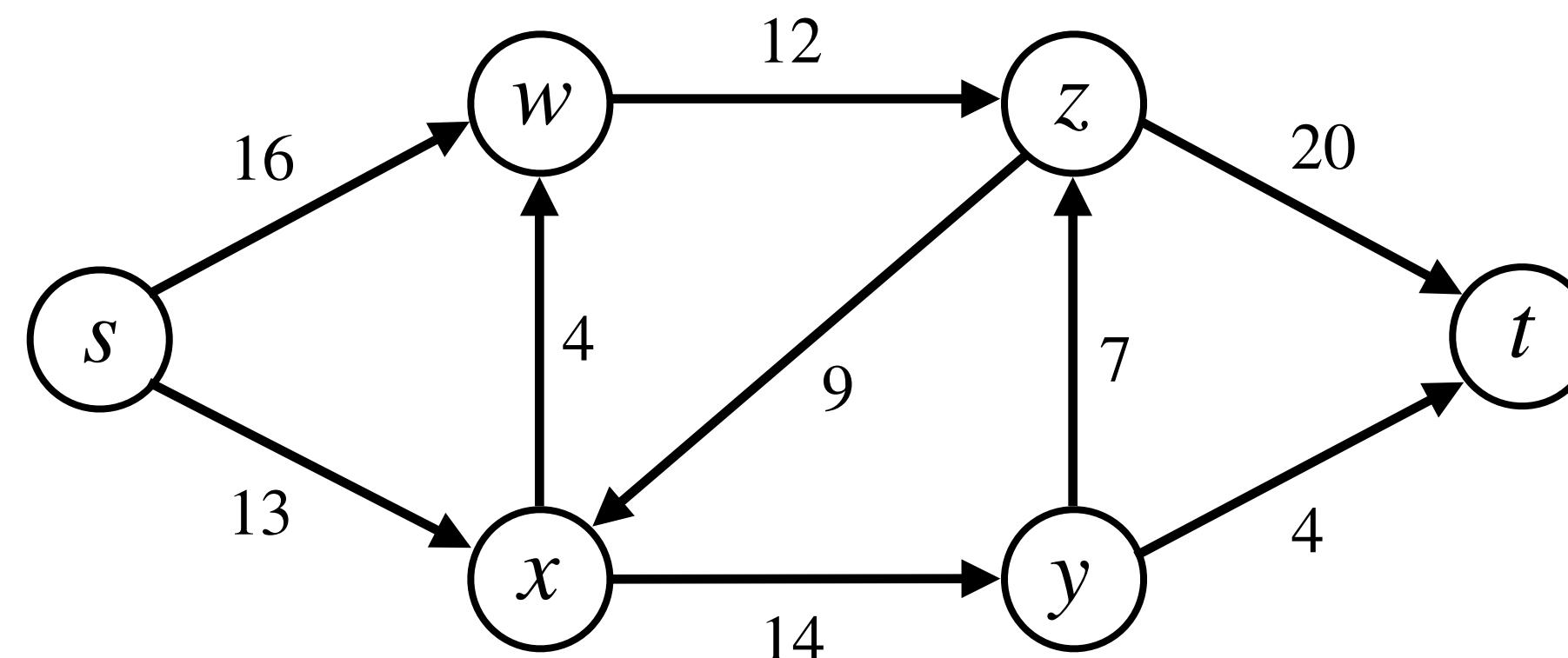
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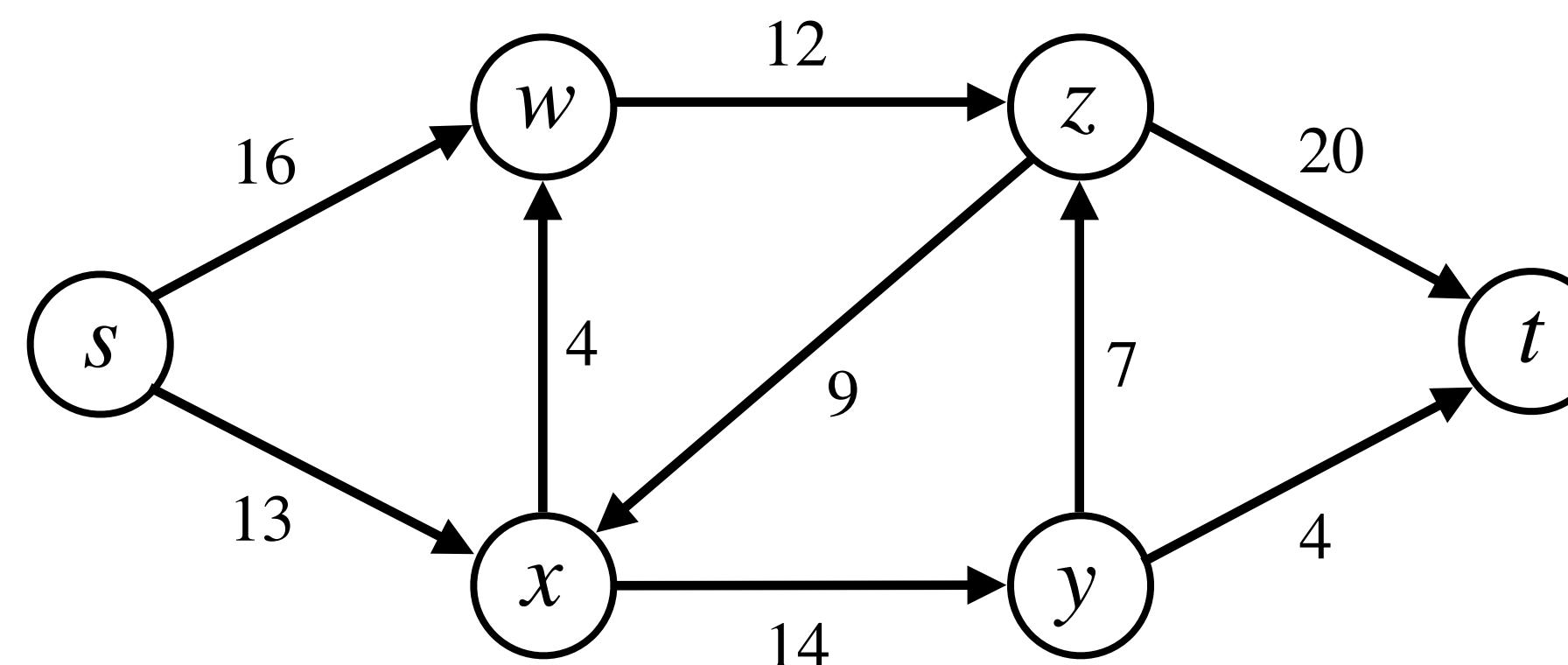
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- Two distinguished vertices: **source s** (no incoming edges) and **sink t** (no outgoing edges).
- For every $v \in V$, some $s \rightsquigarrow v \rightsquigarrow t$ path exists. Hence, $|E| \geq |V| - 1$.



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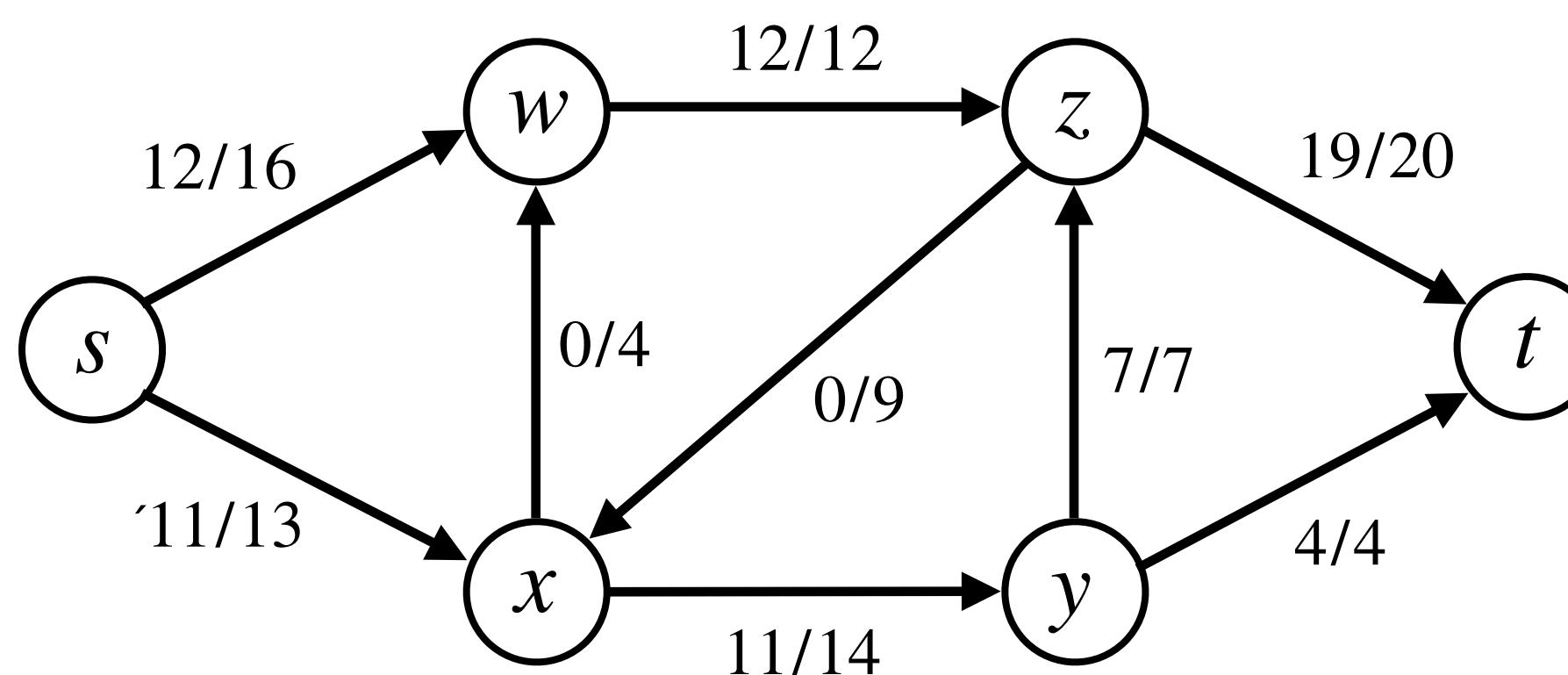
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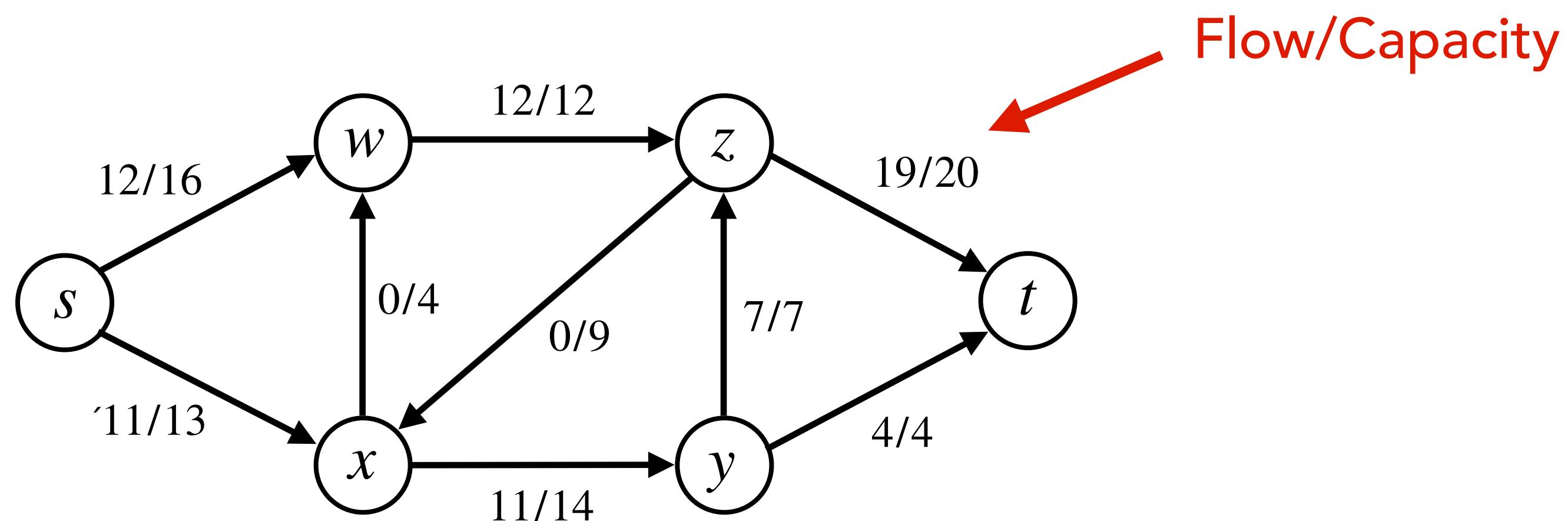
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Output: Flow of maximum value.

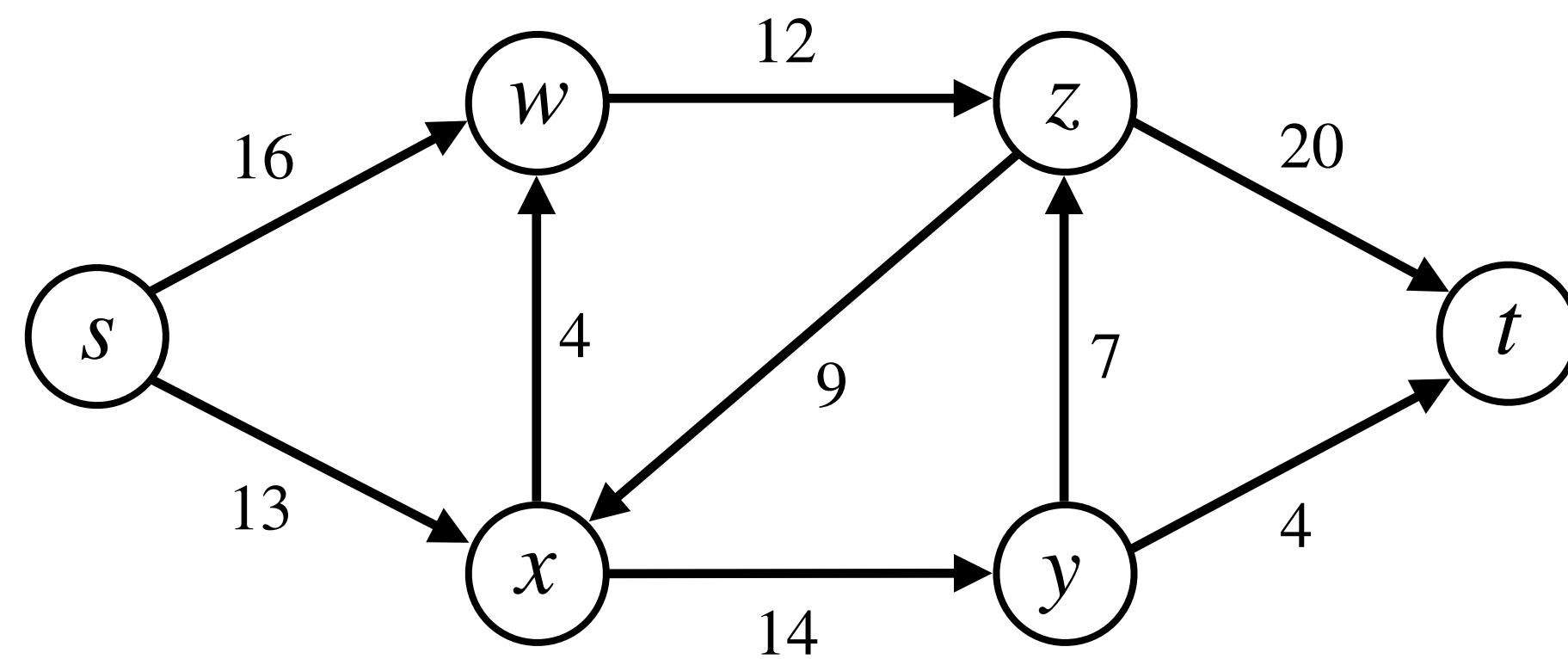
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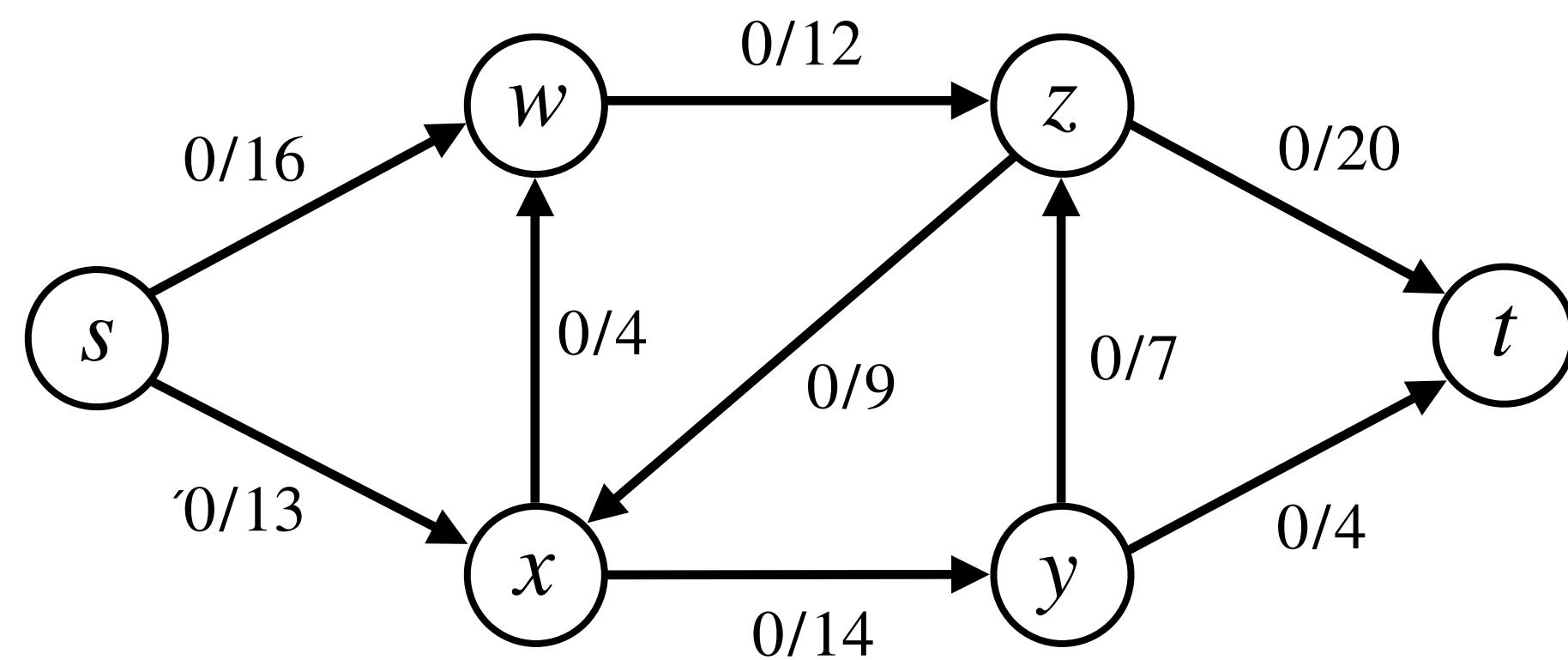


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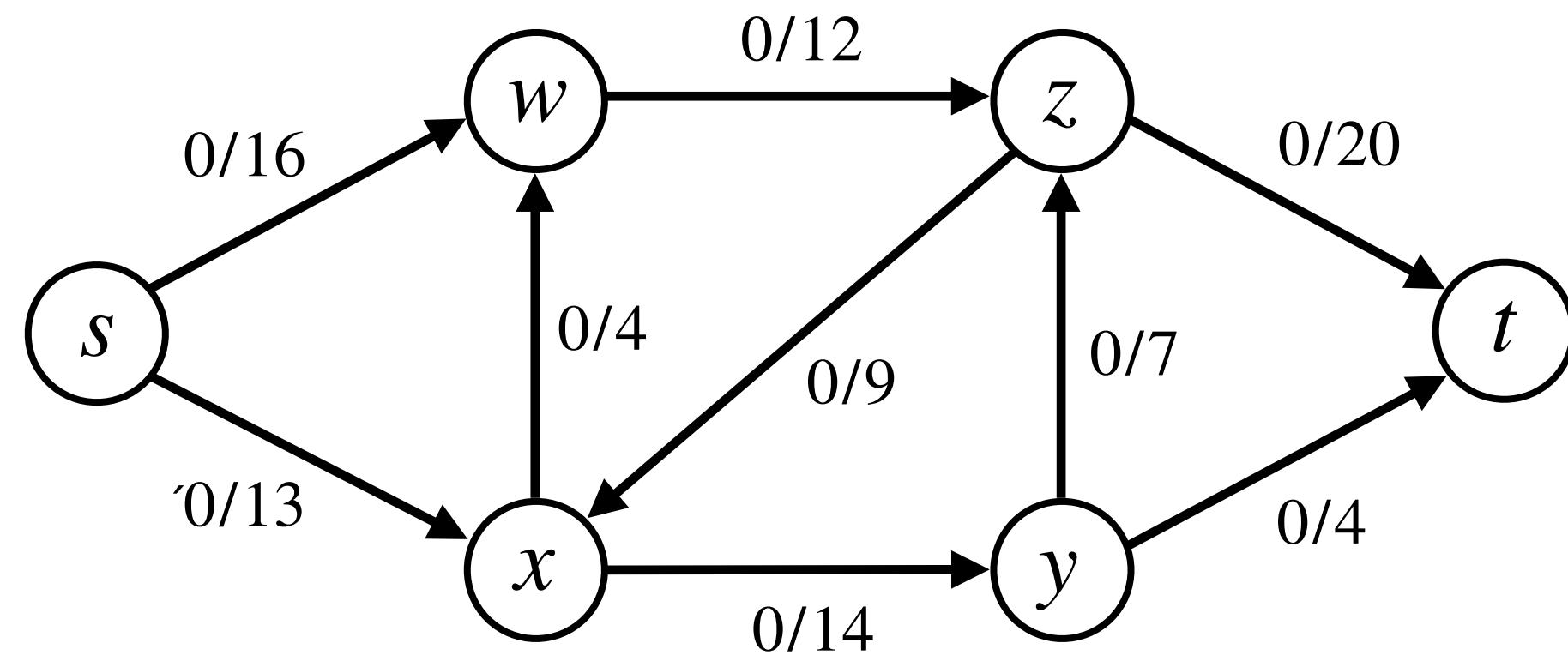


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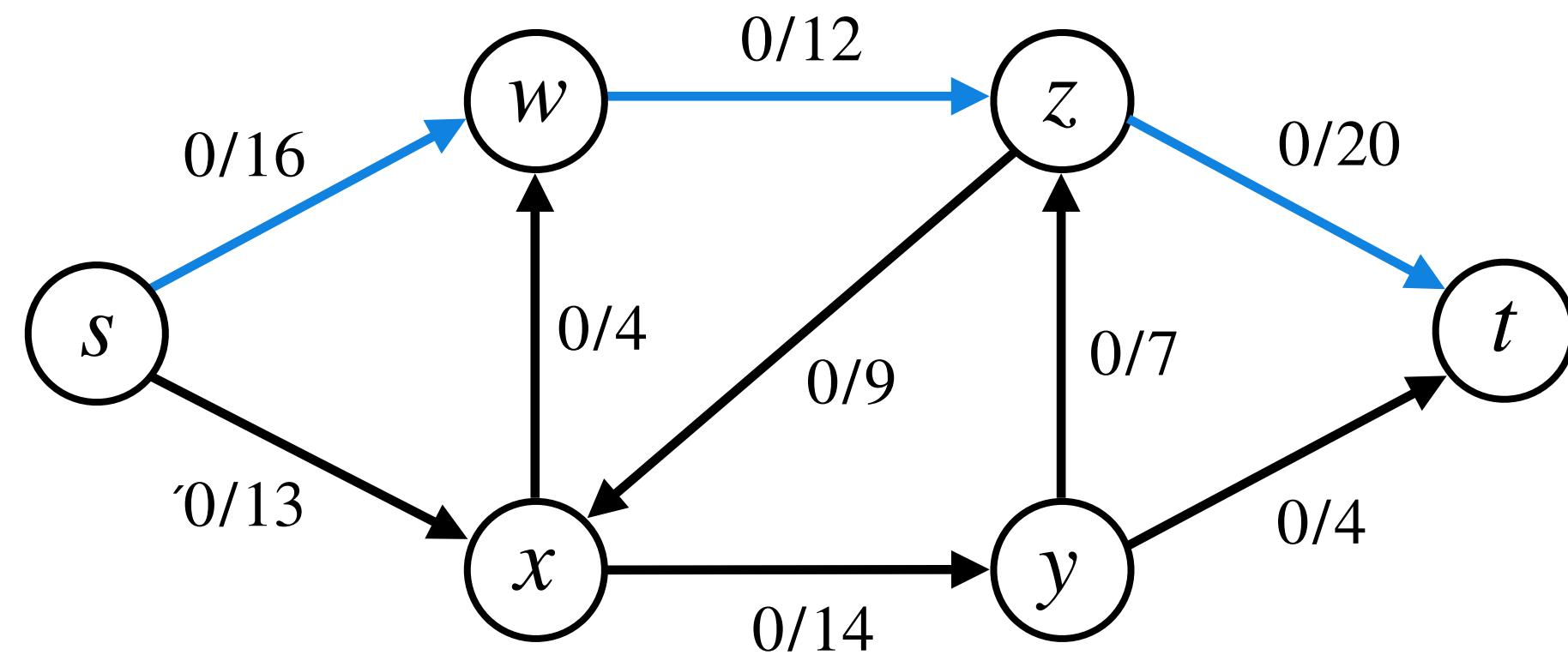


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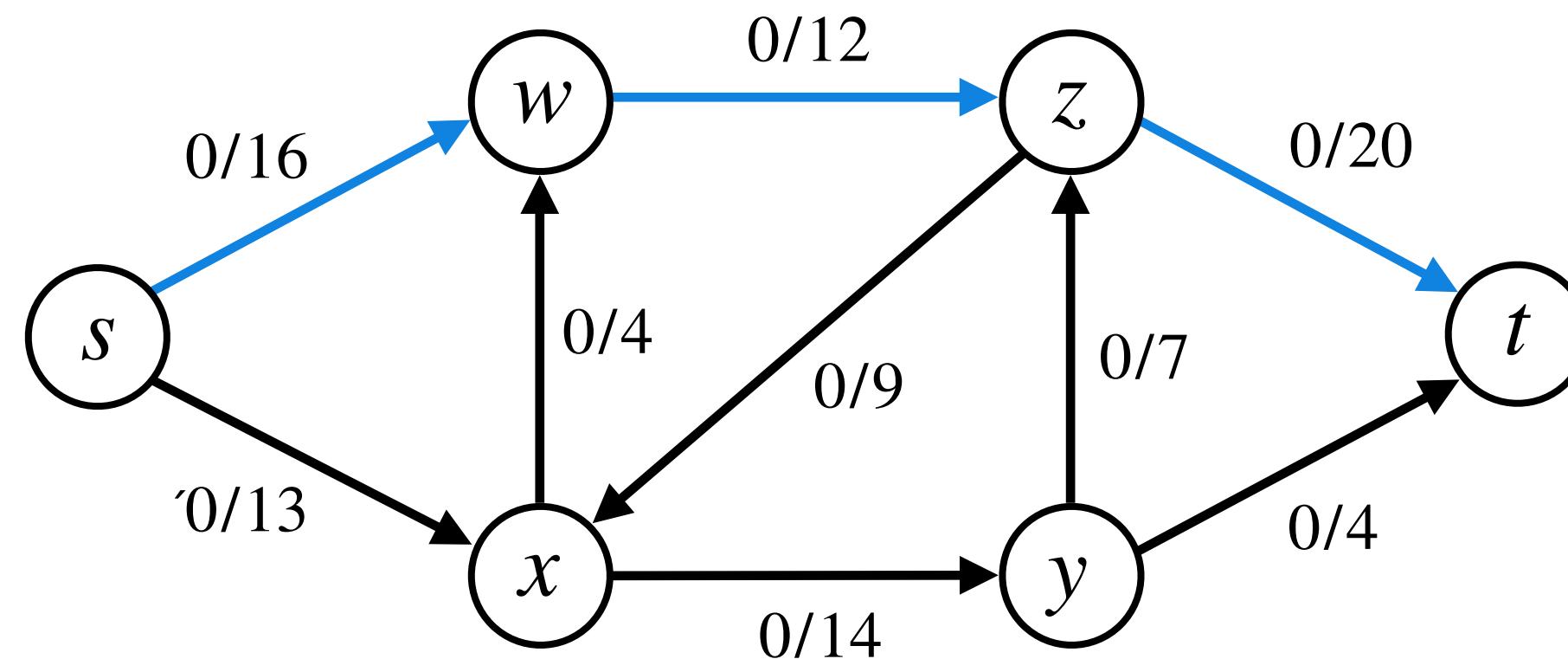


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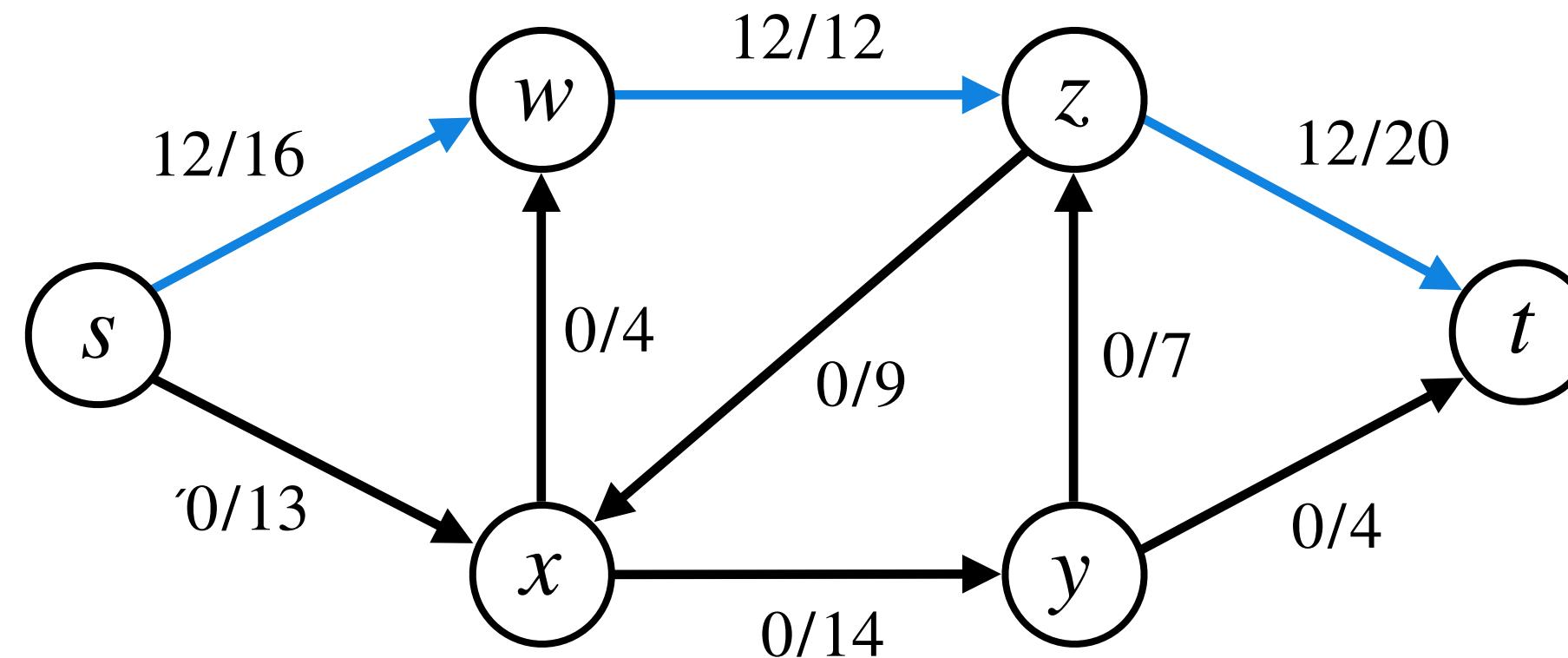


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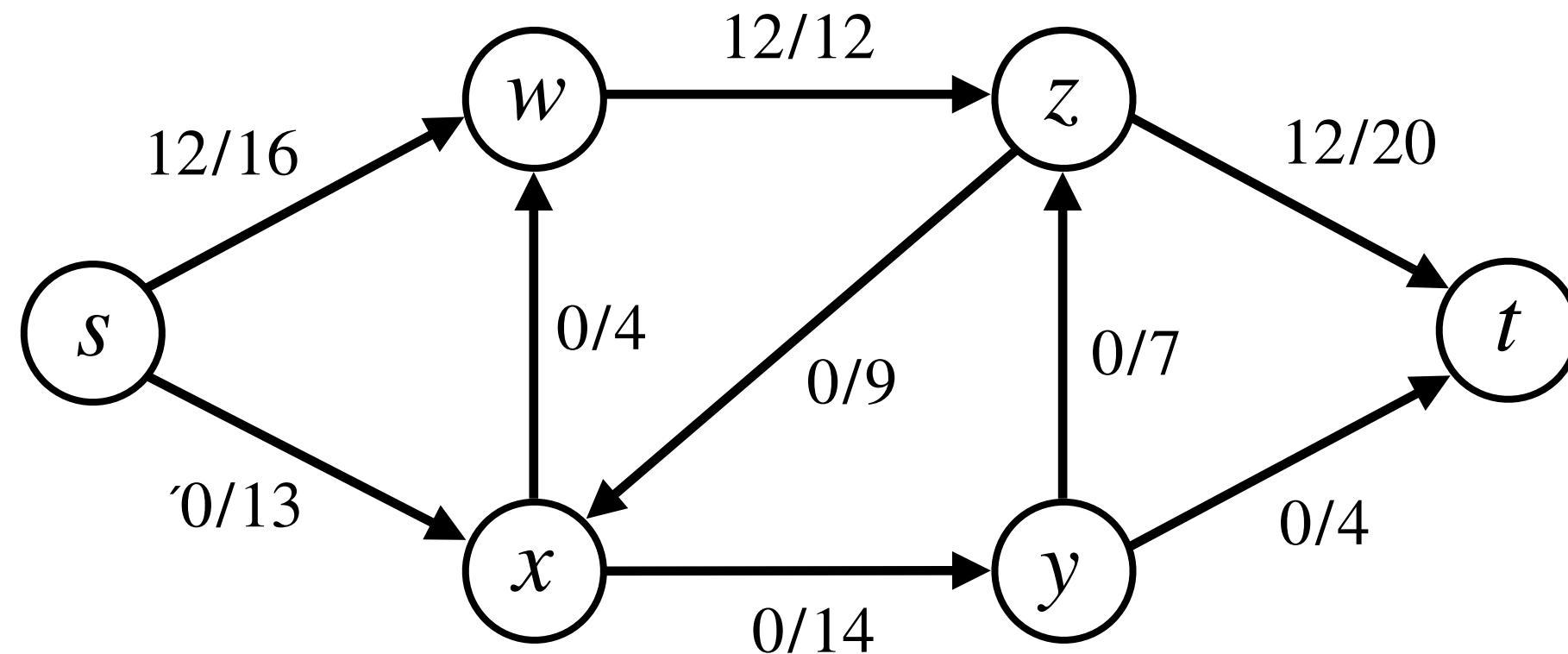


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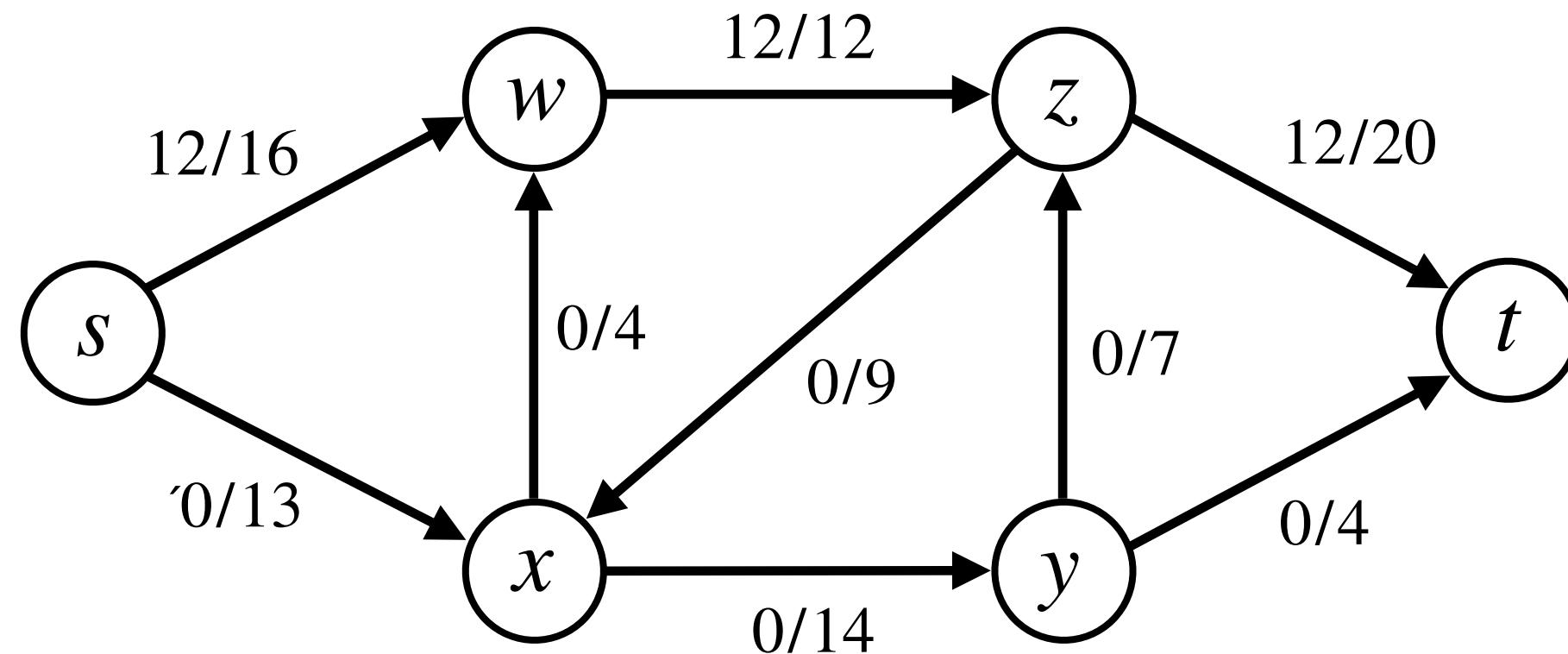


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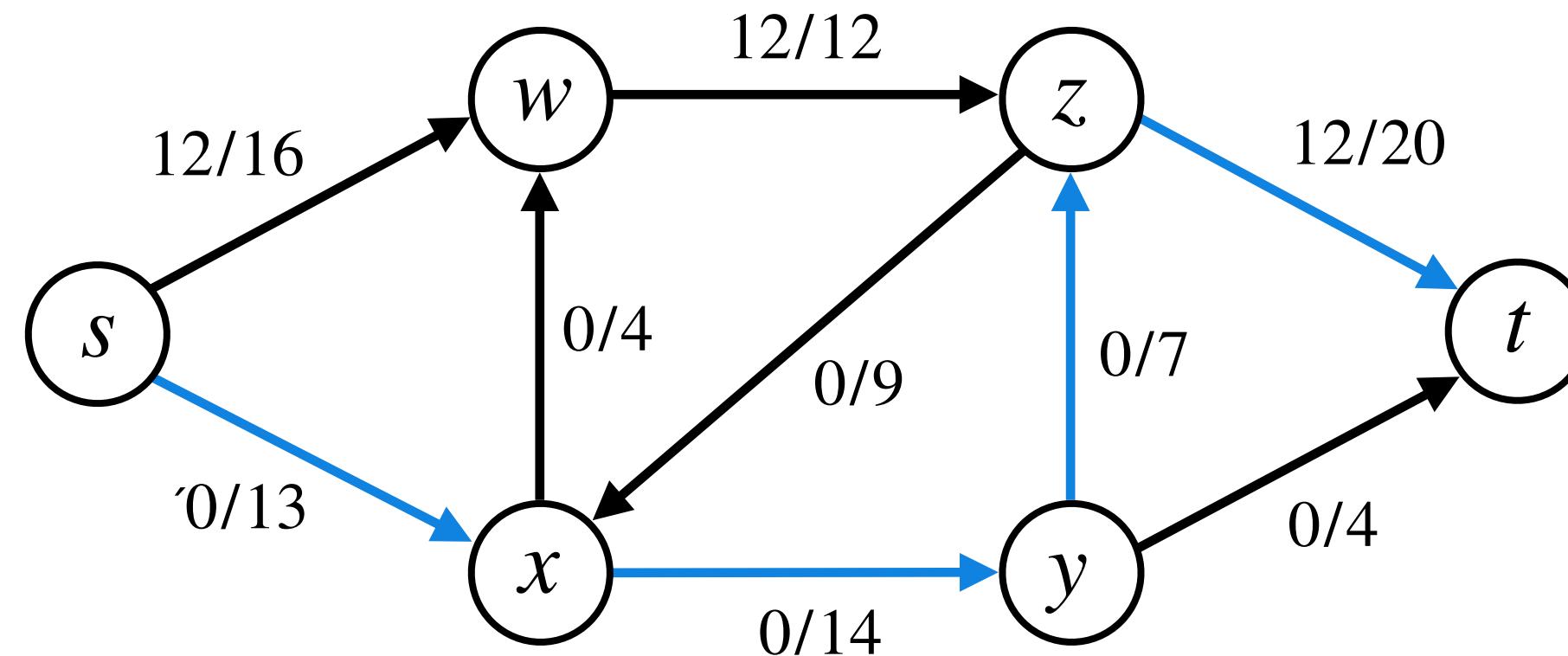


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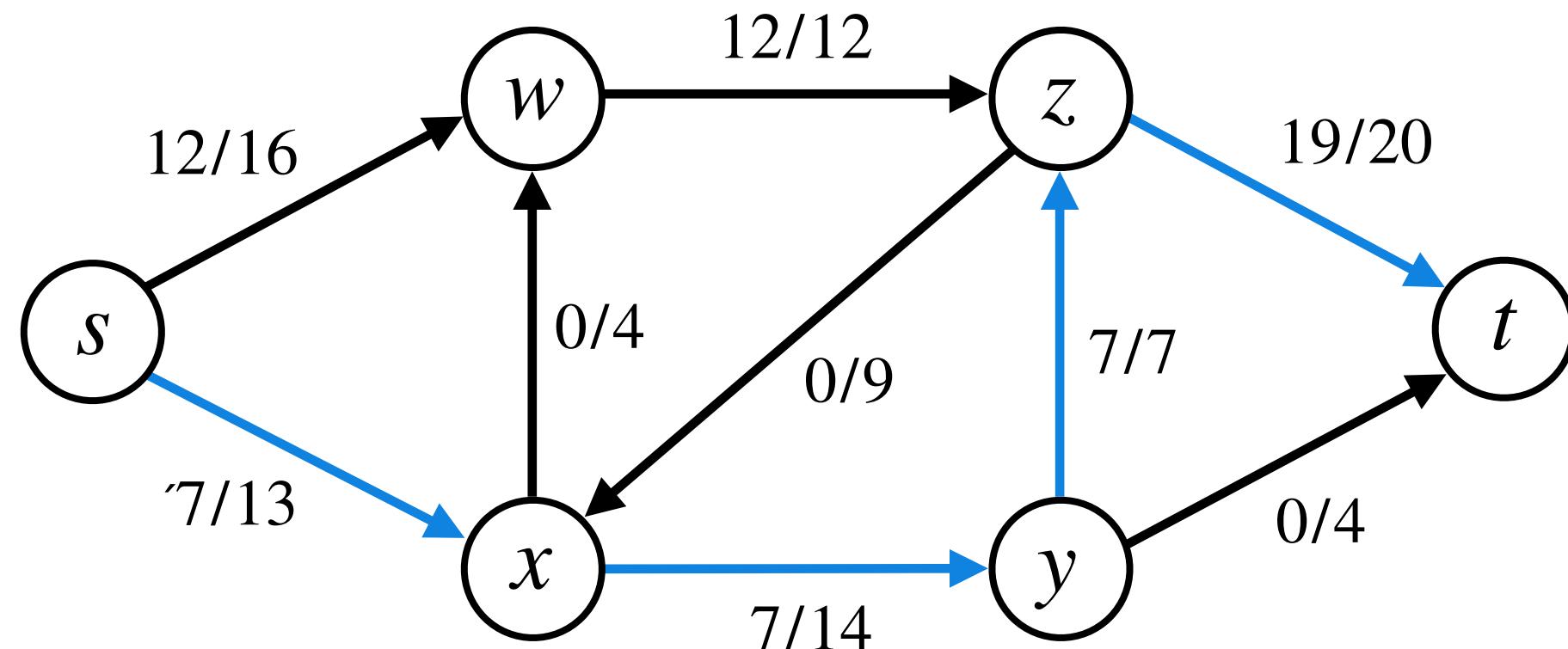


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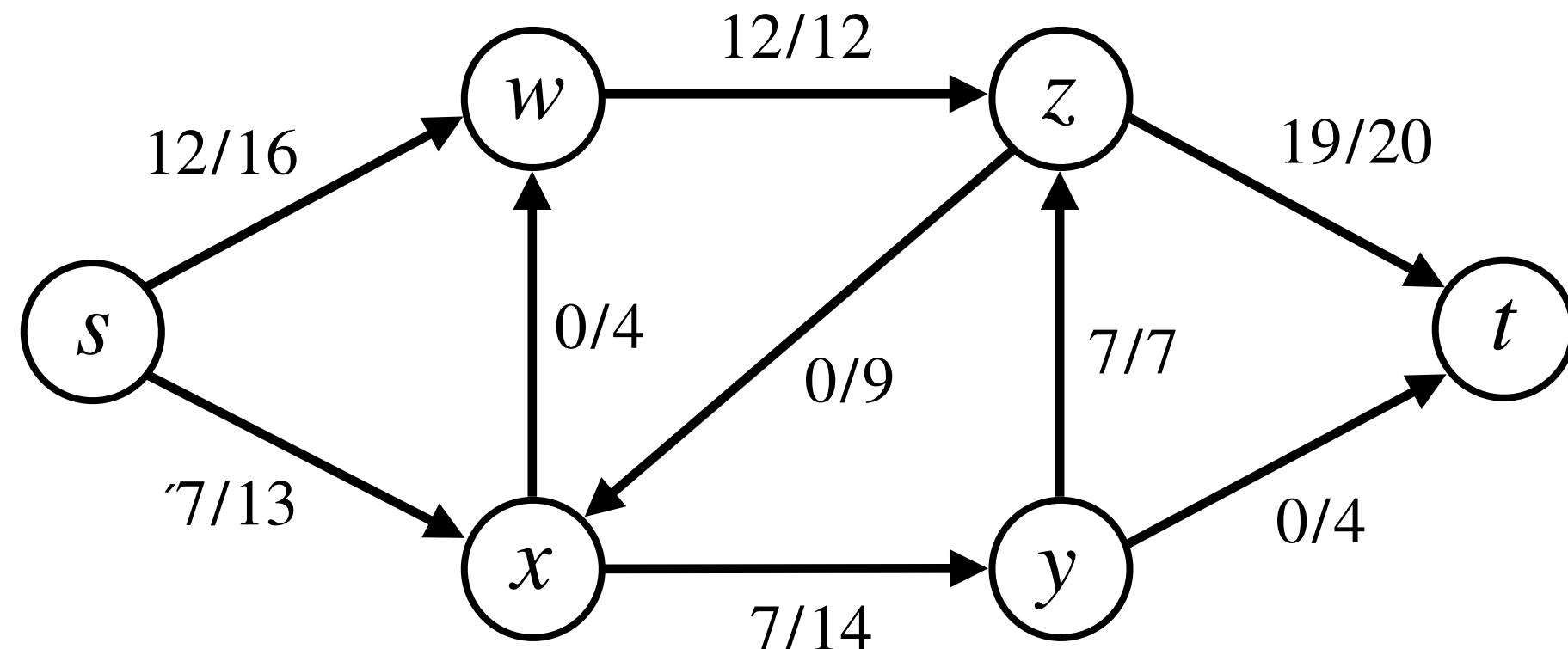


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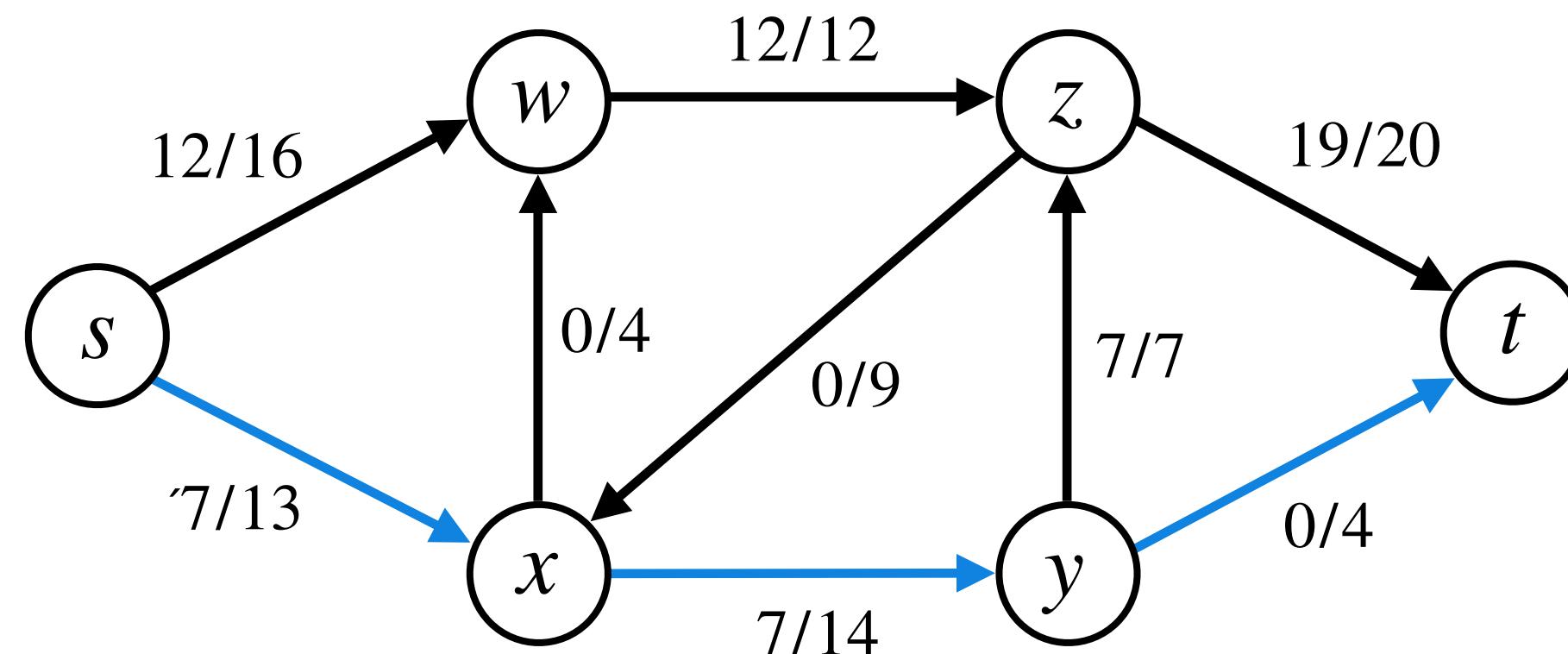


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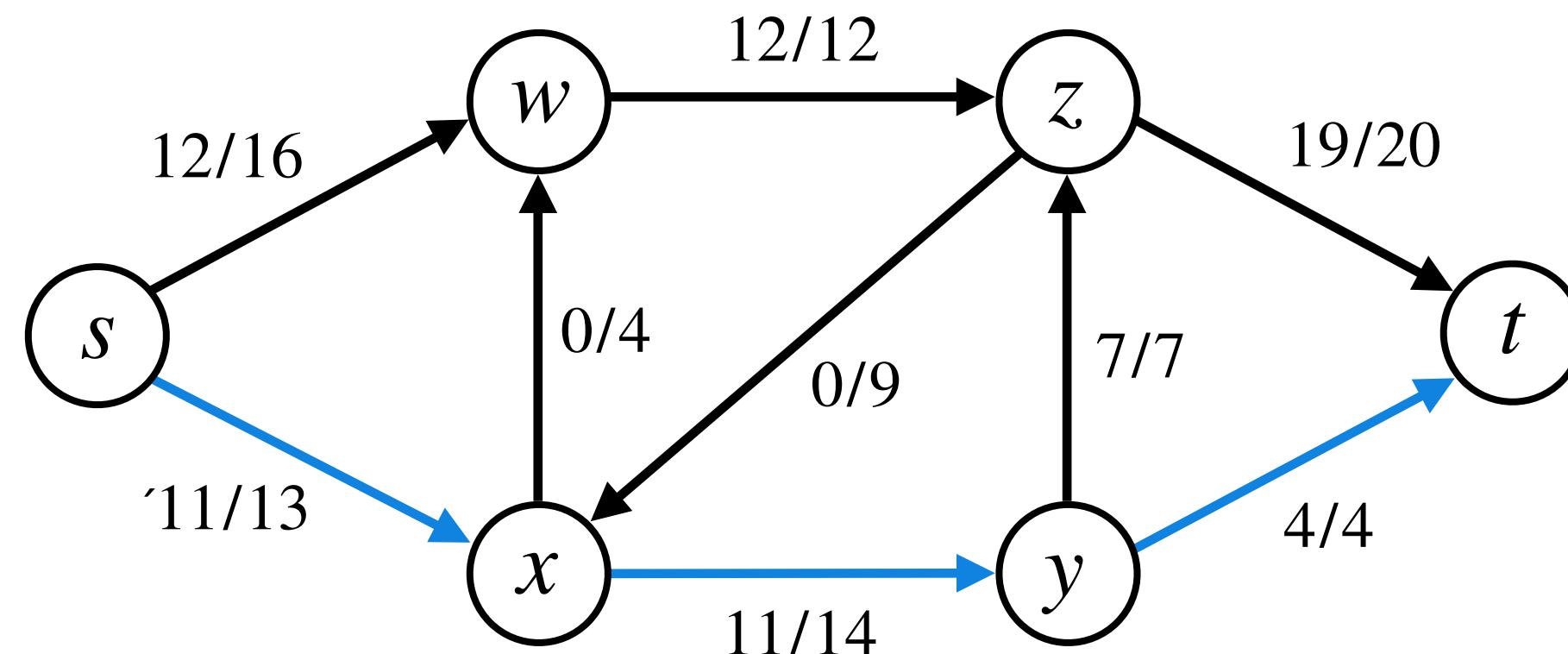


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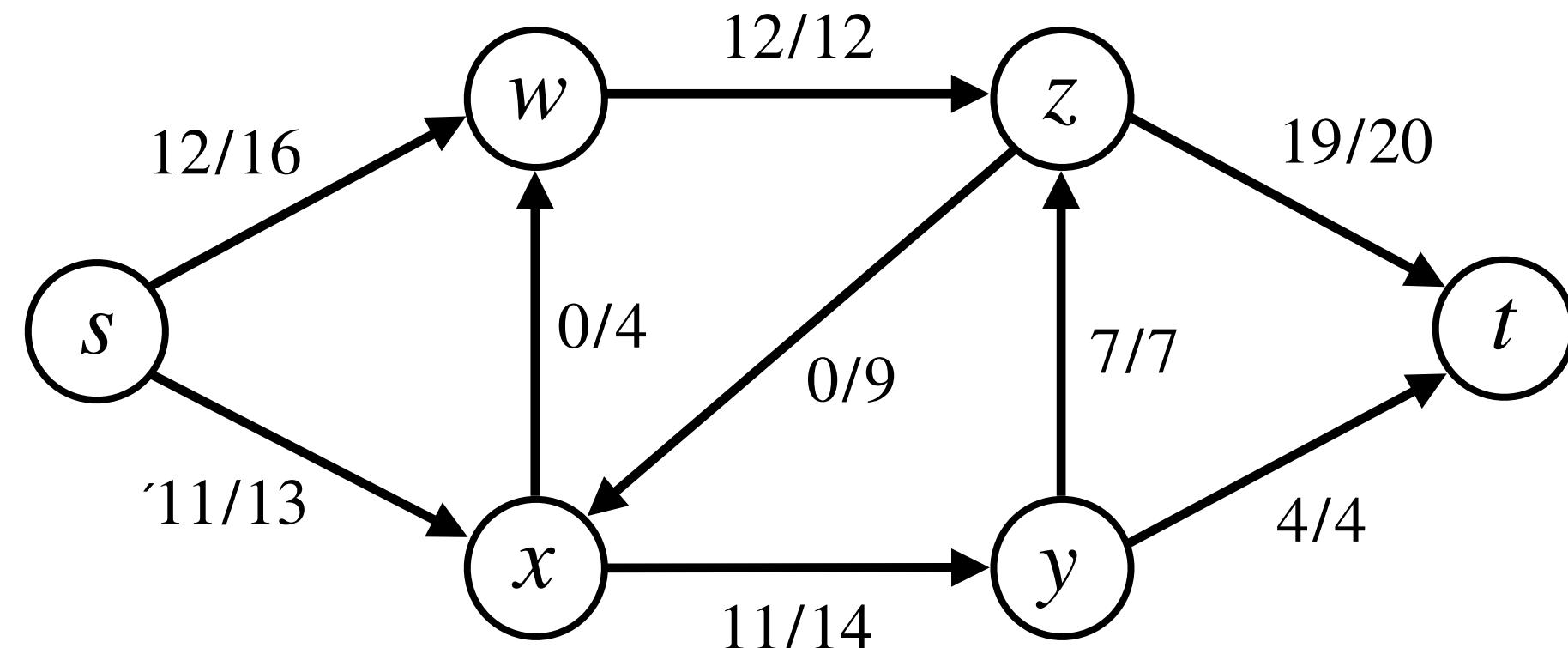


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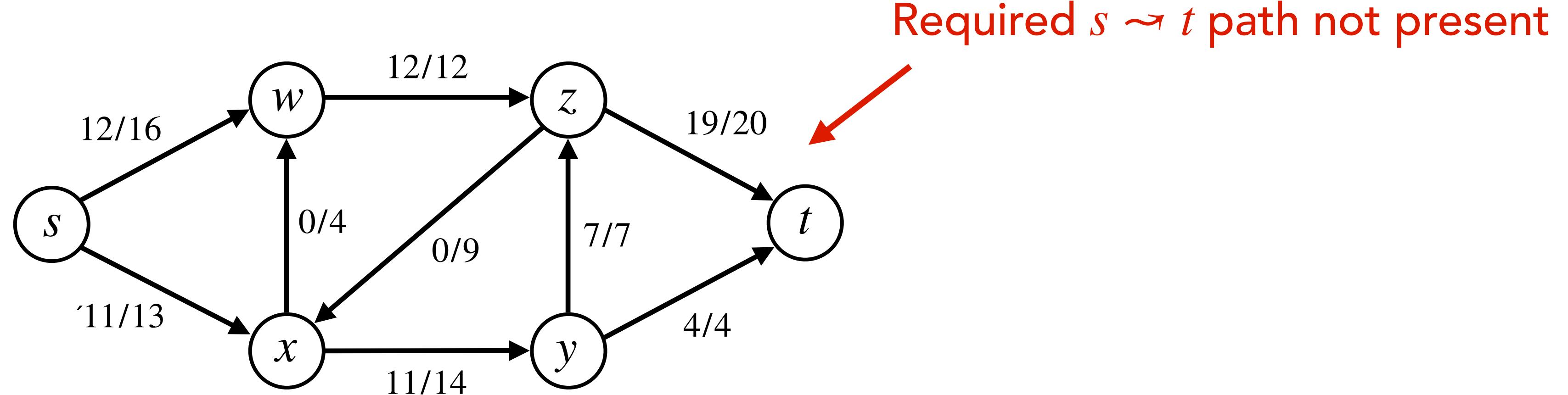


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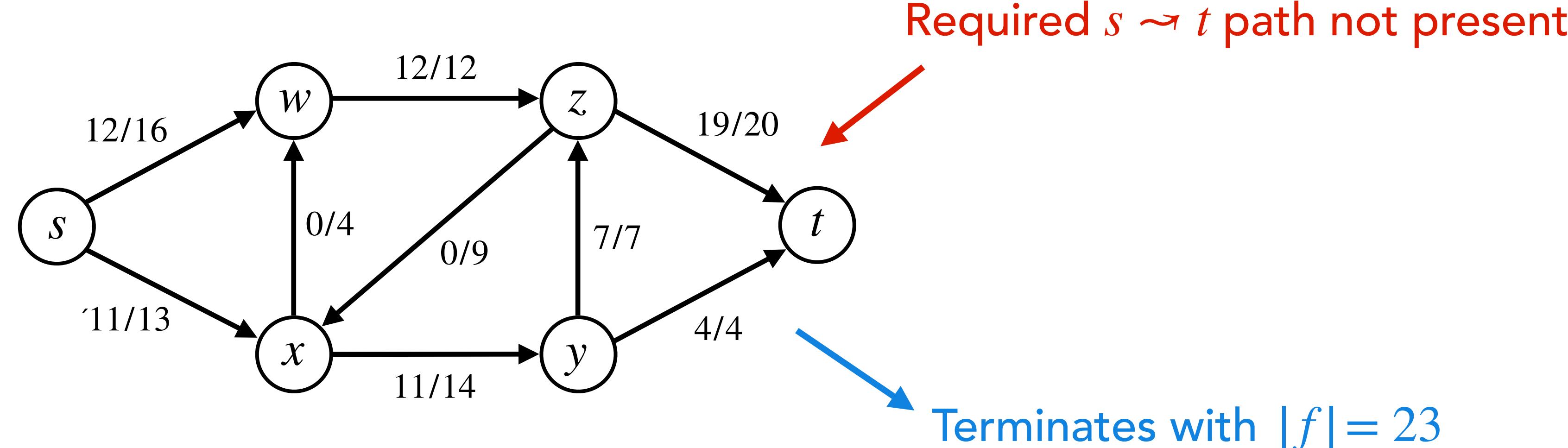


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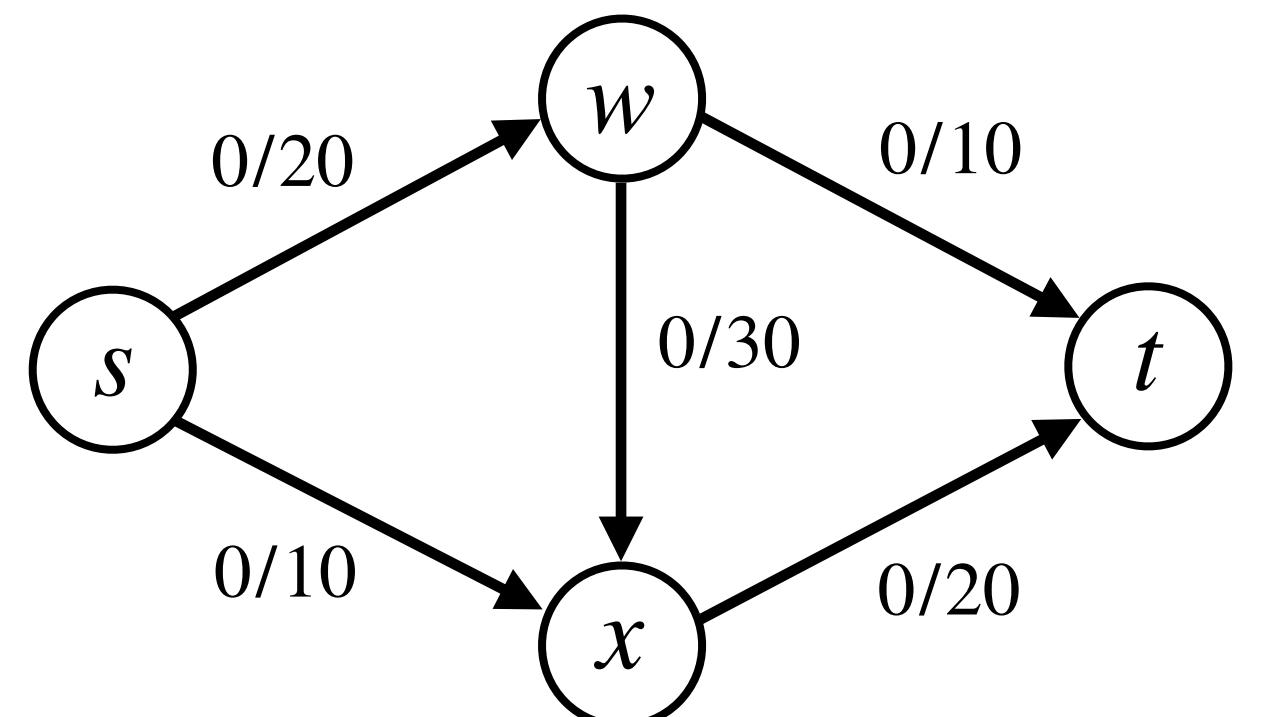
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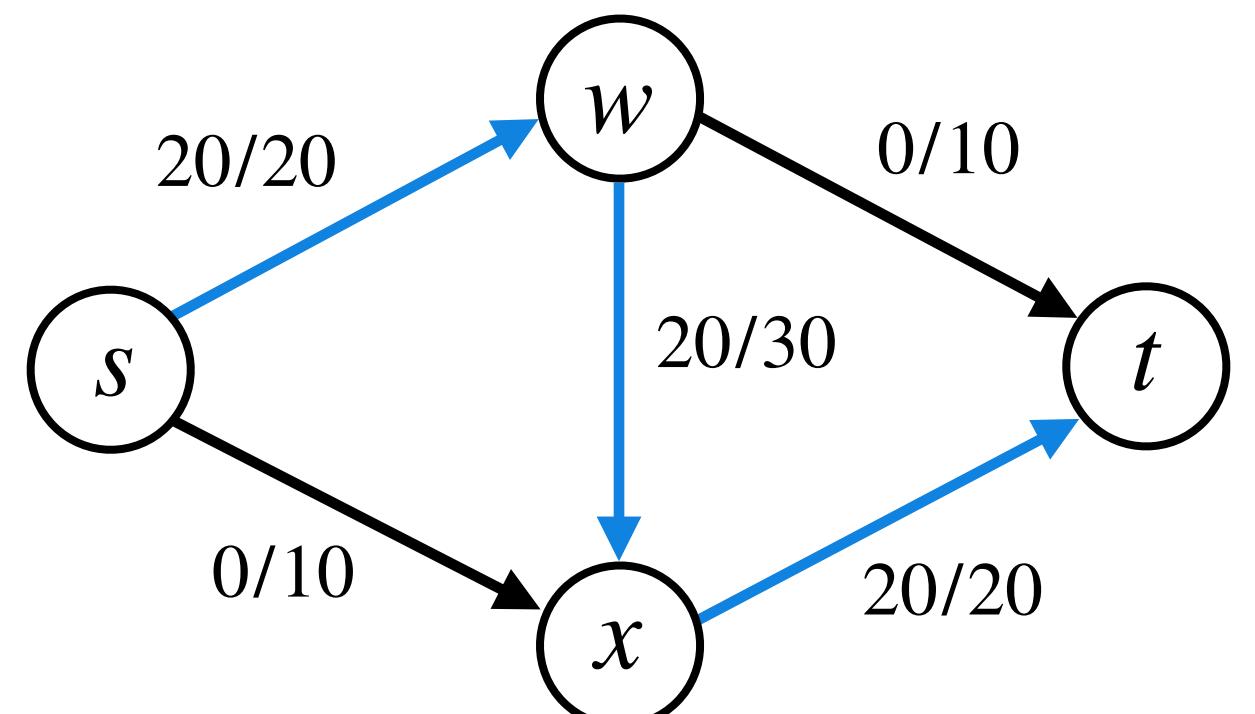


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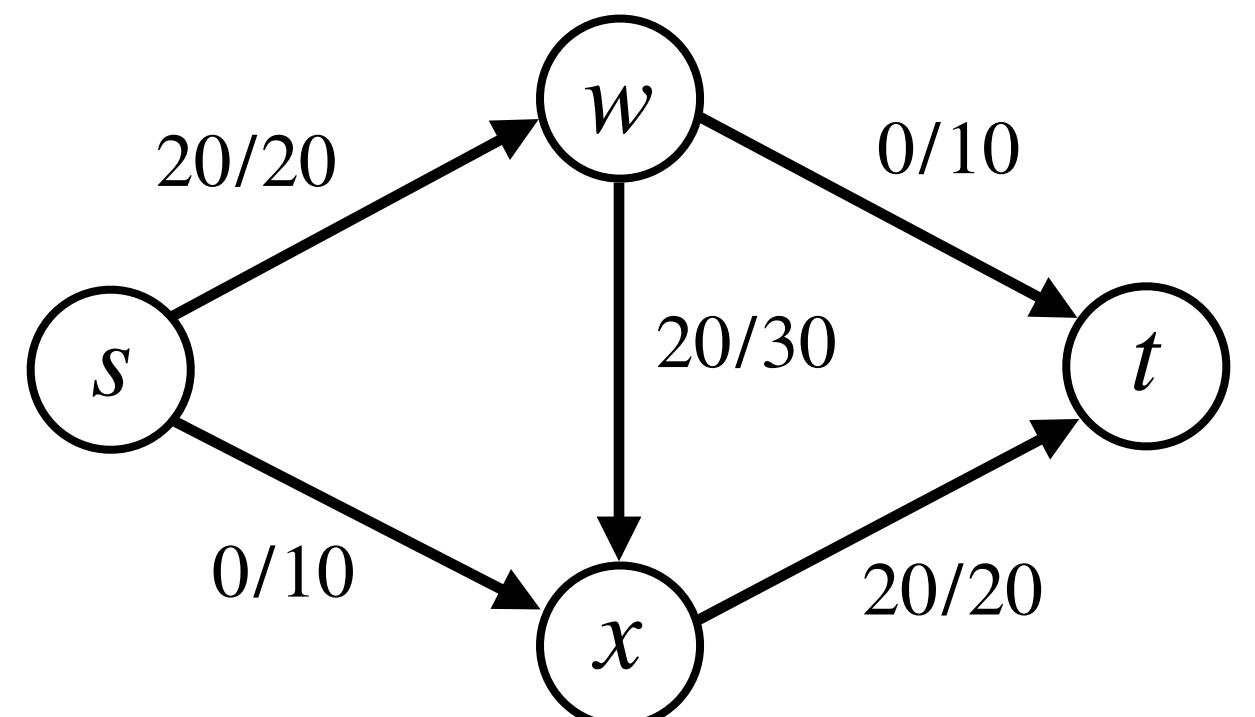


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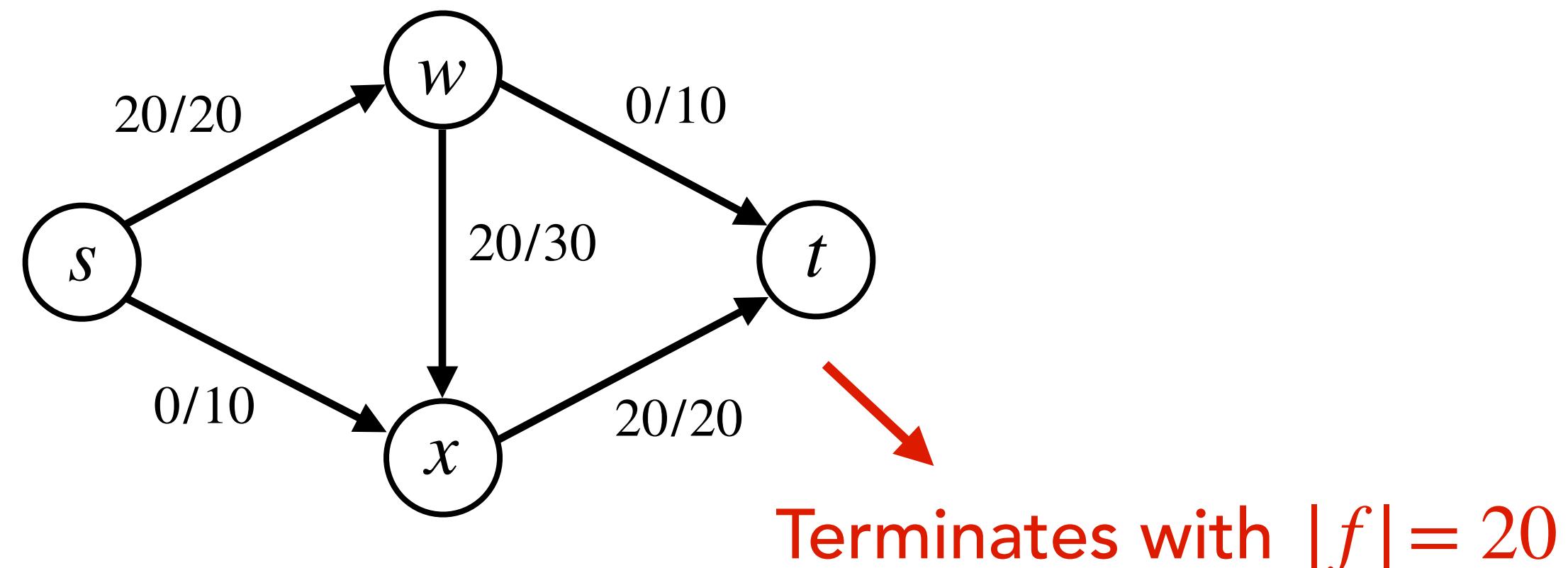


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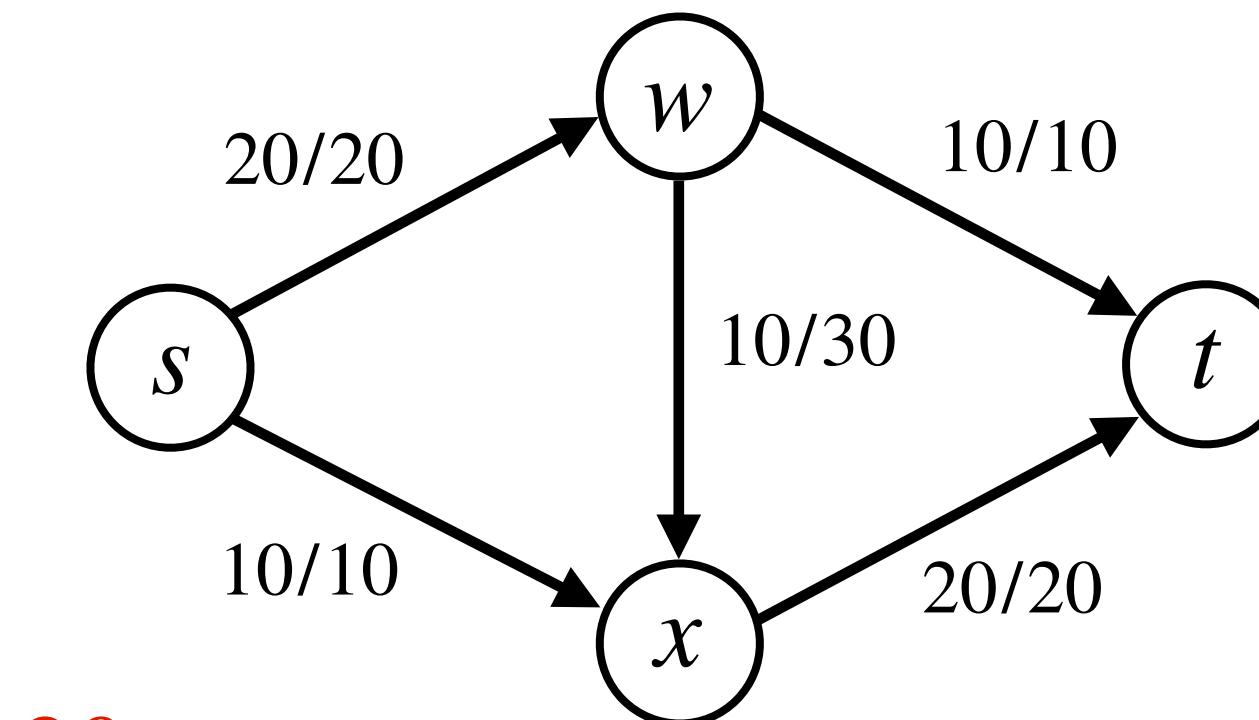
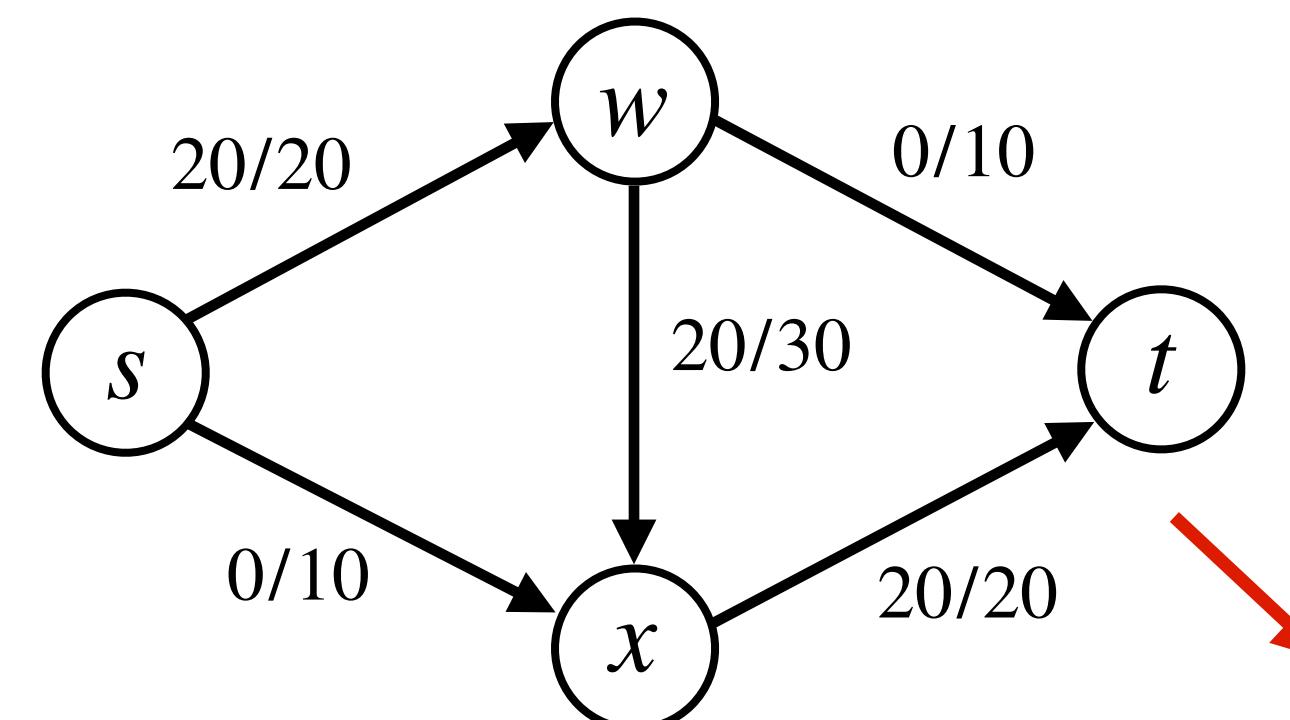


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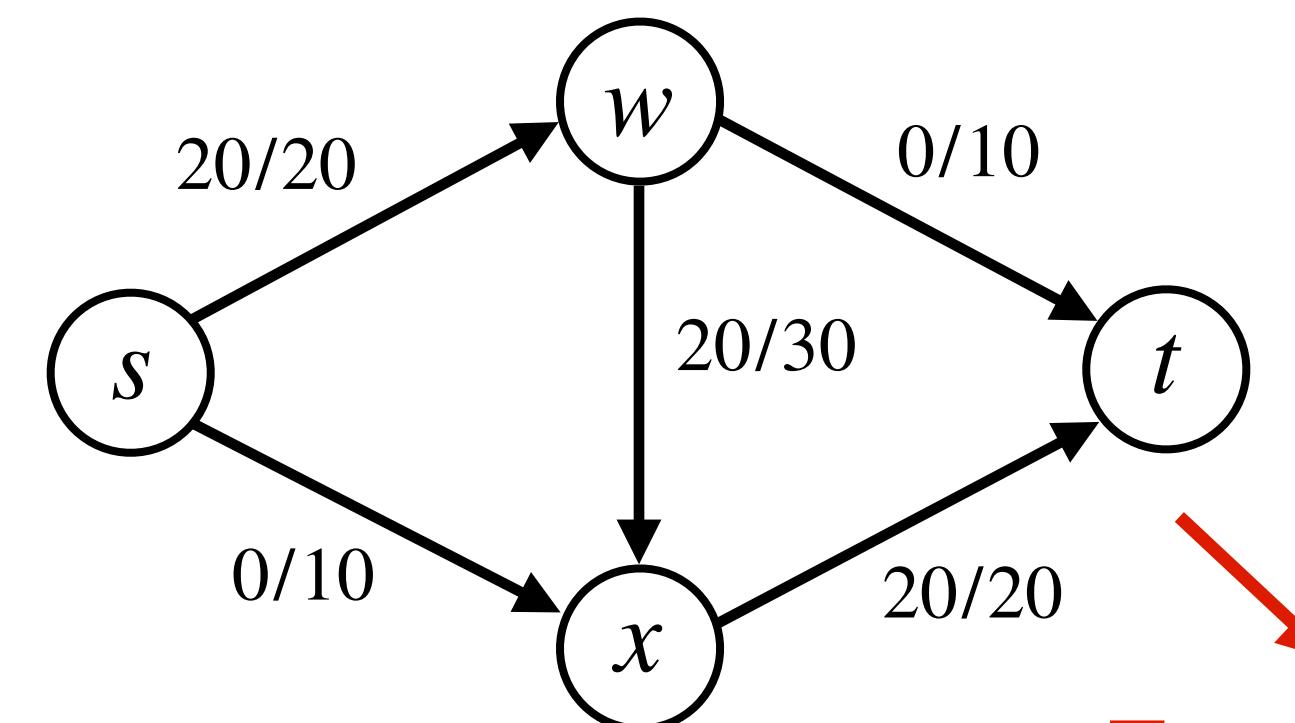
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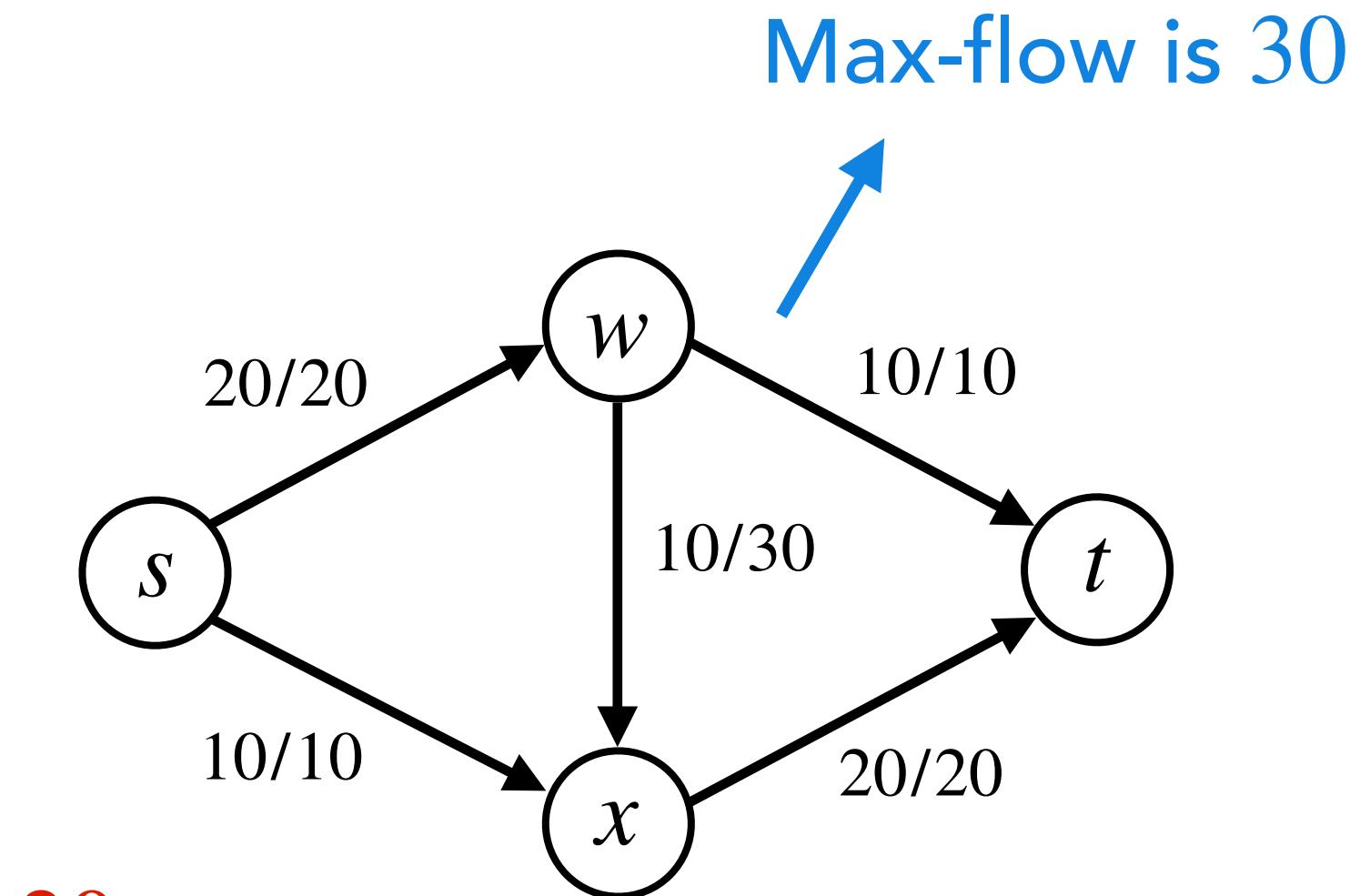
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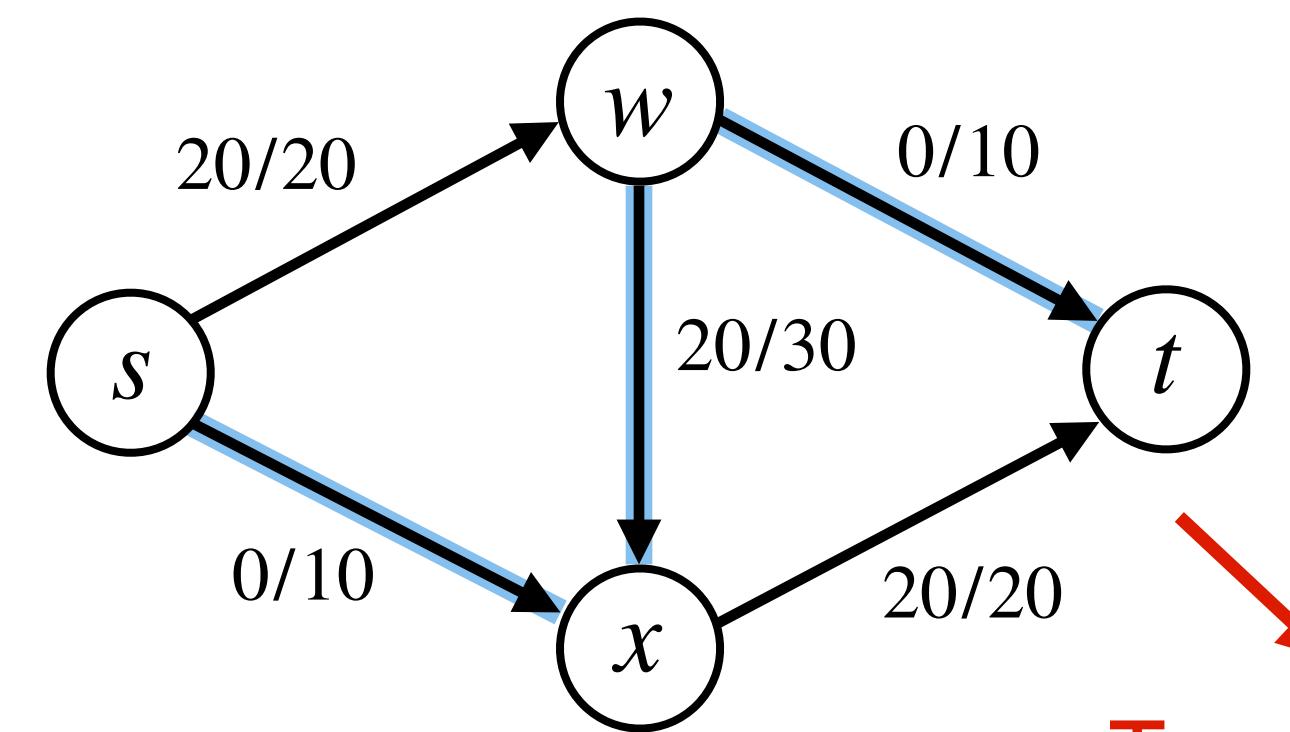


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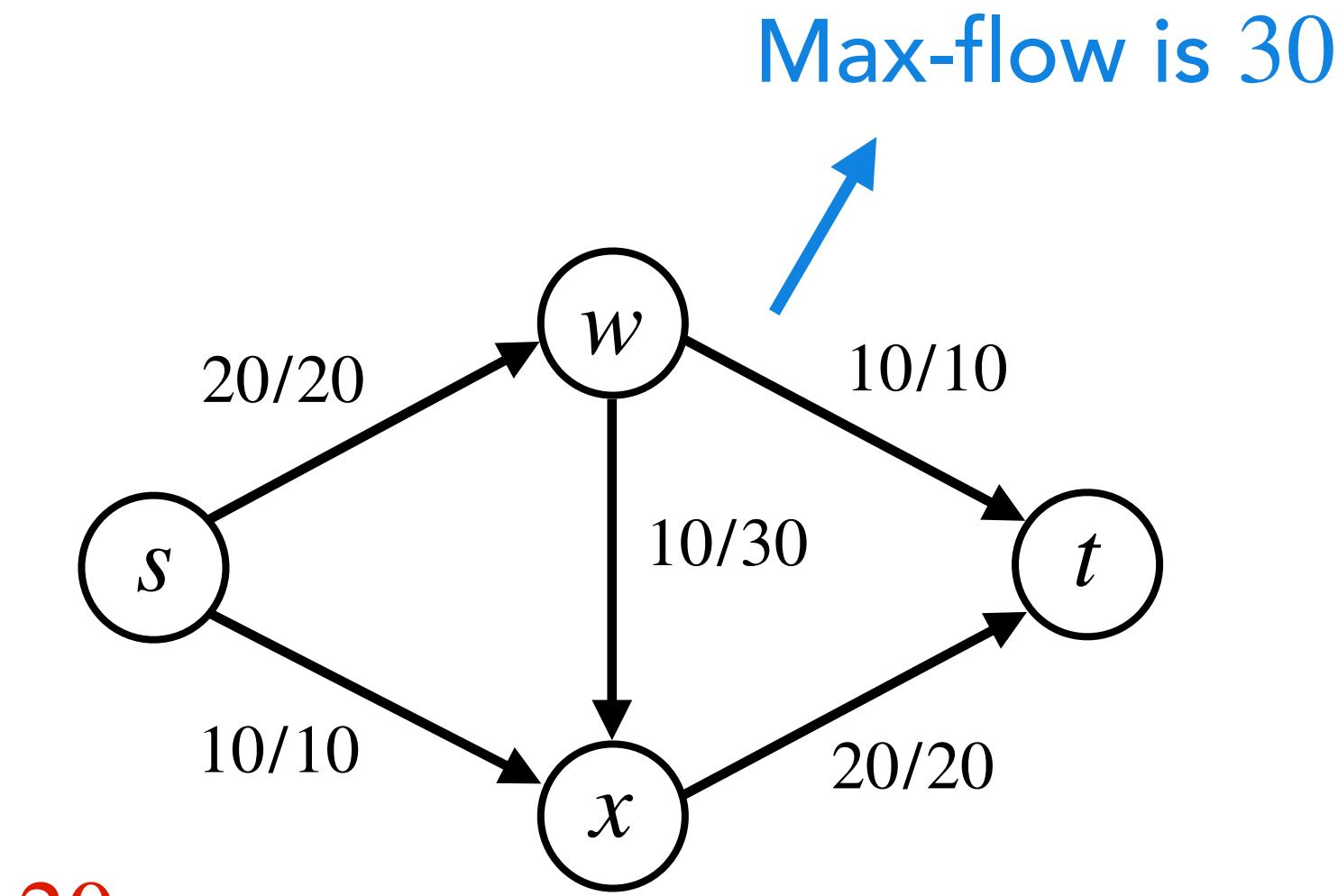
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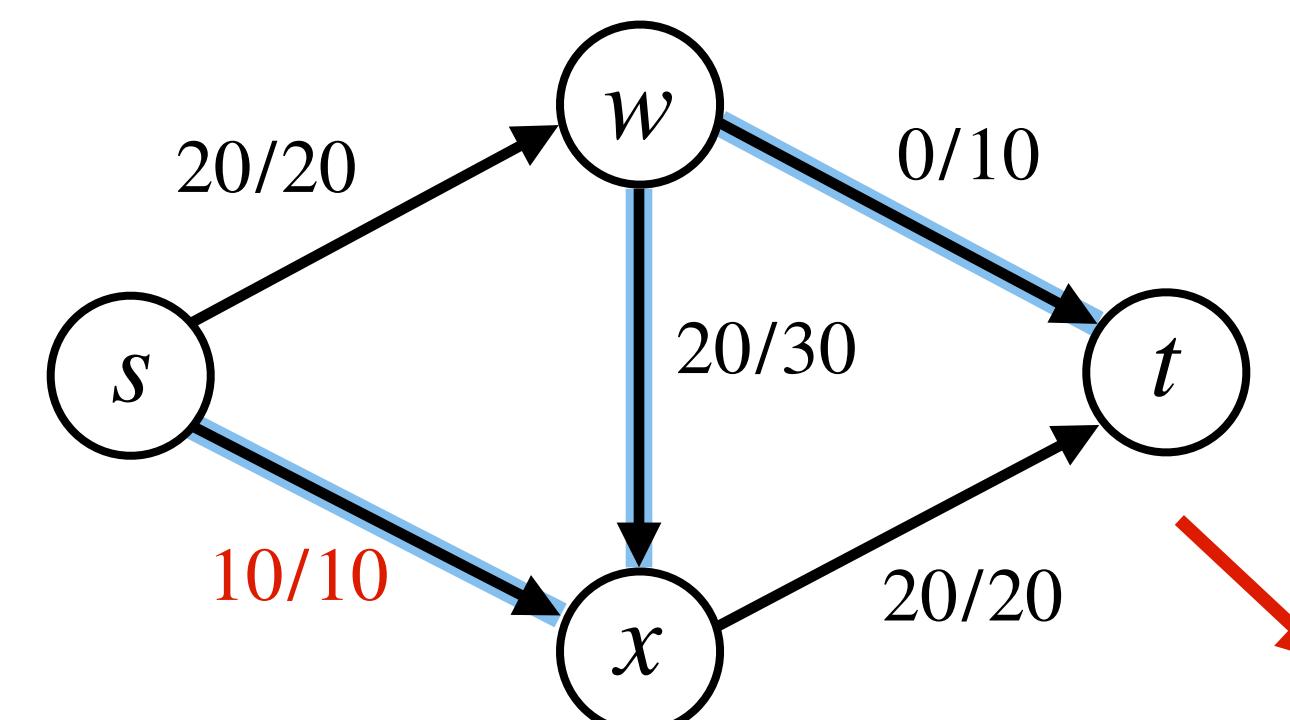
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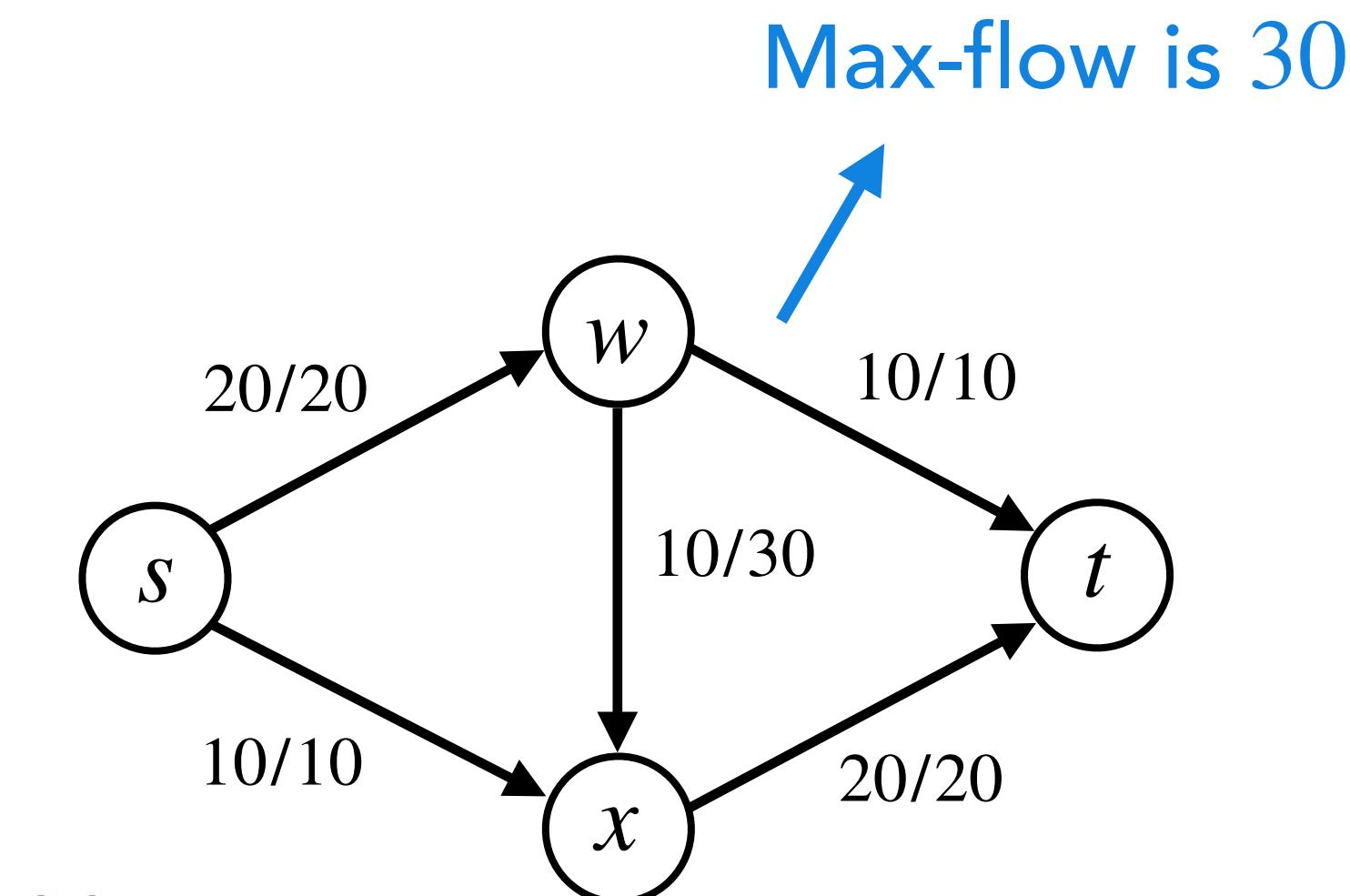
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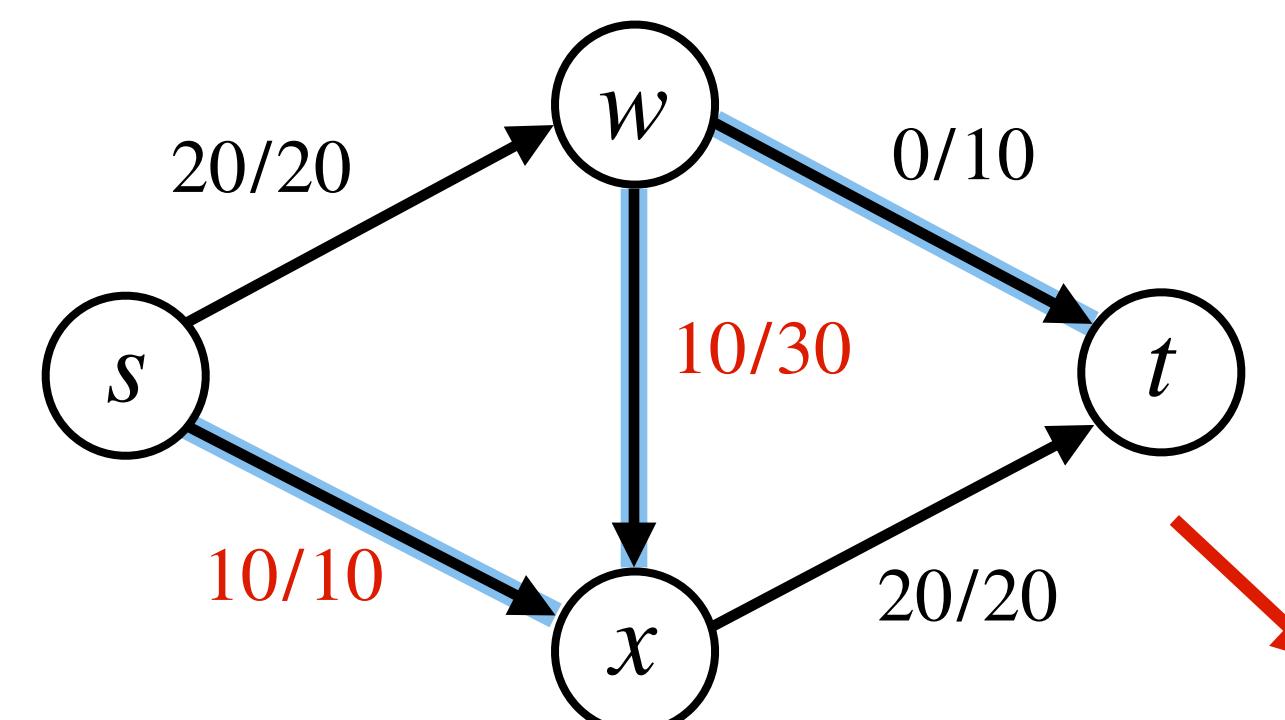
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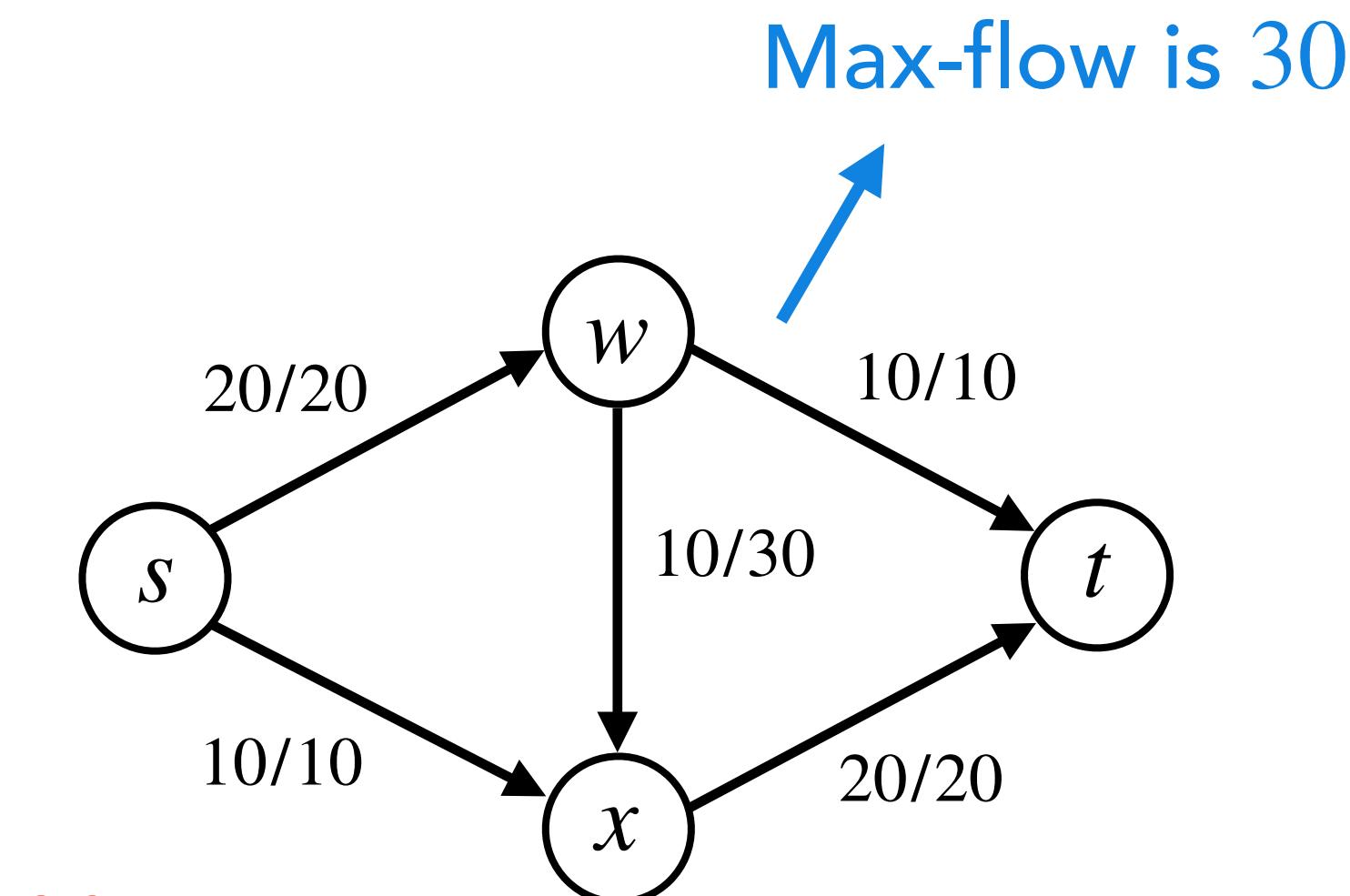
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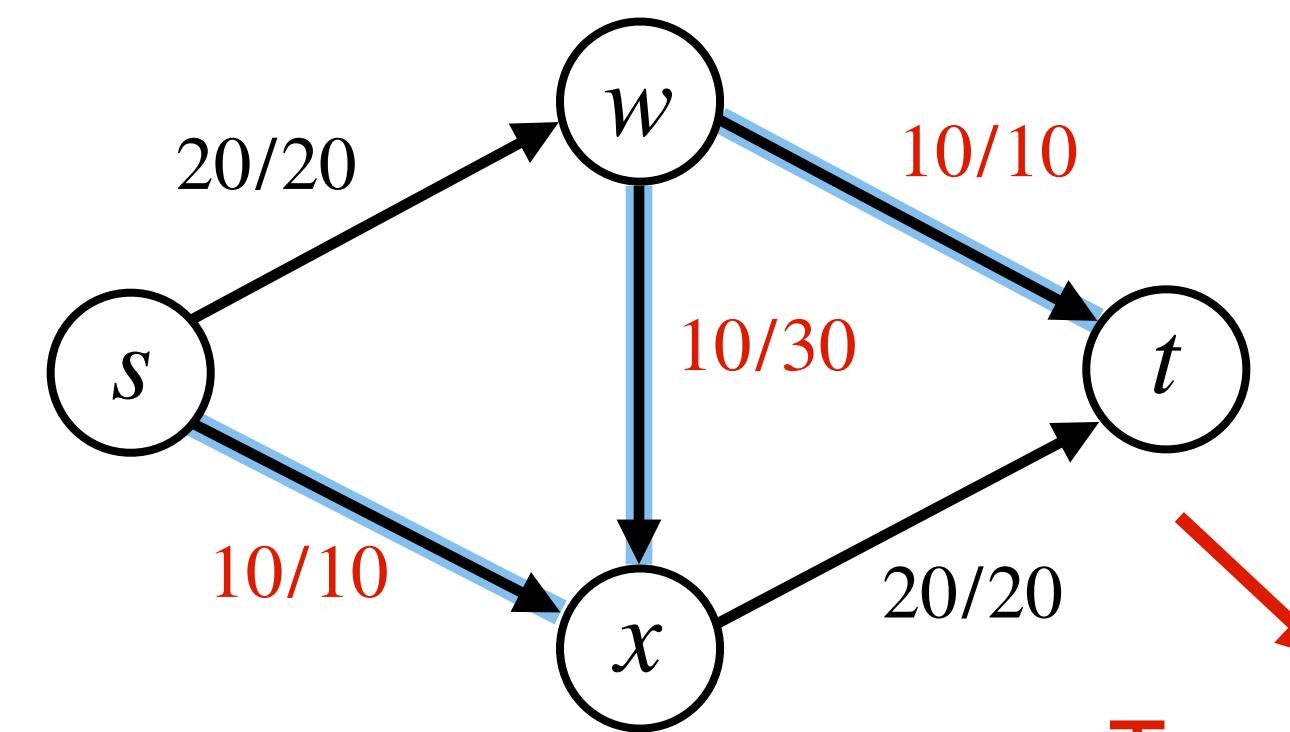
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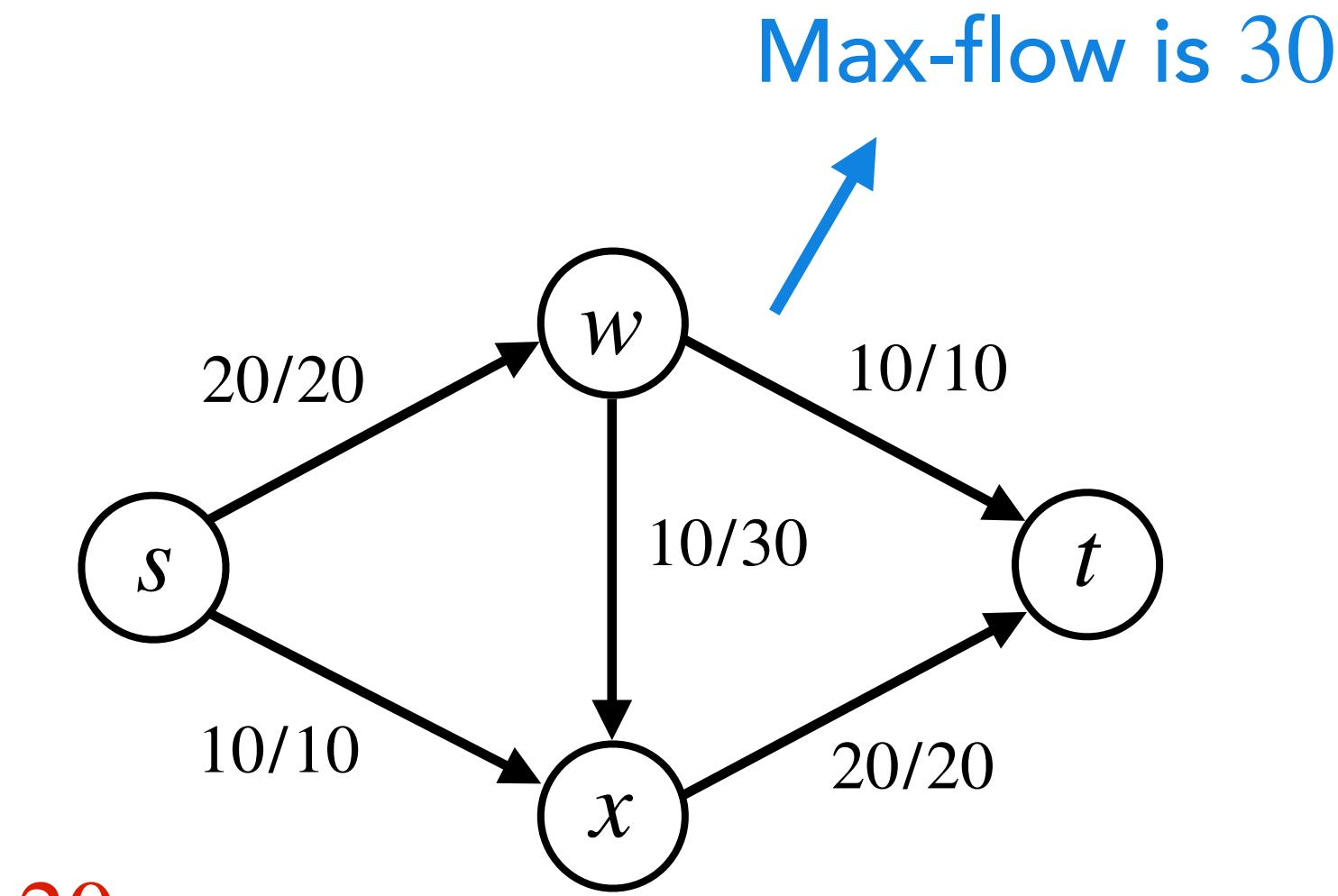
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Need to learn a new structure for that!

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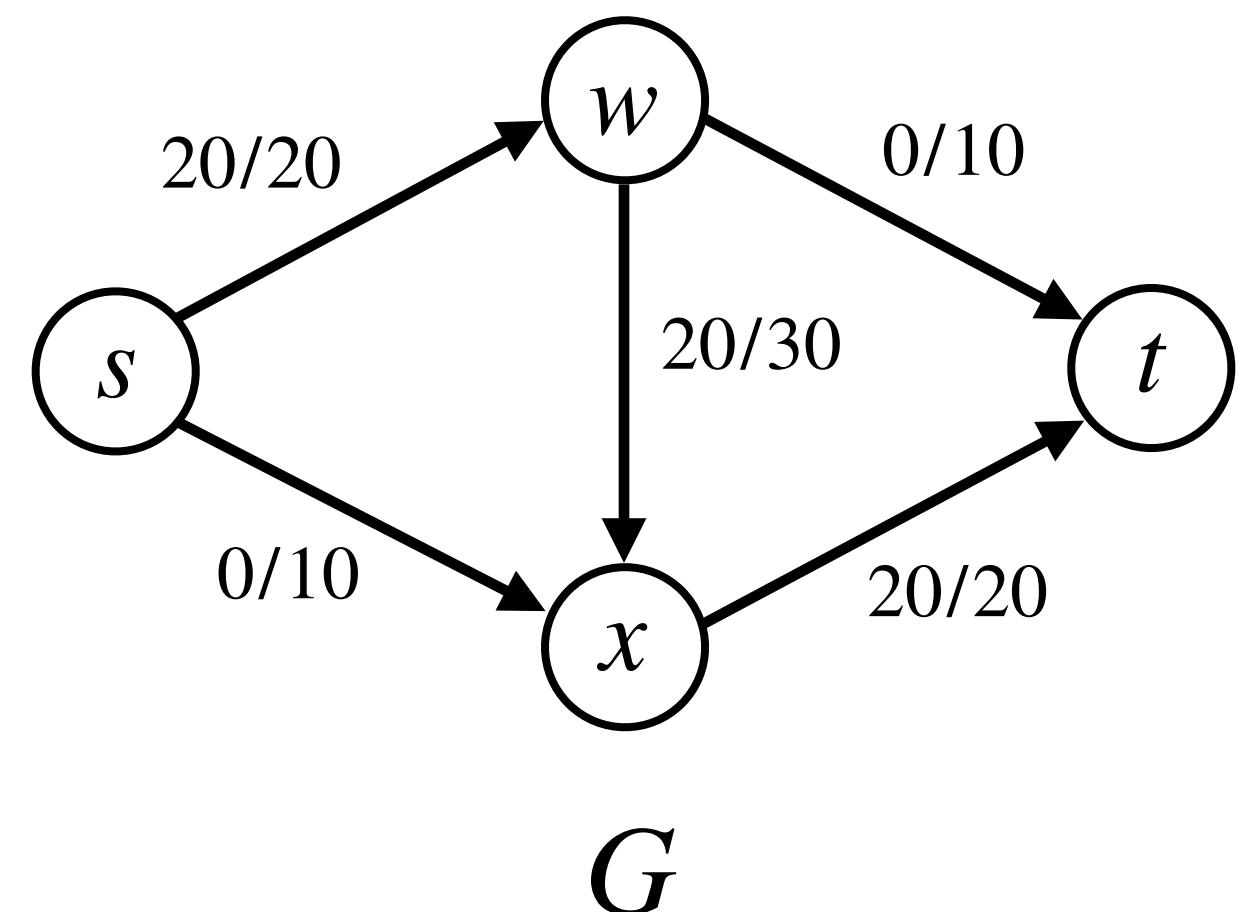
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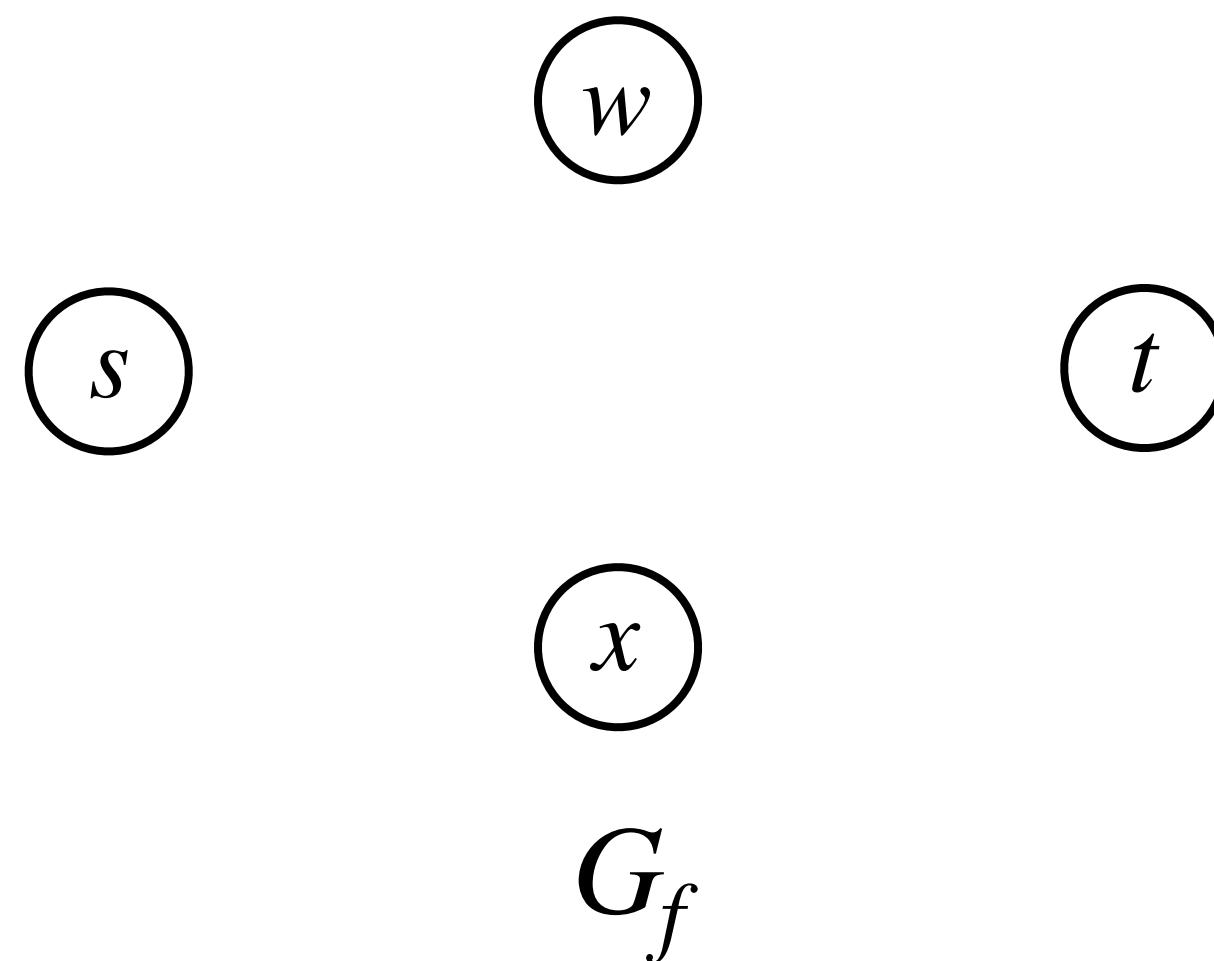
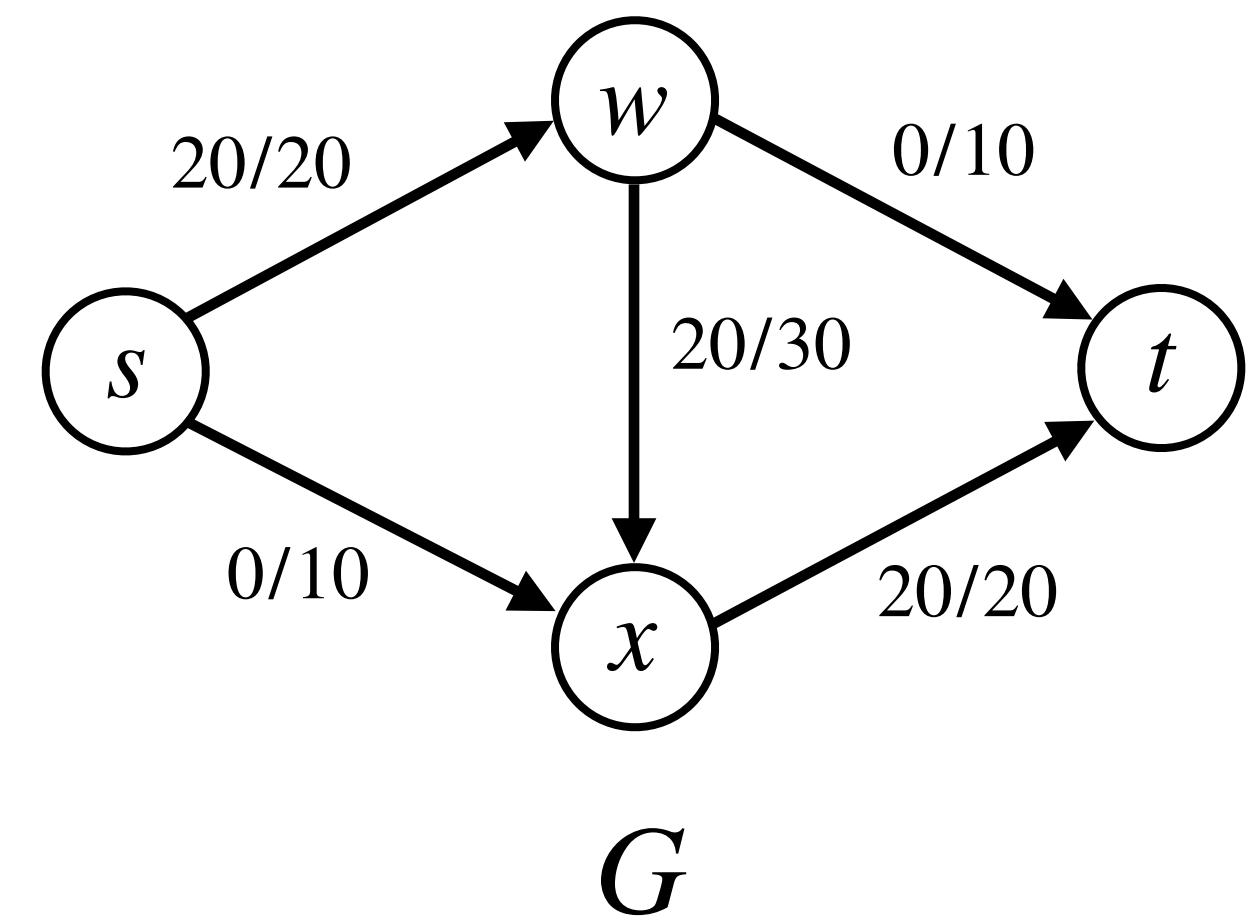


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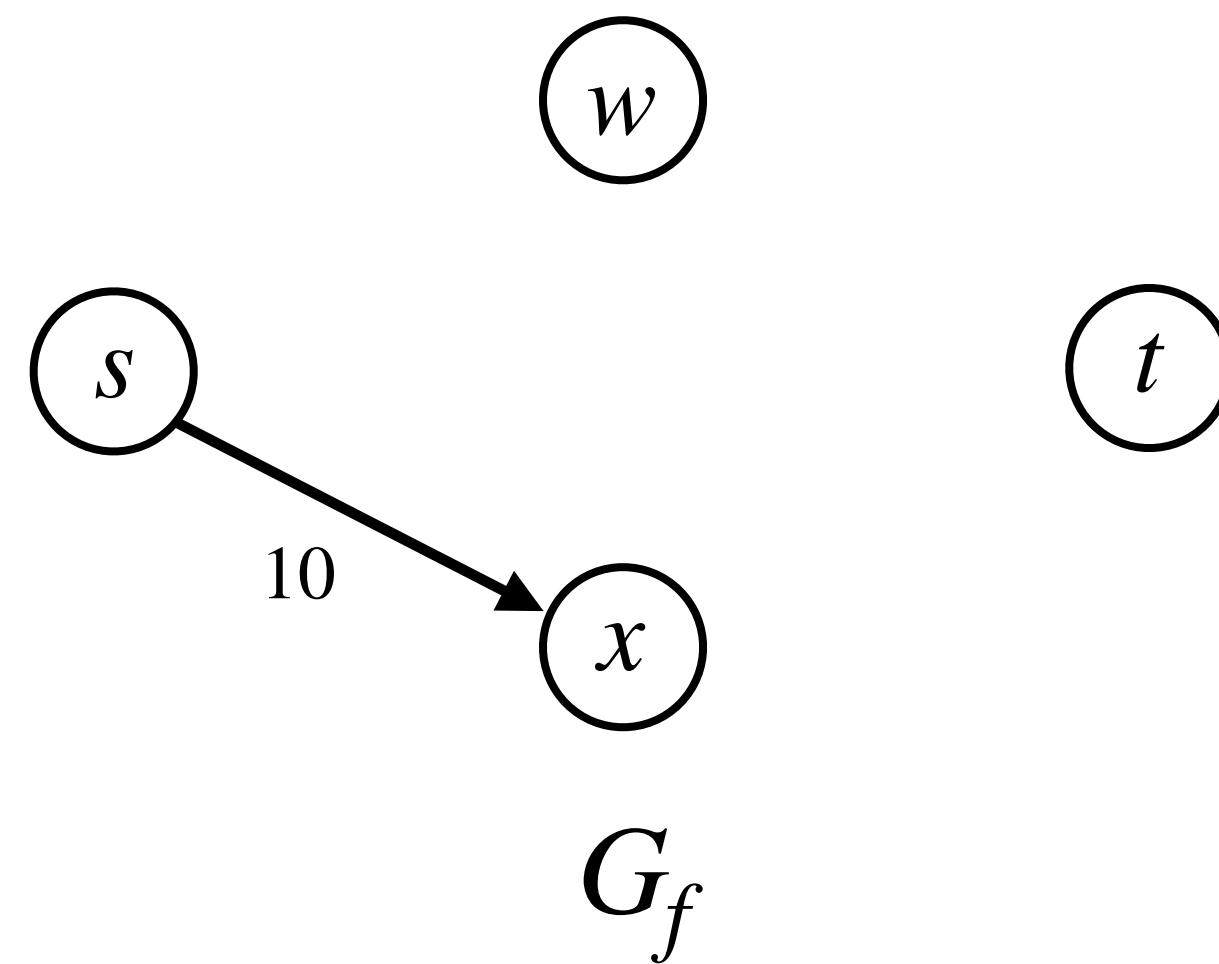
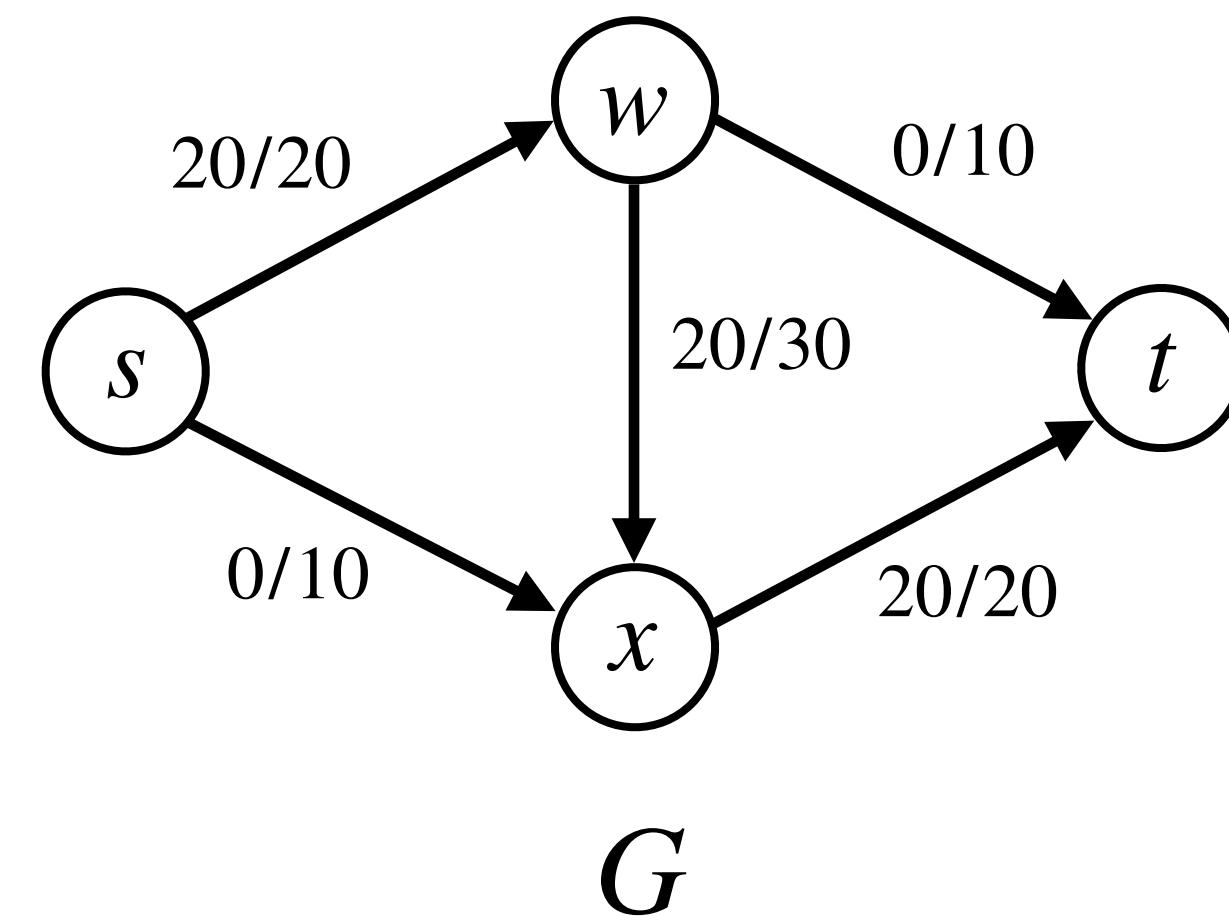


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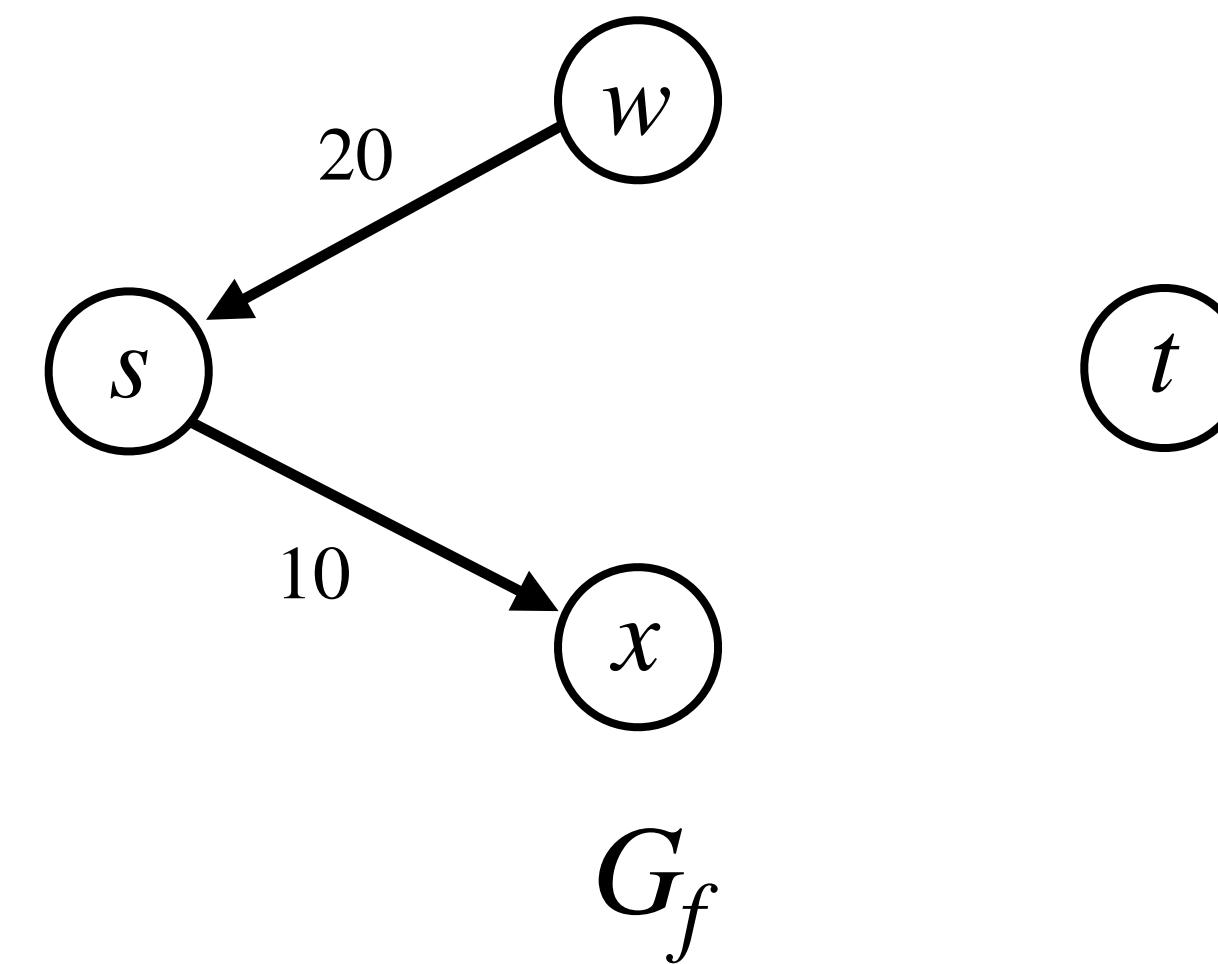
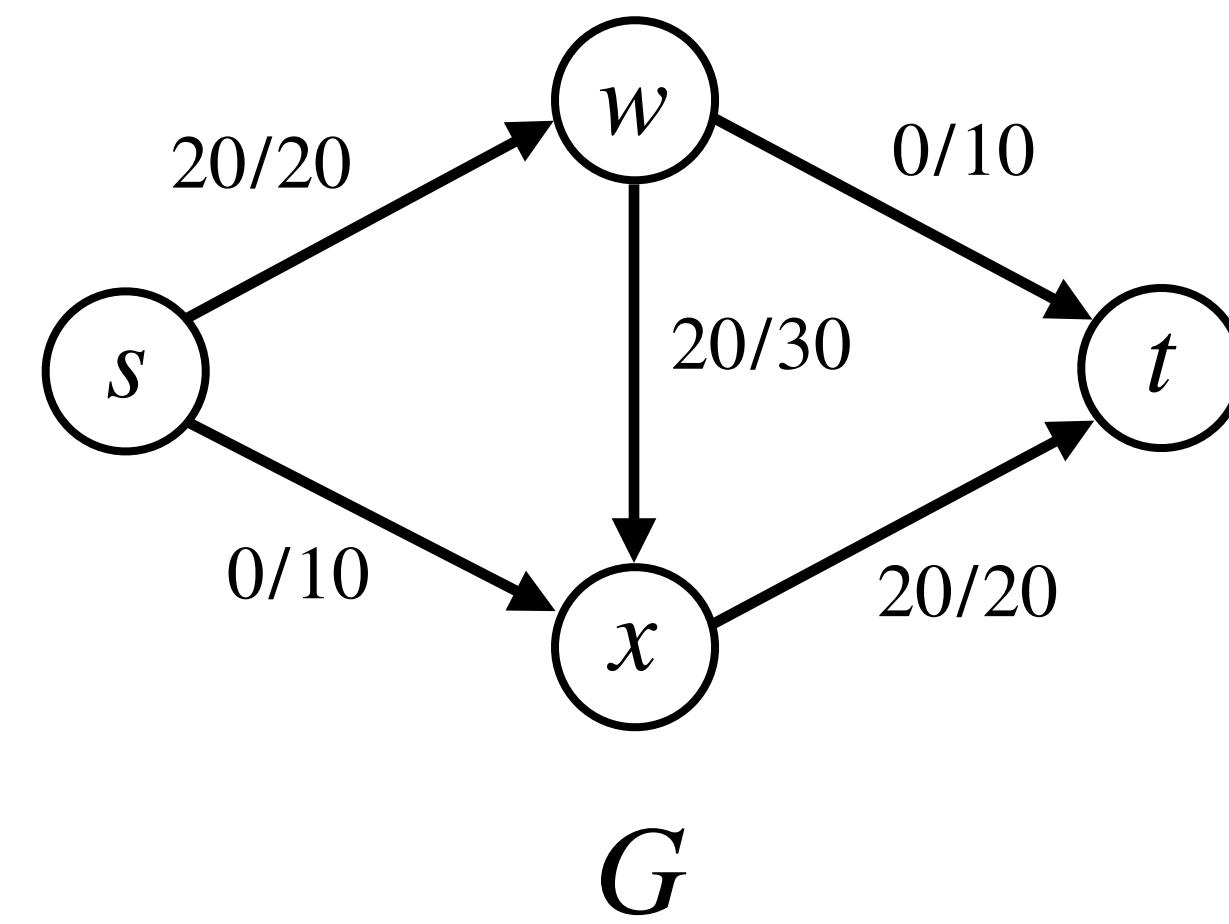


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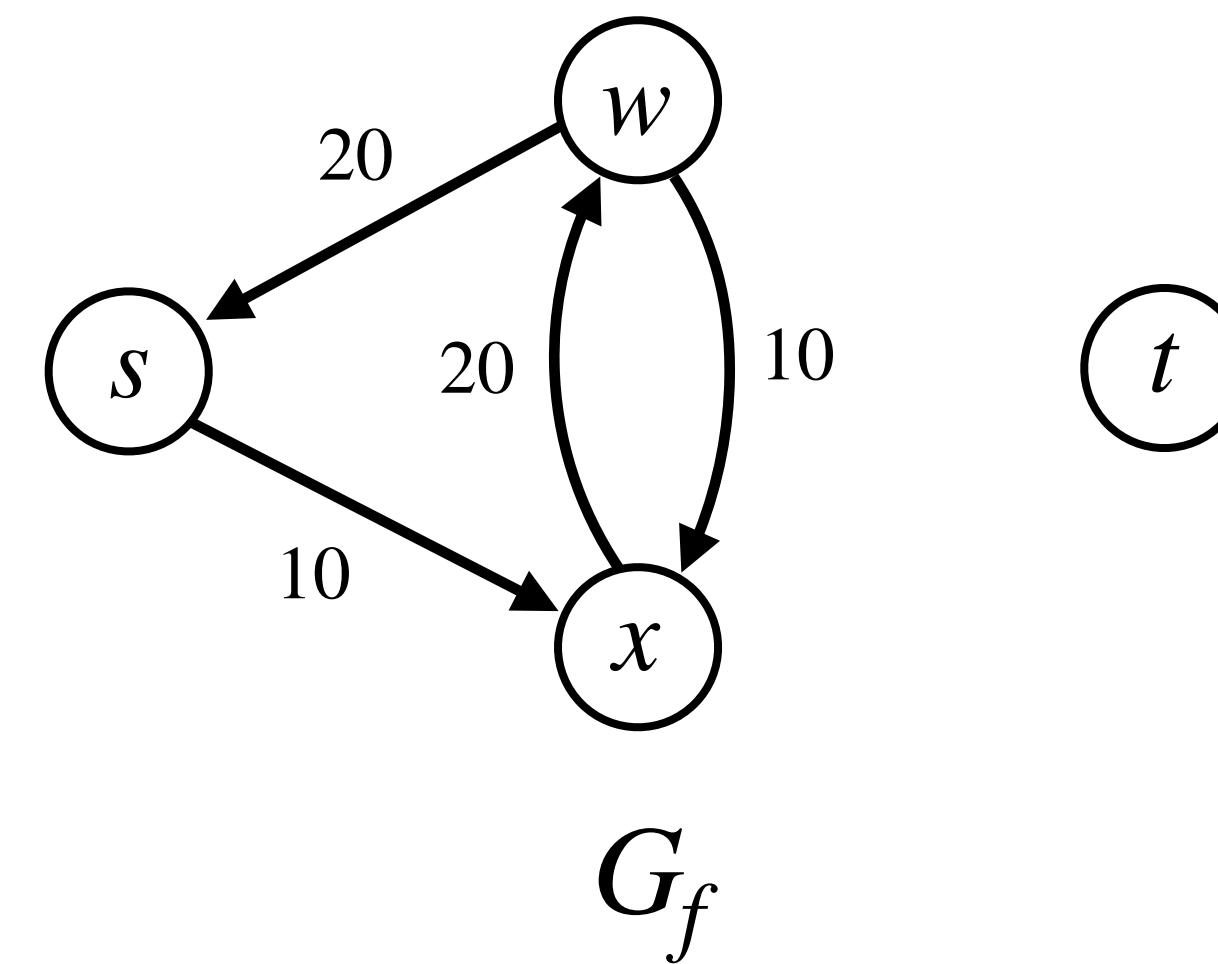
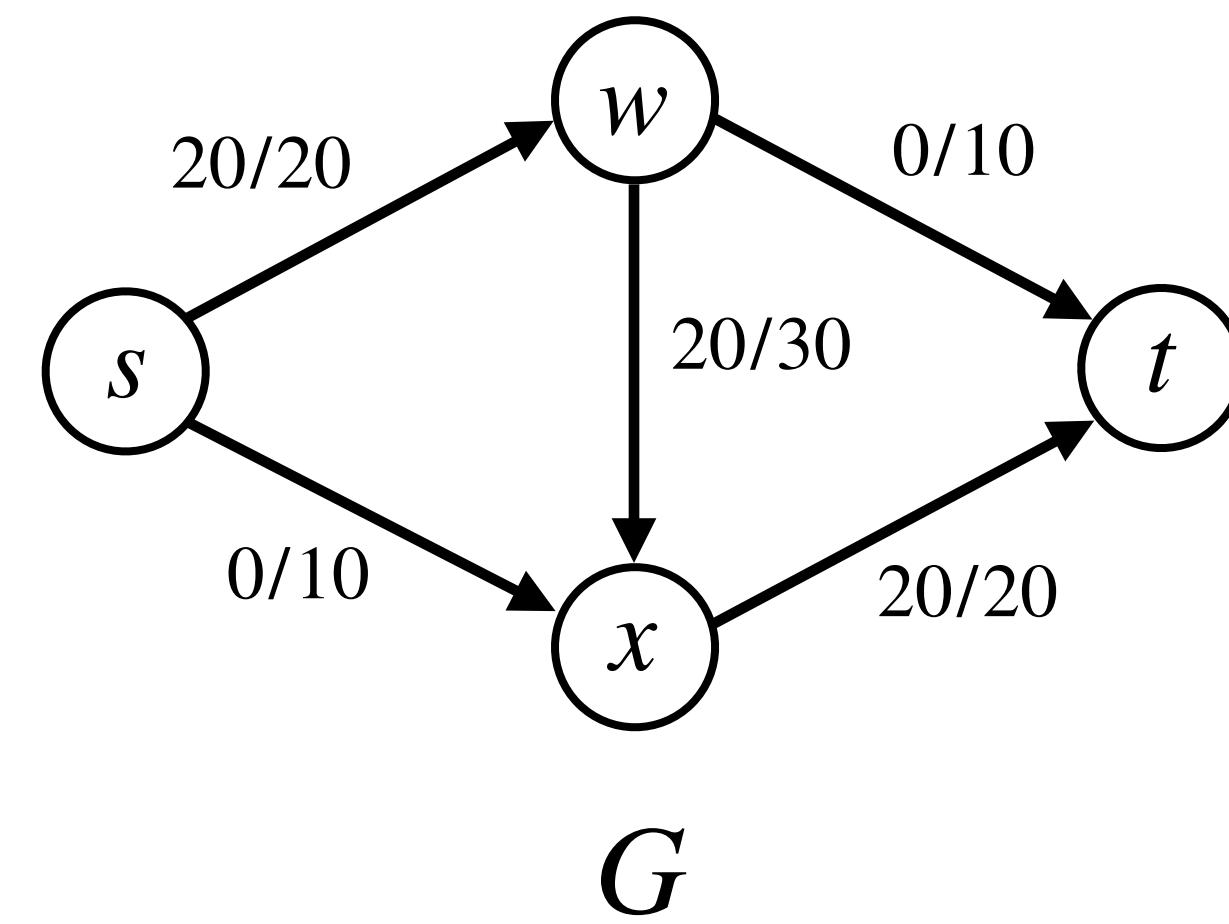


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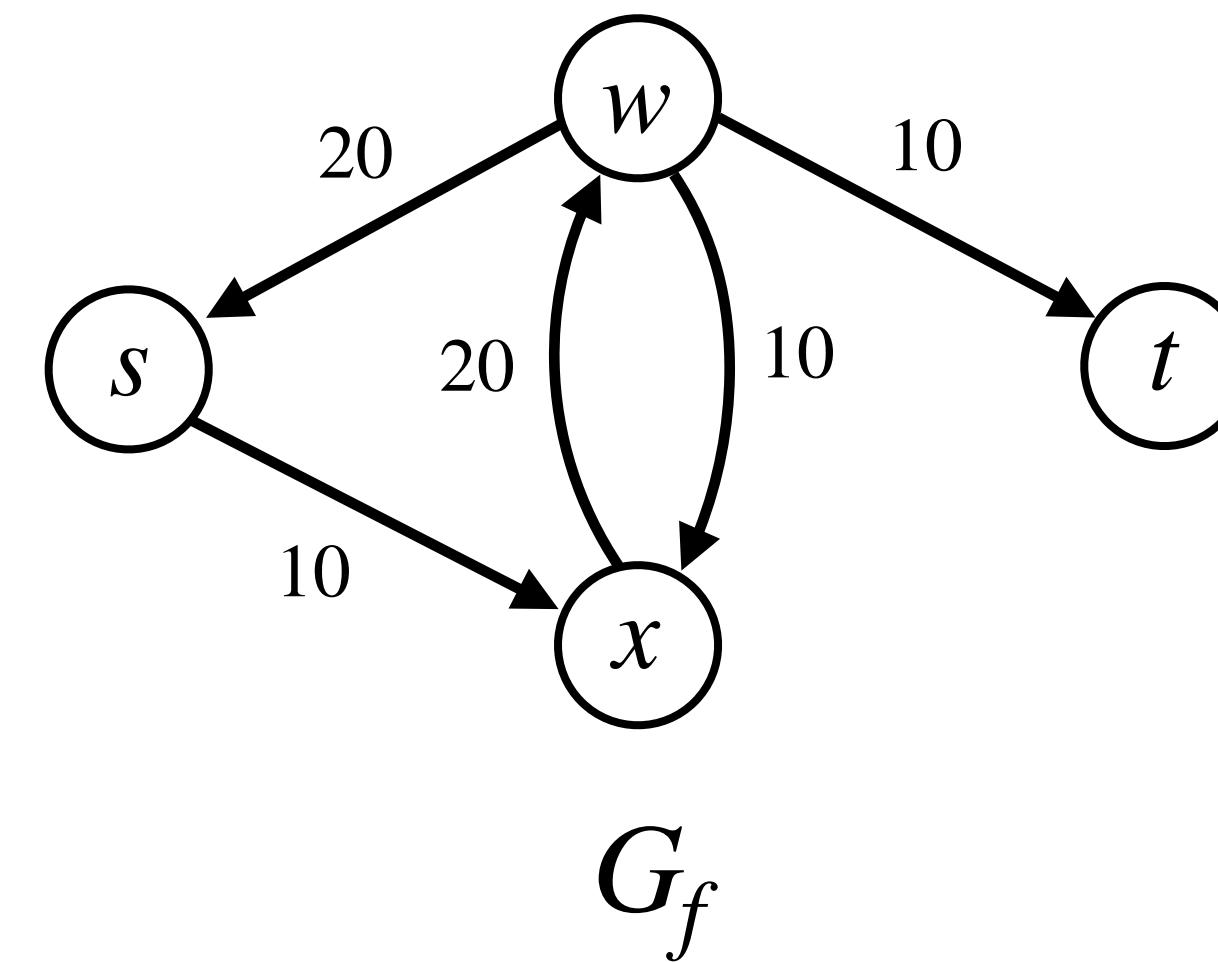
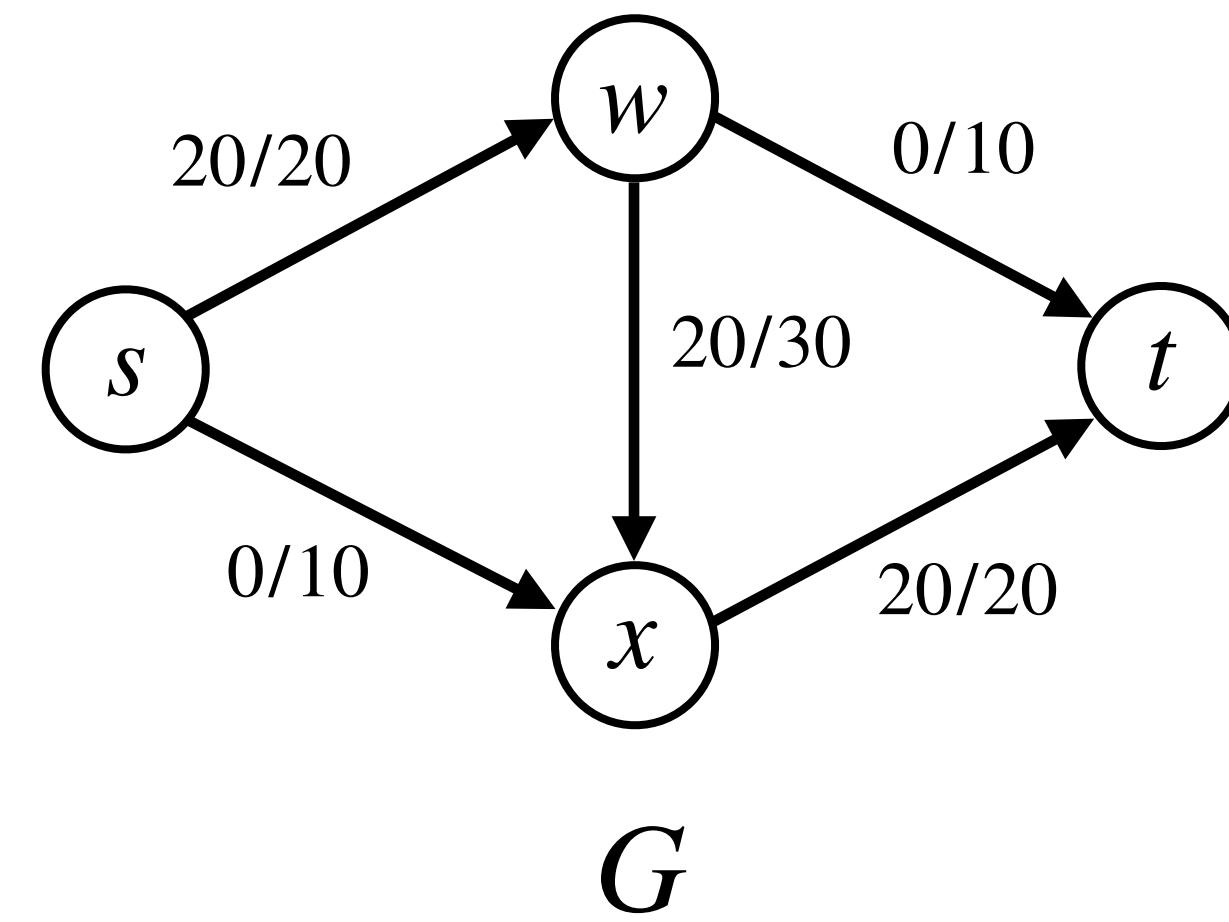


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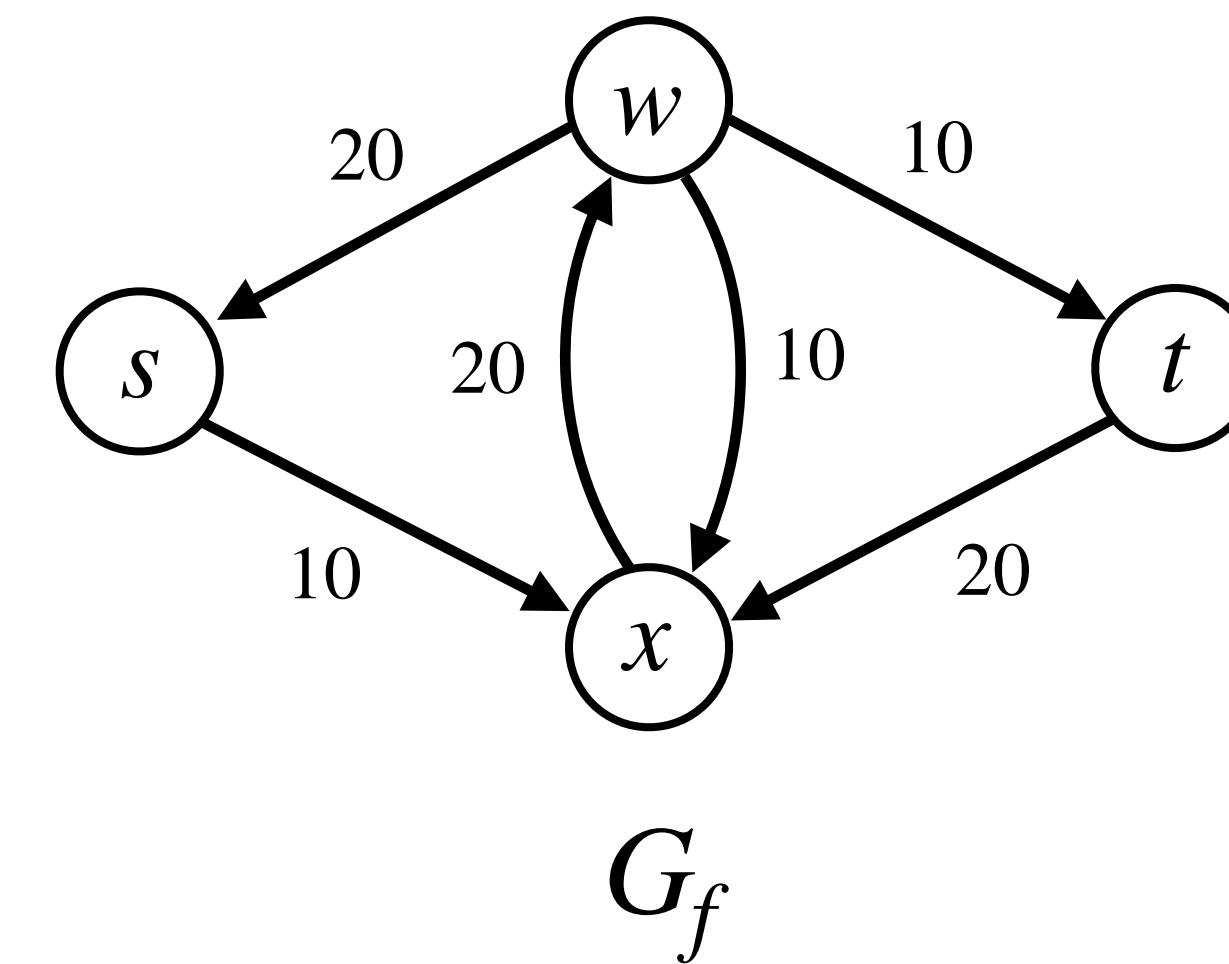
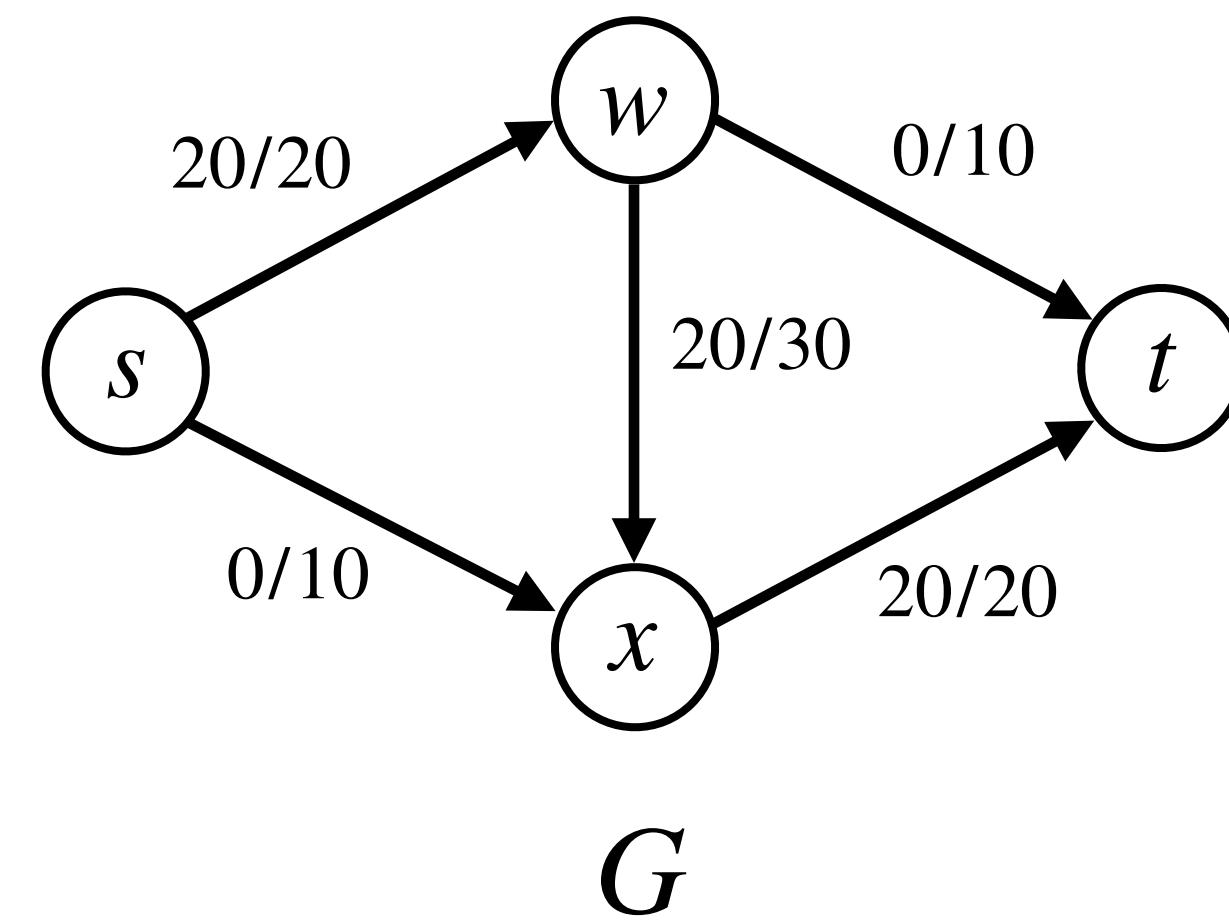


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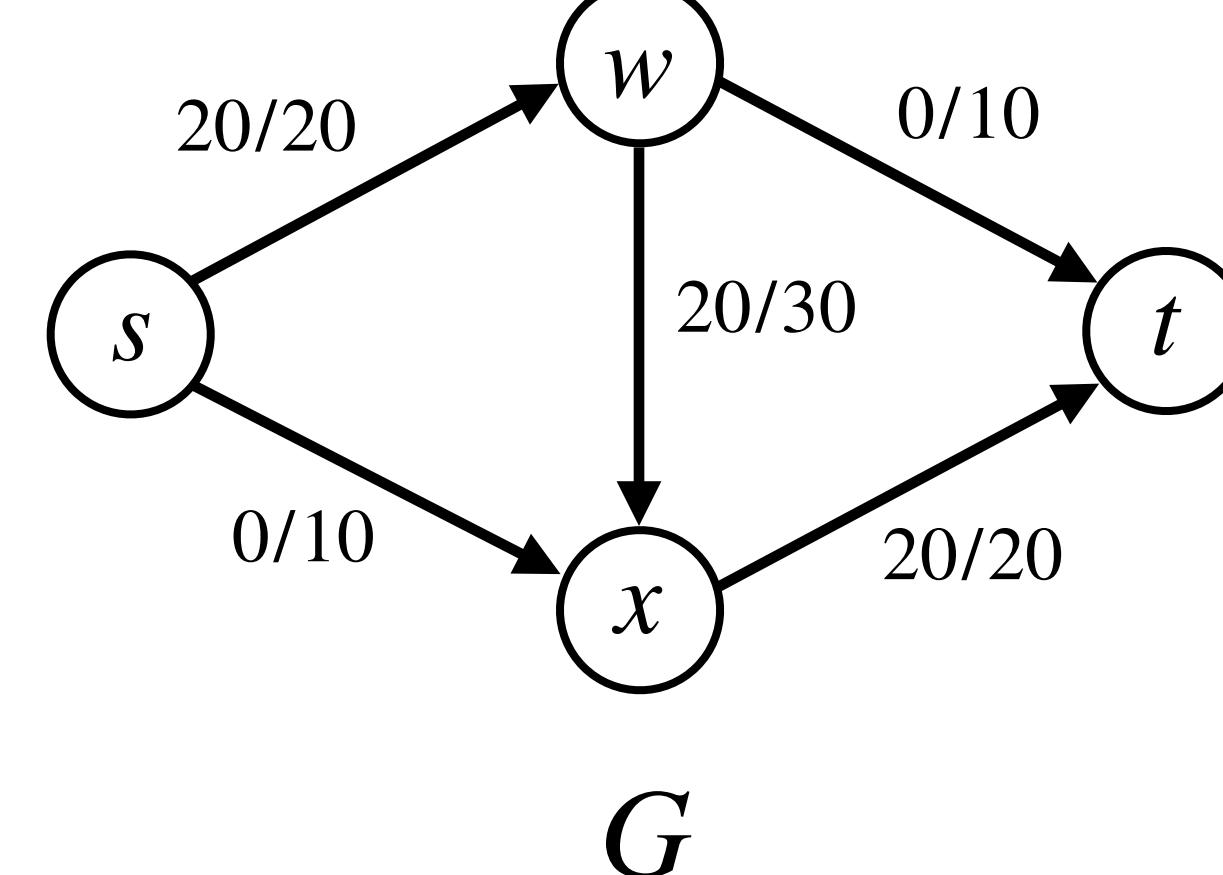


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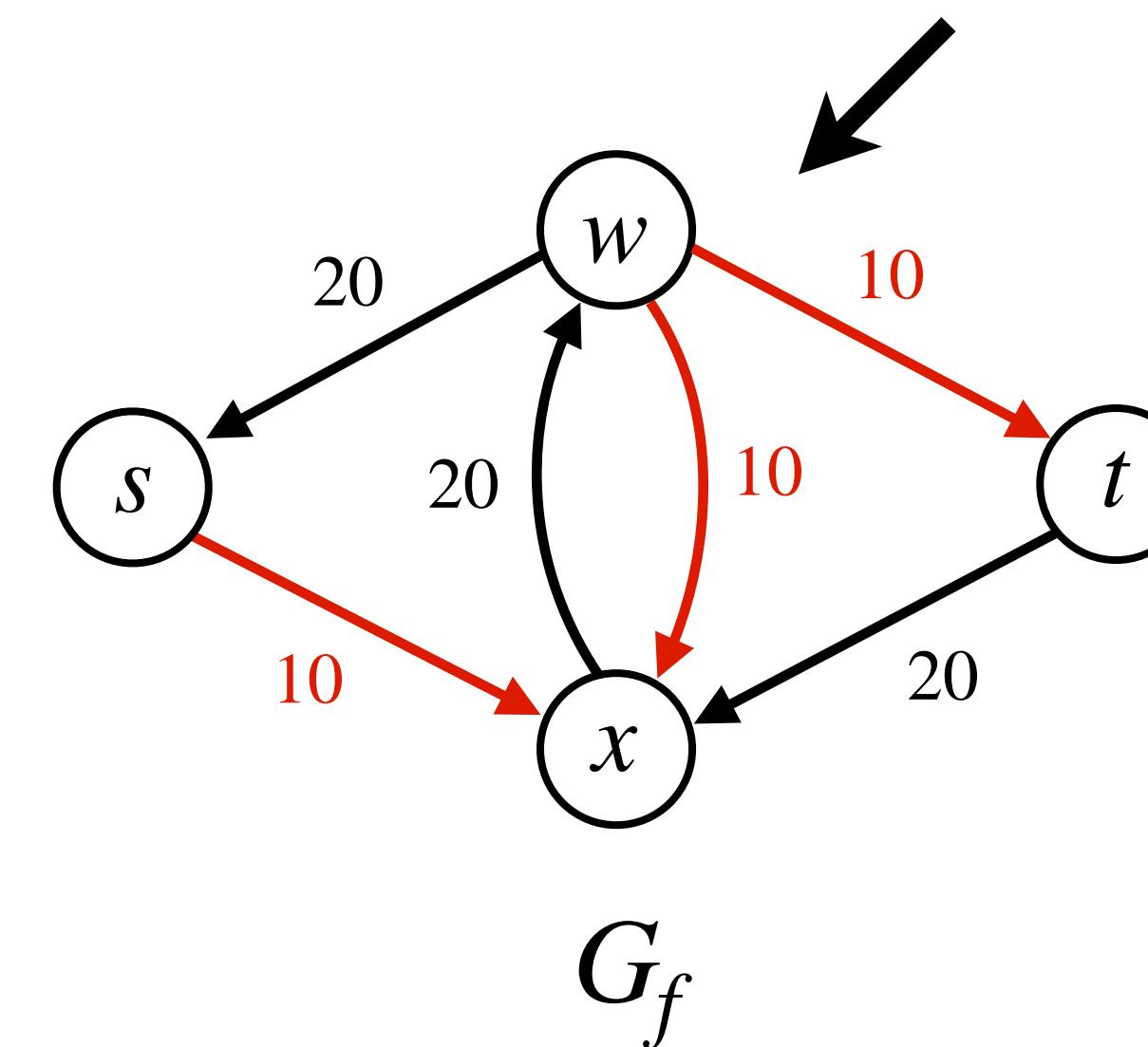
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Example:



Forward edges to use the remaining capacity

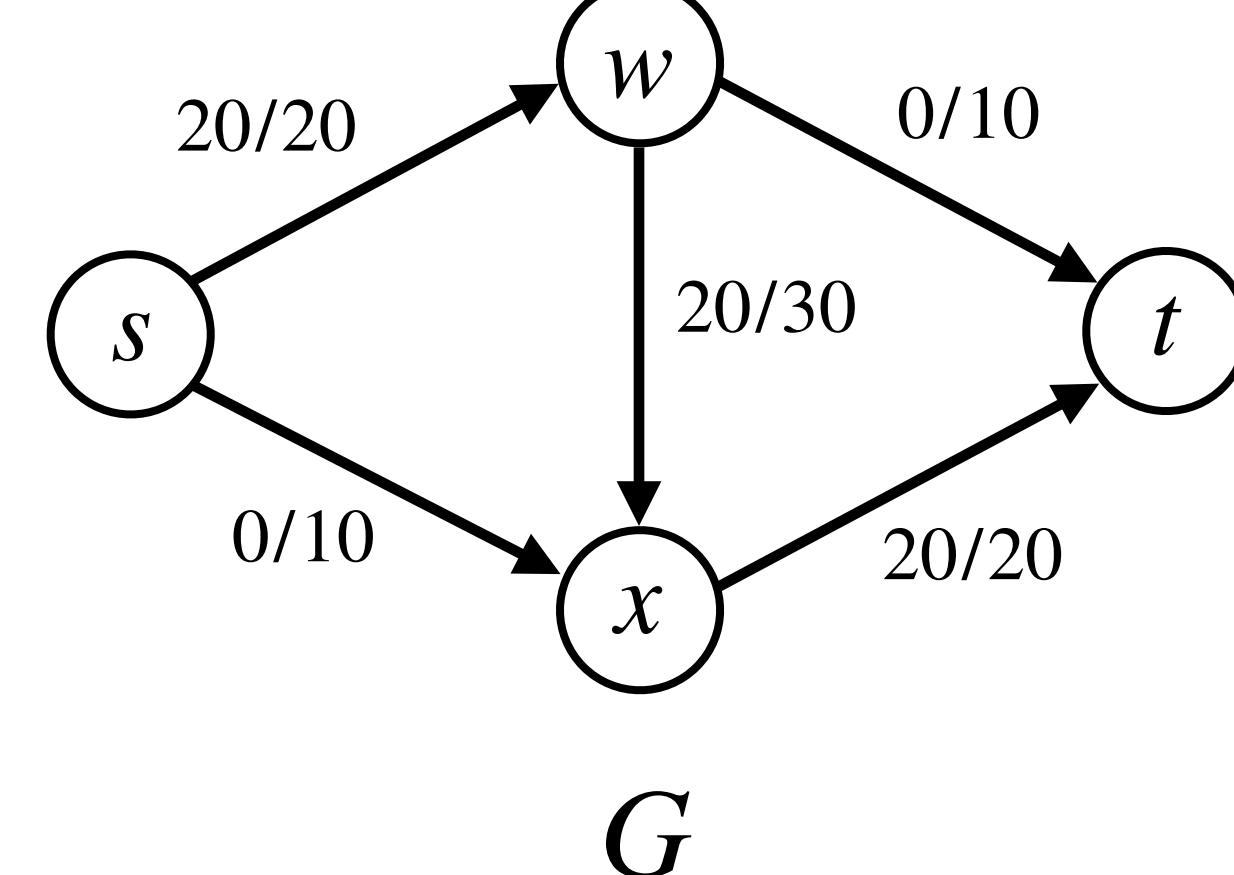


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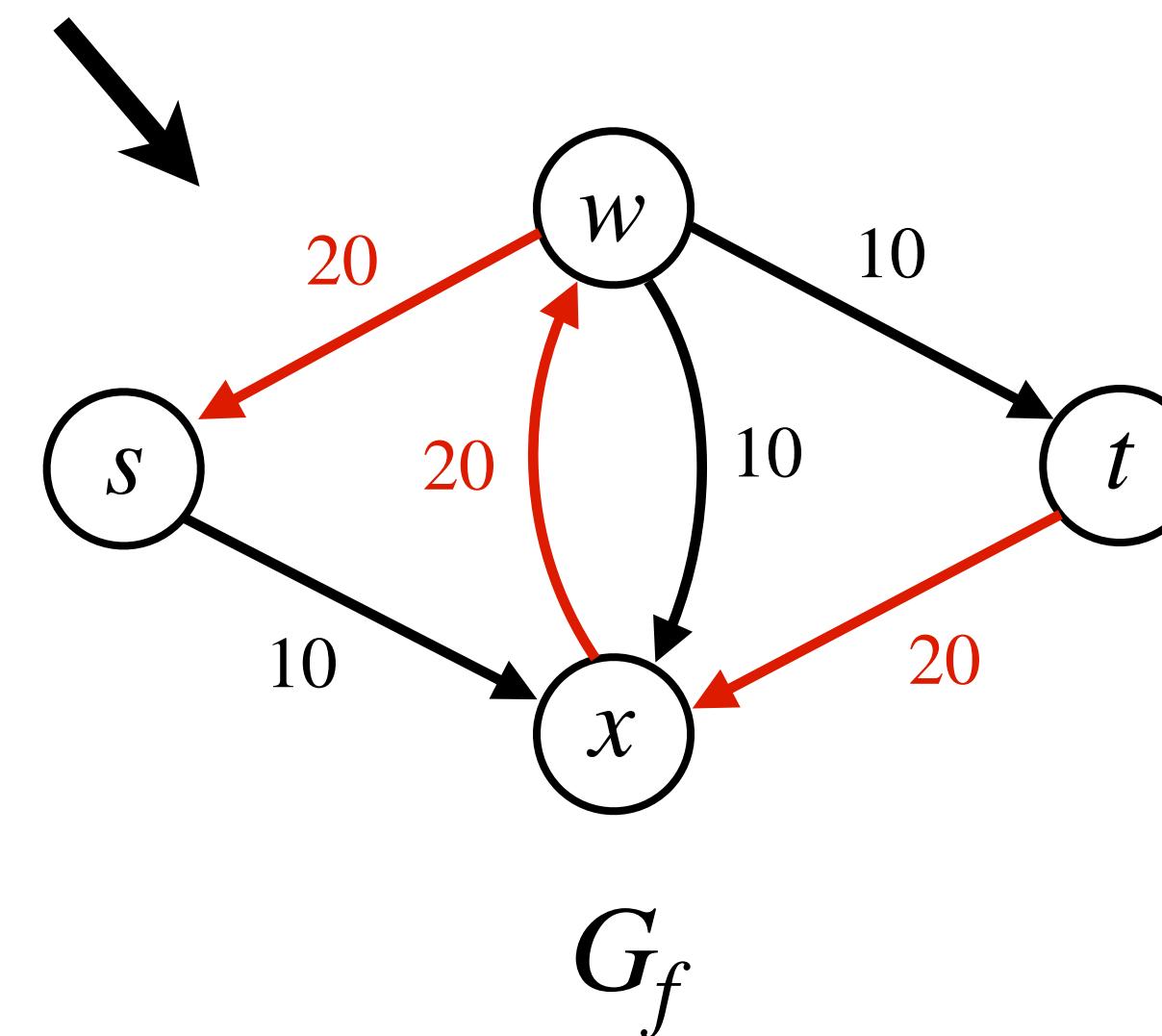
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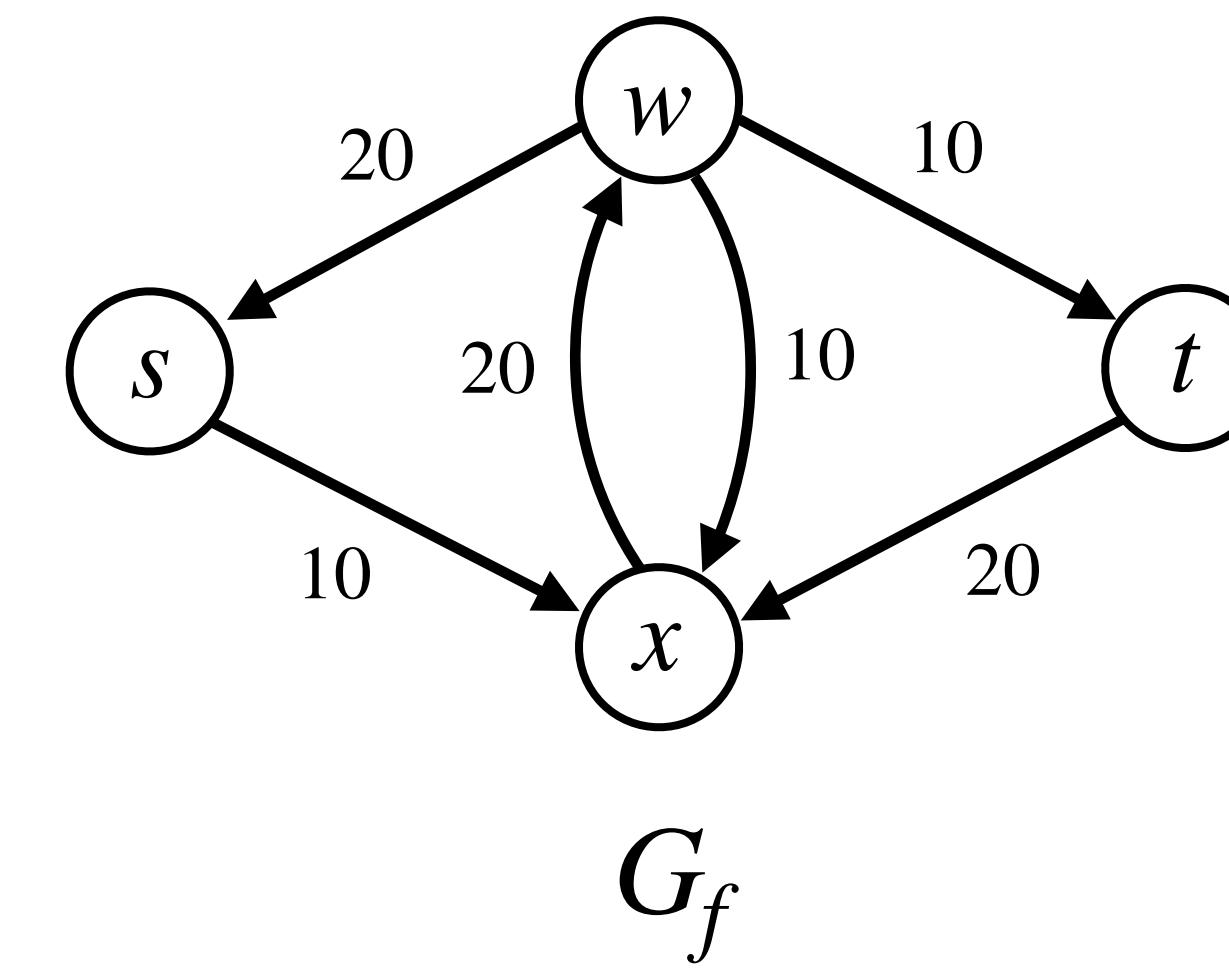
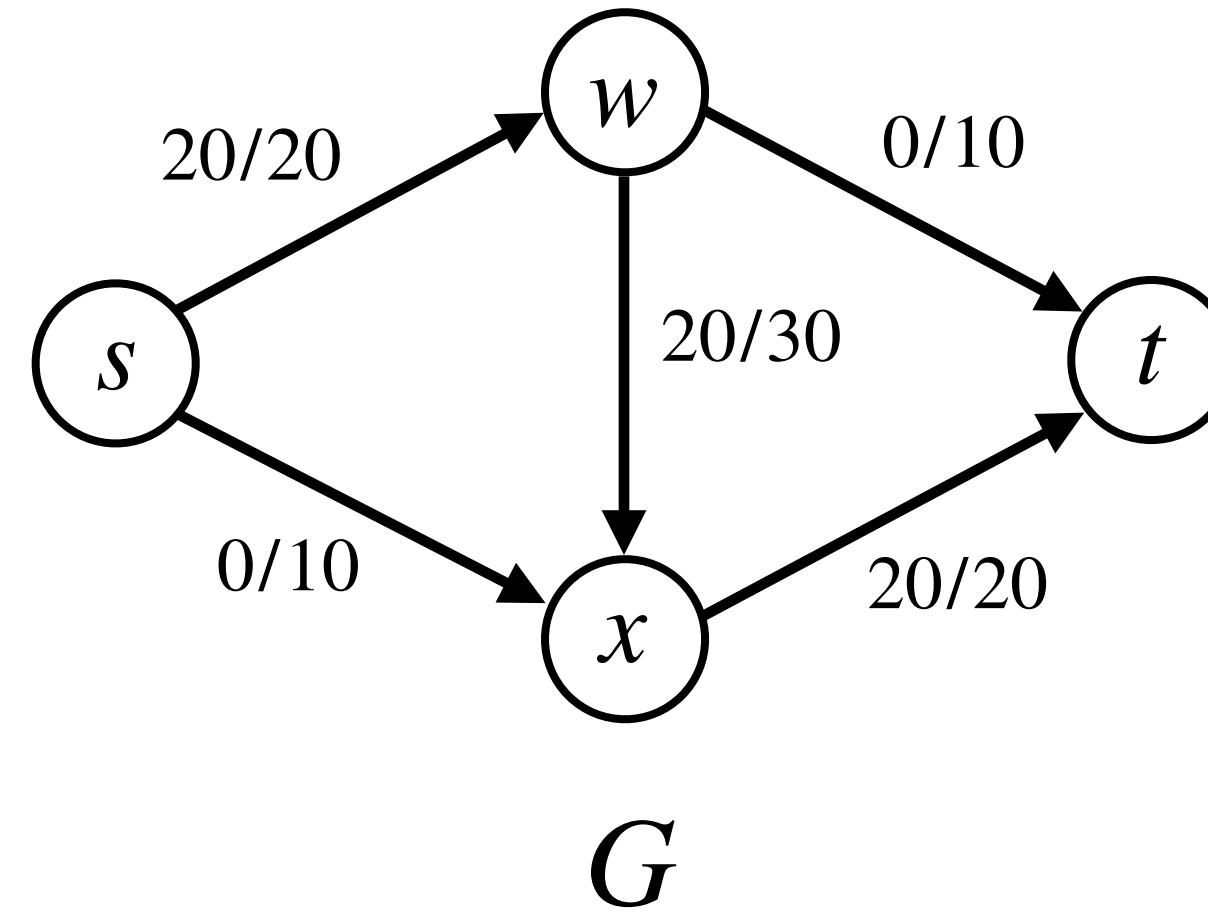


Backward edges to decrease the flow on some edges



Augmenting Flows via Residual Networks

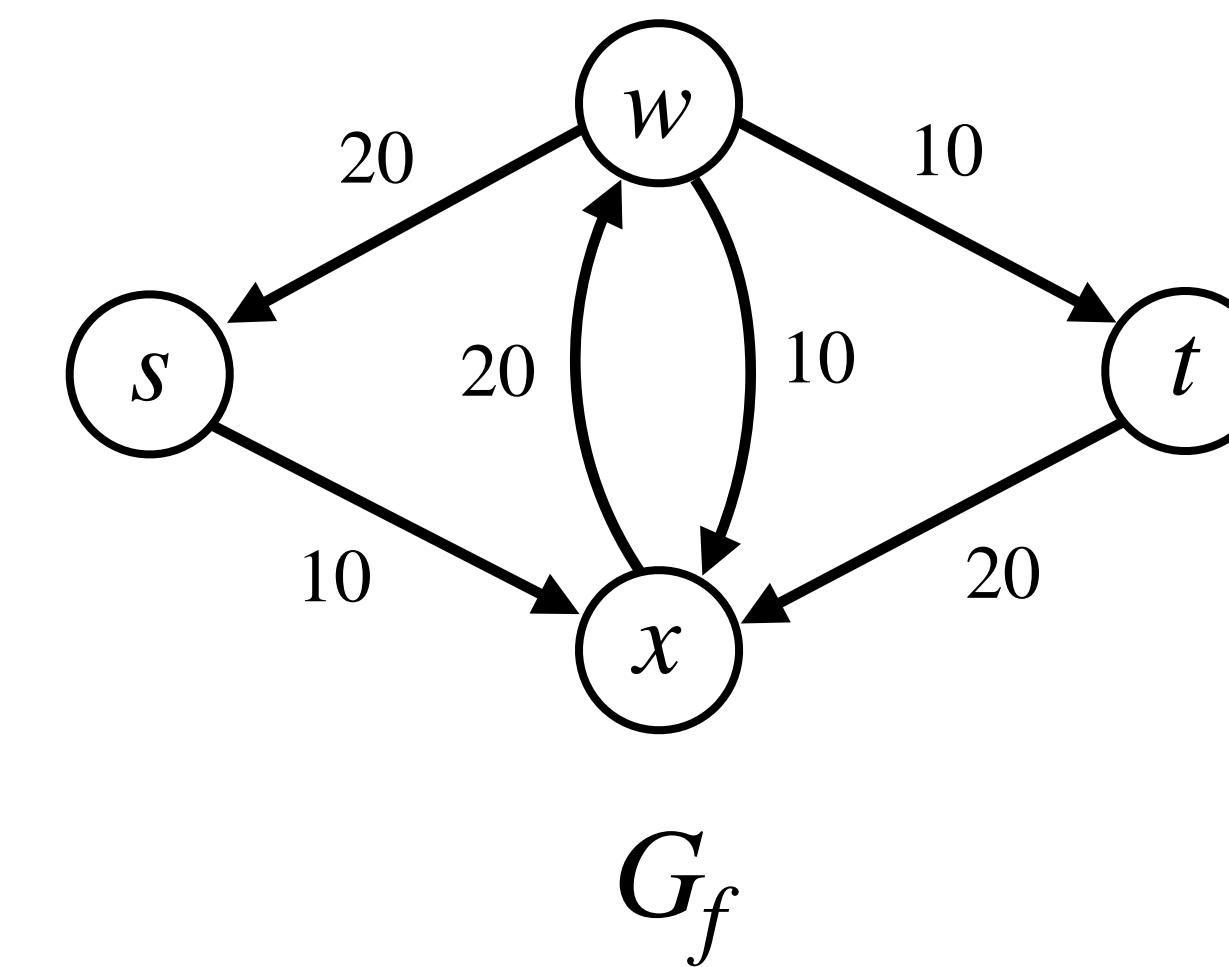
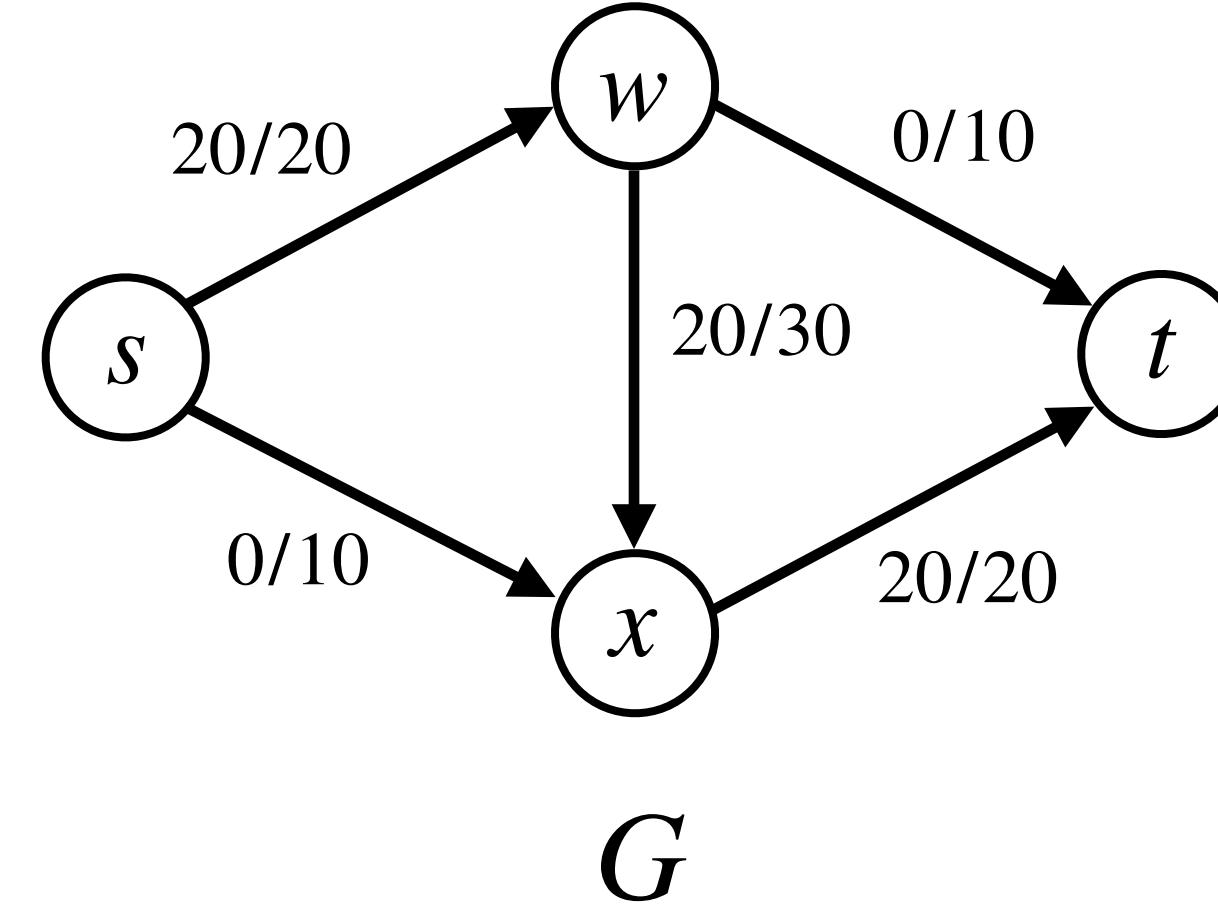
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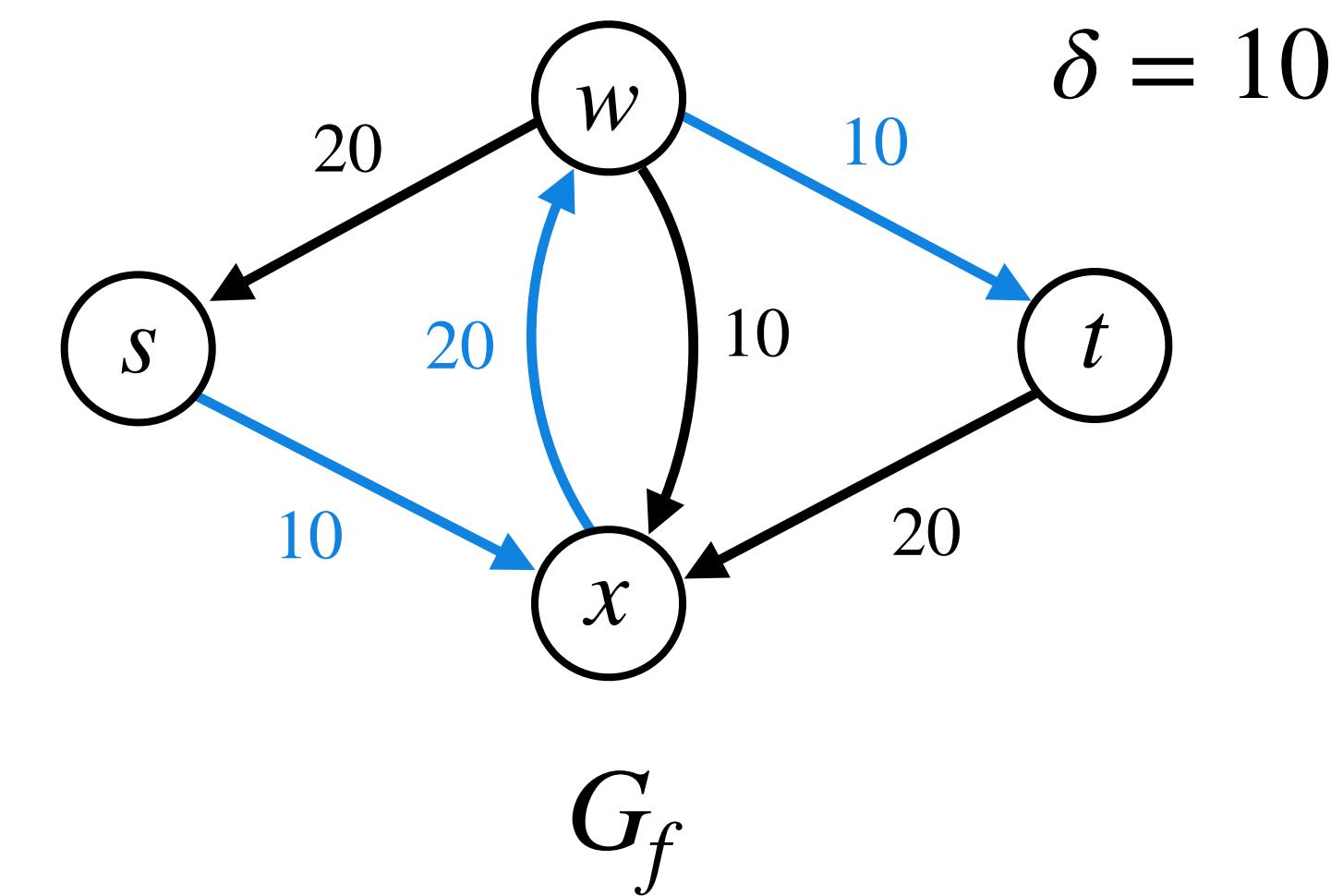
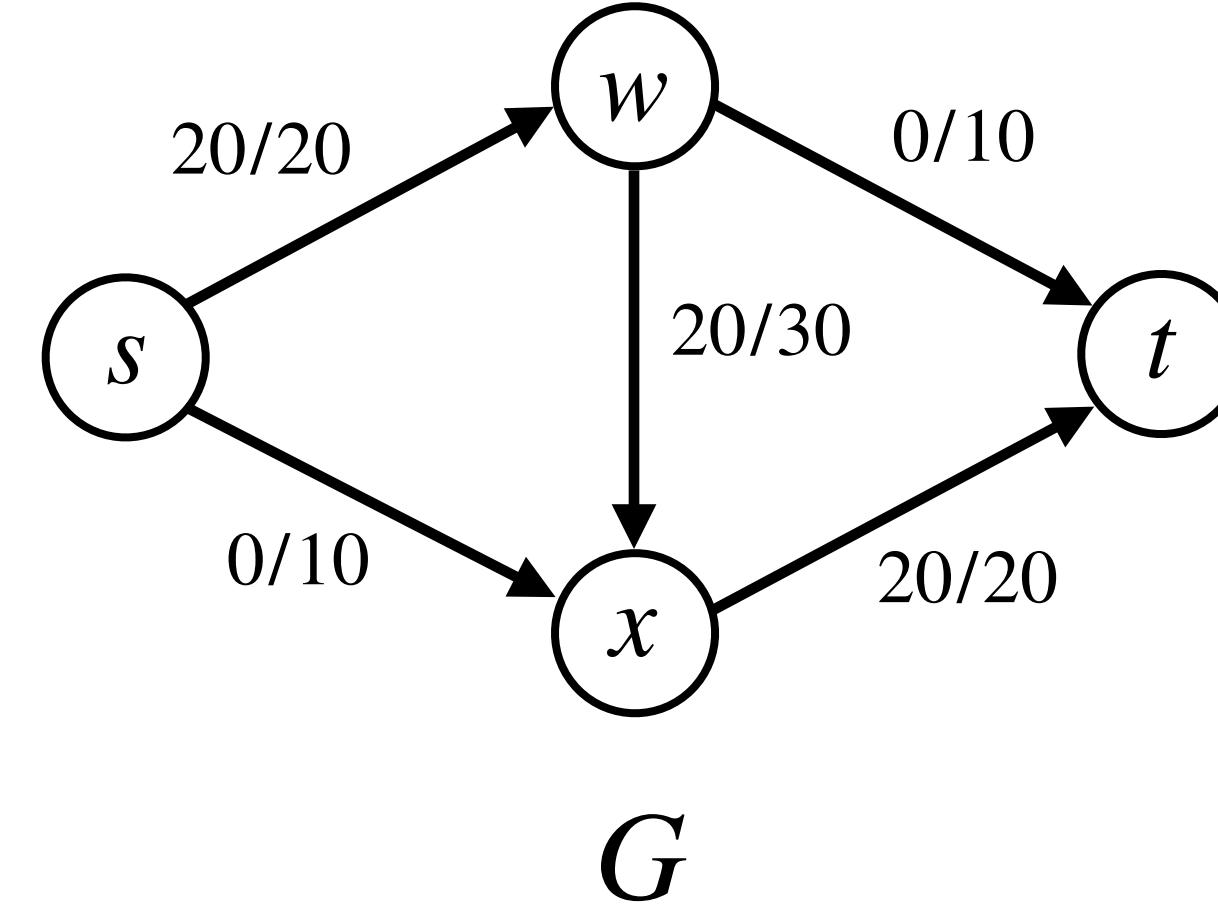
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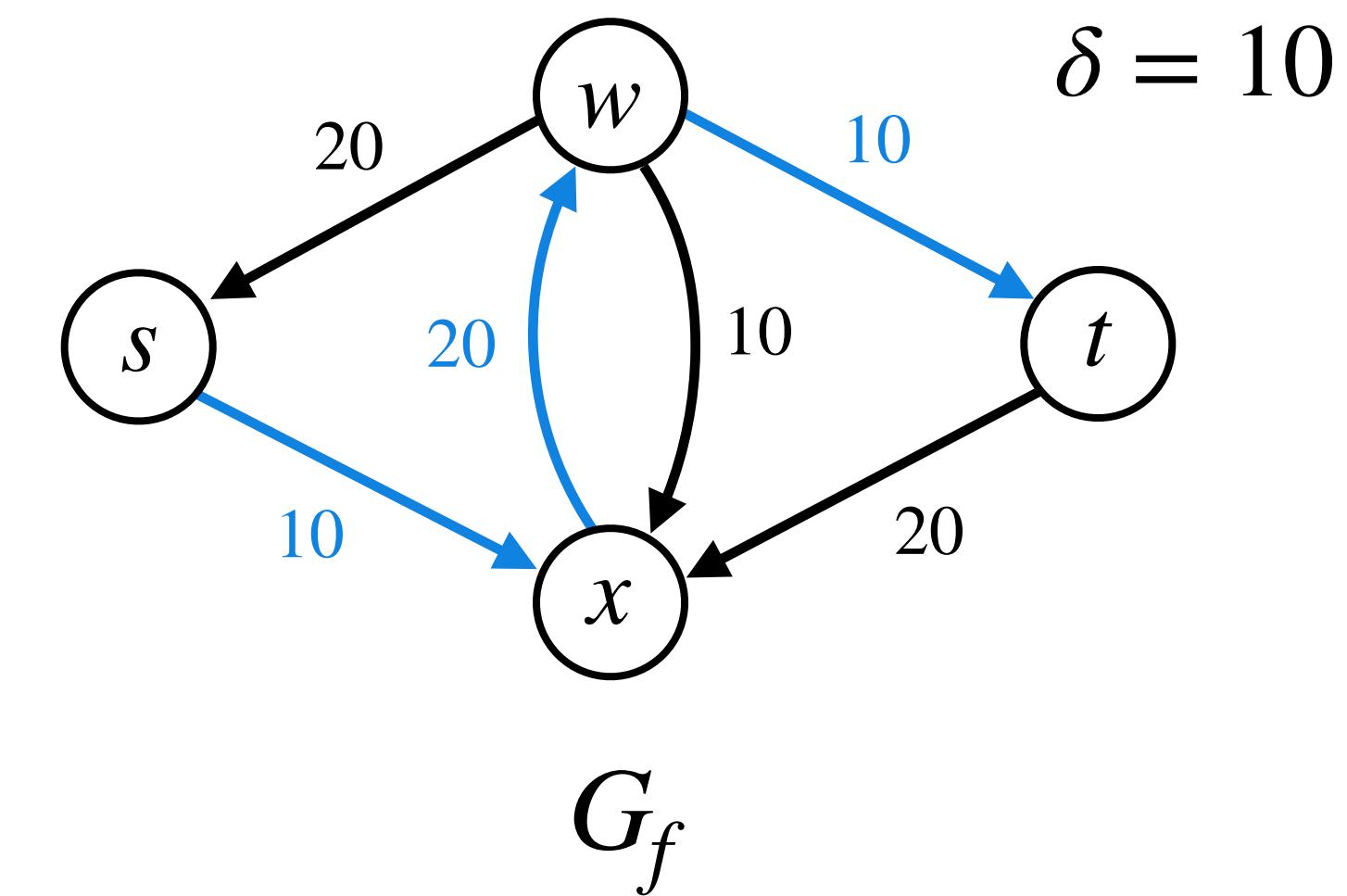
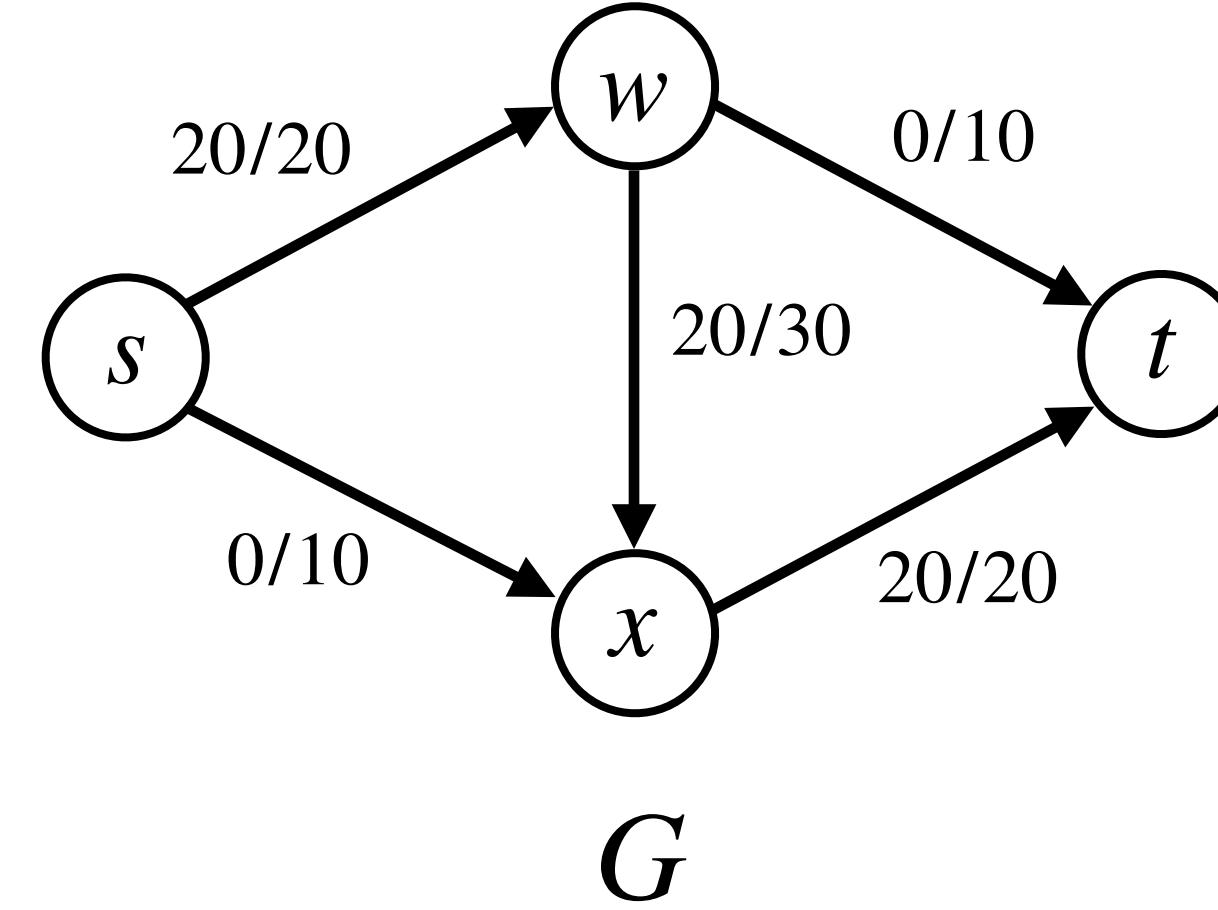
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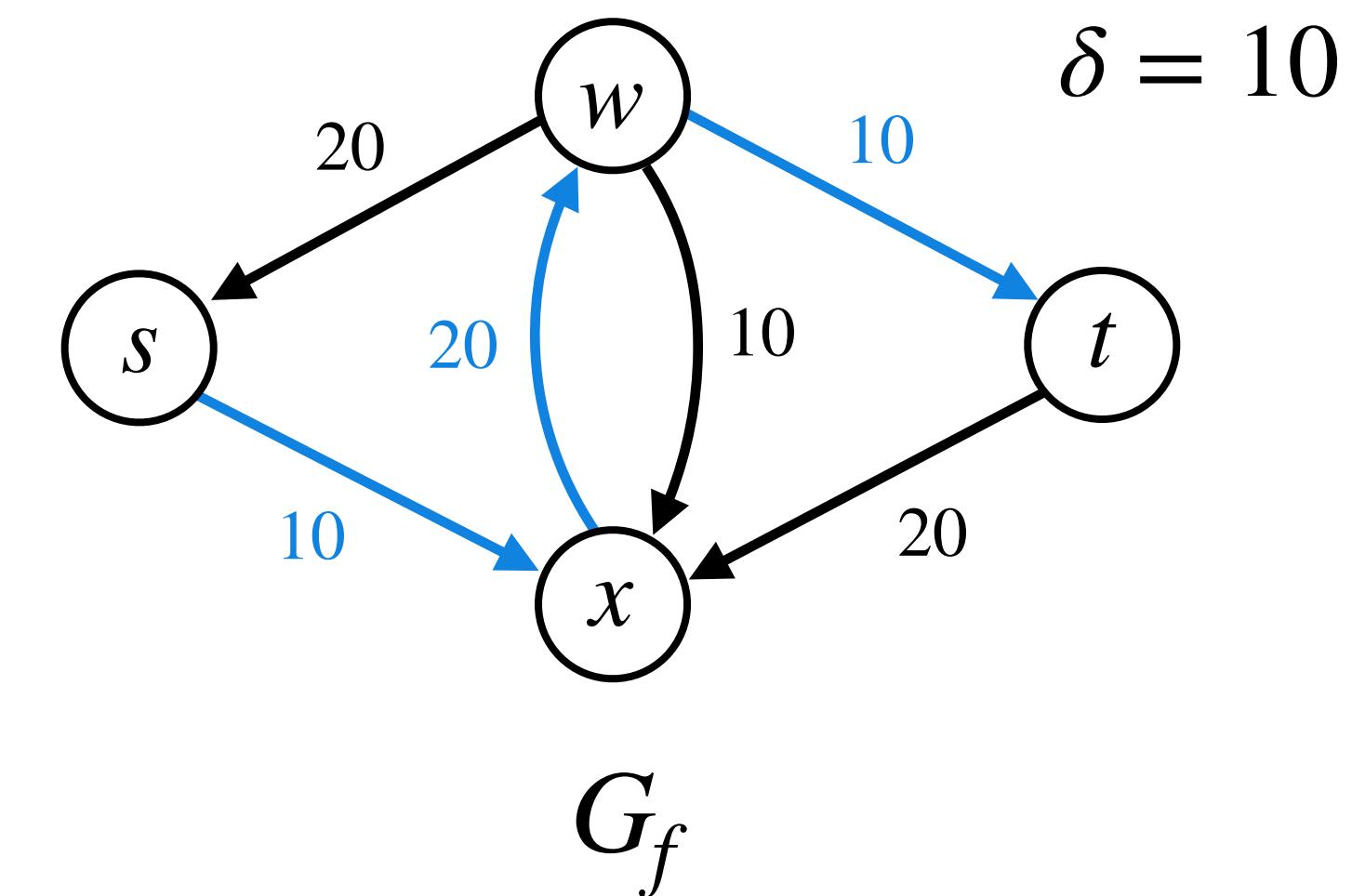
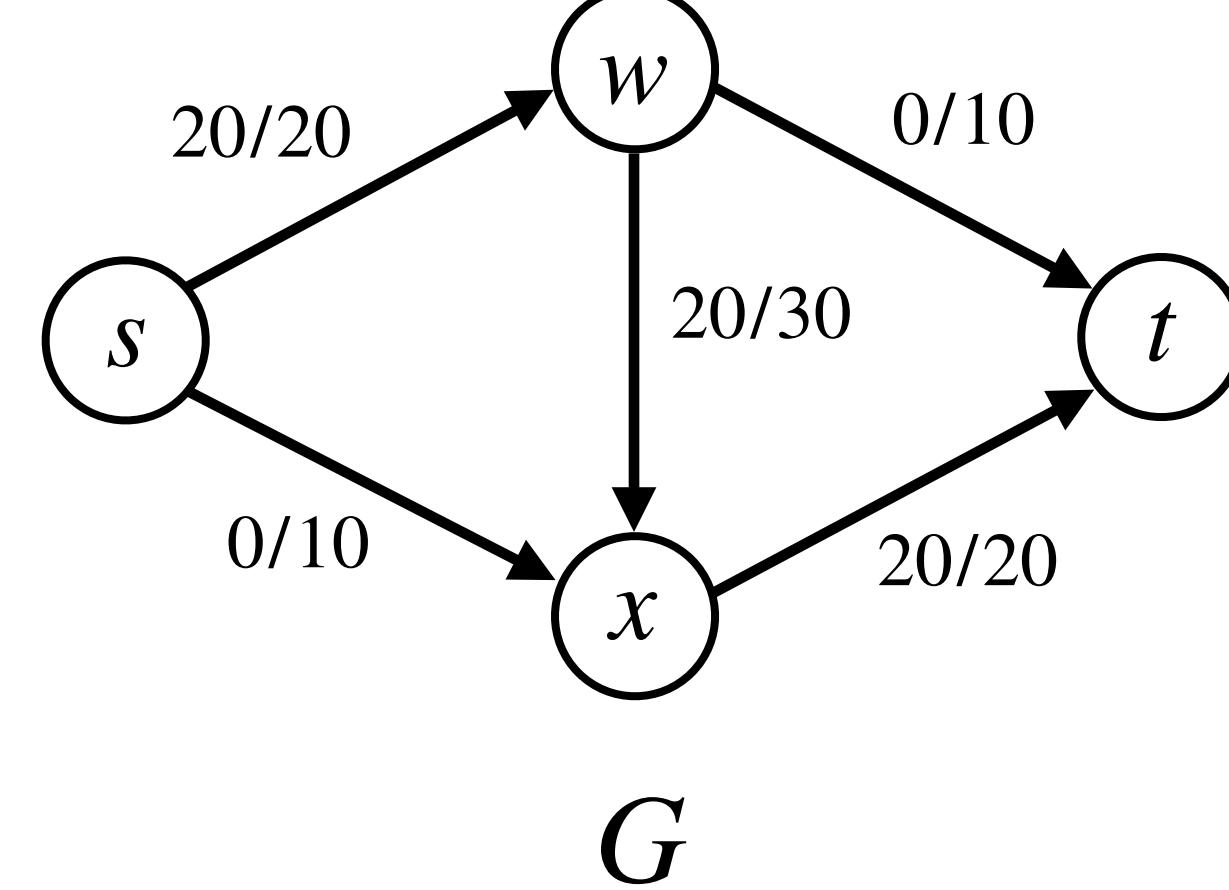
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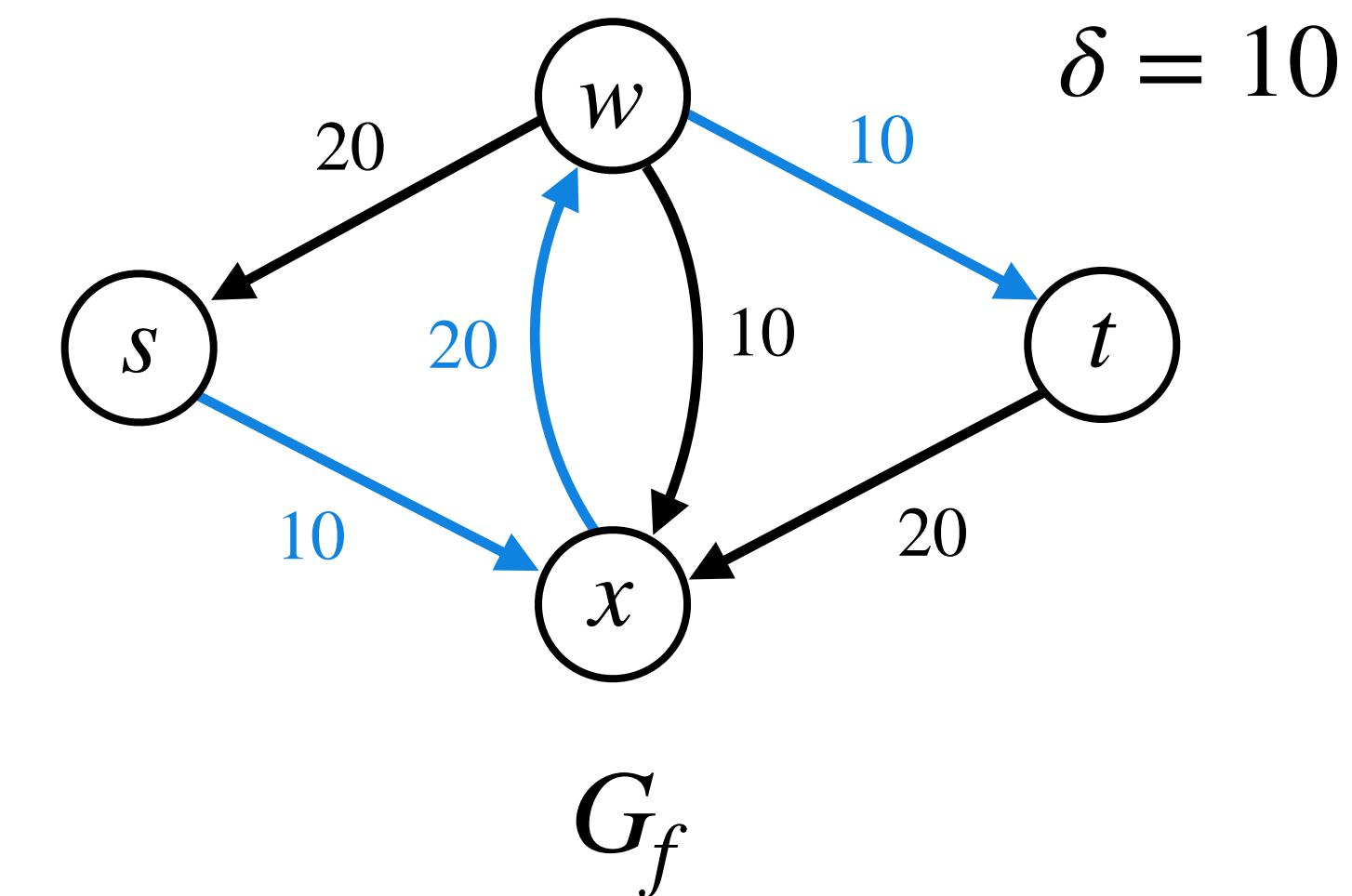
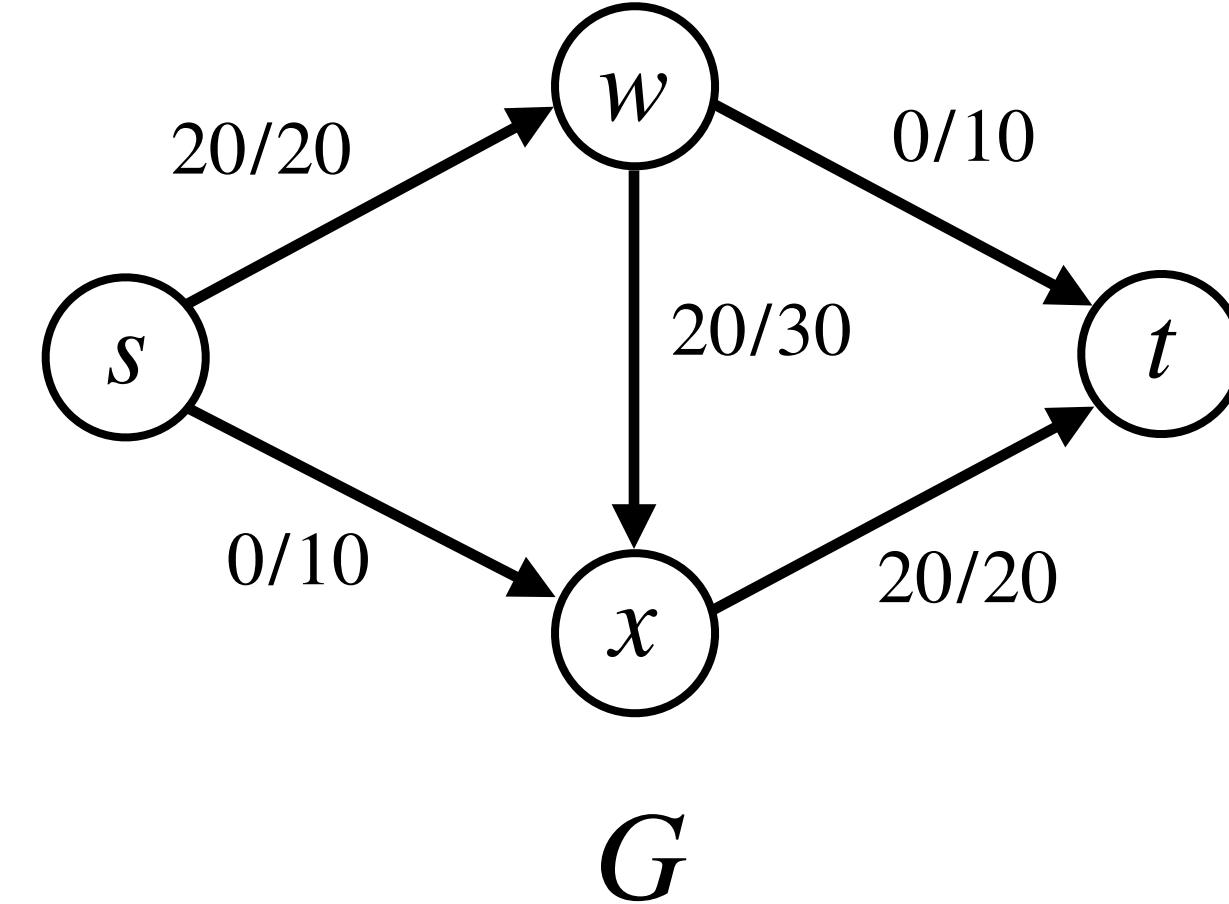
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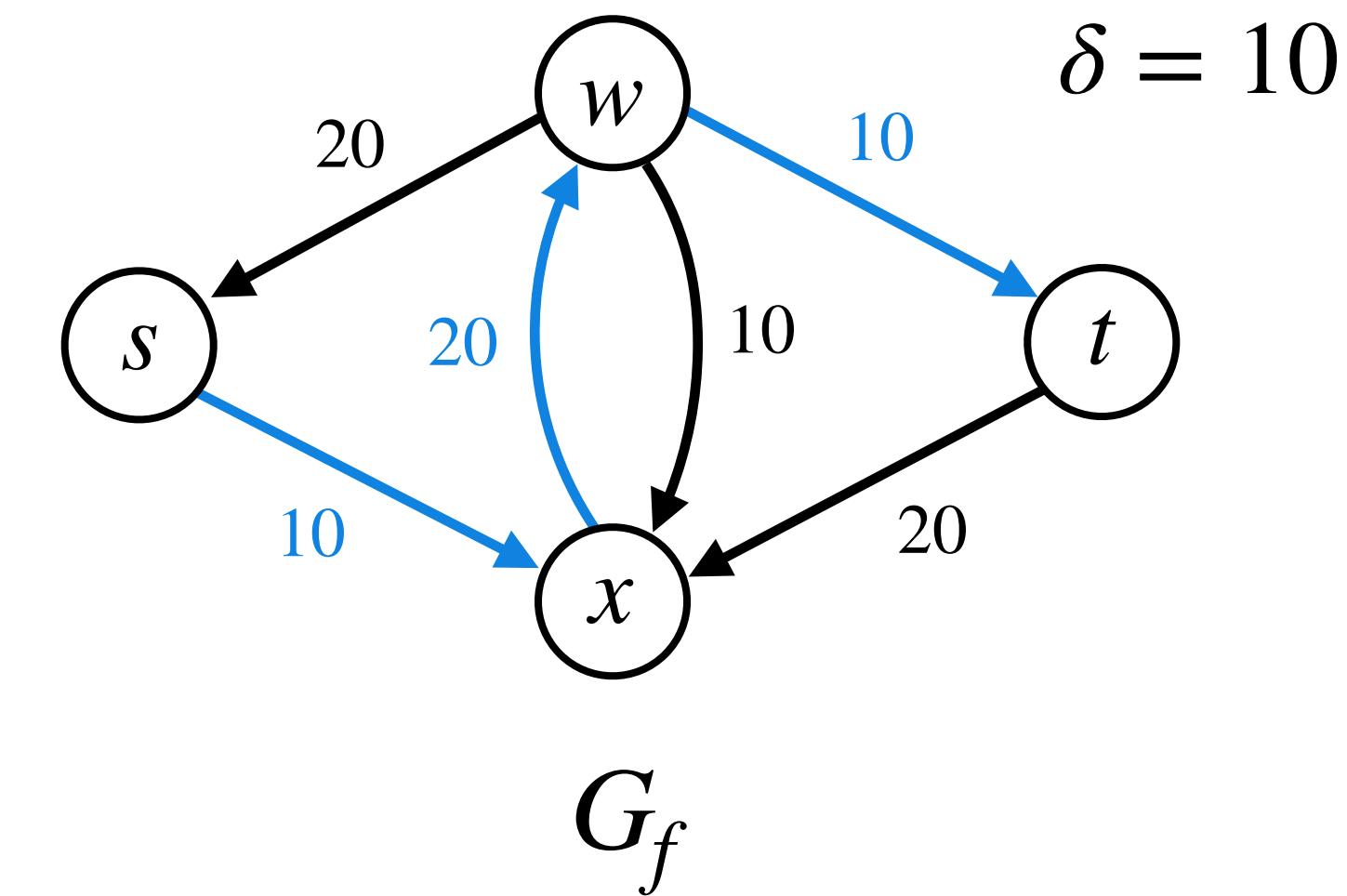
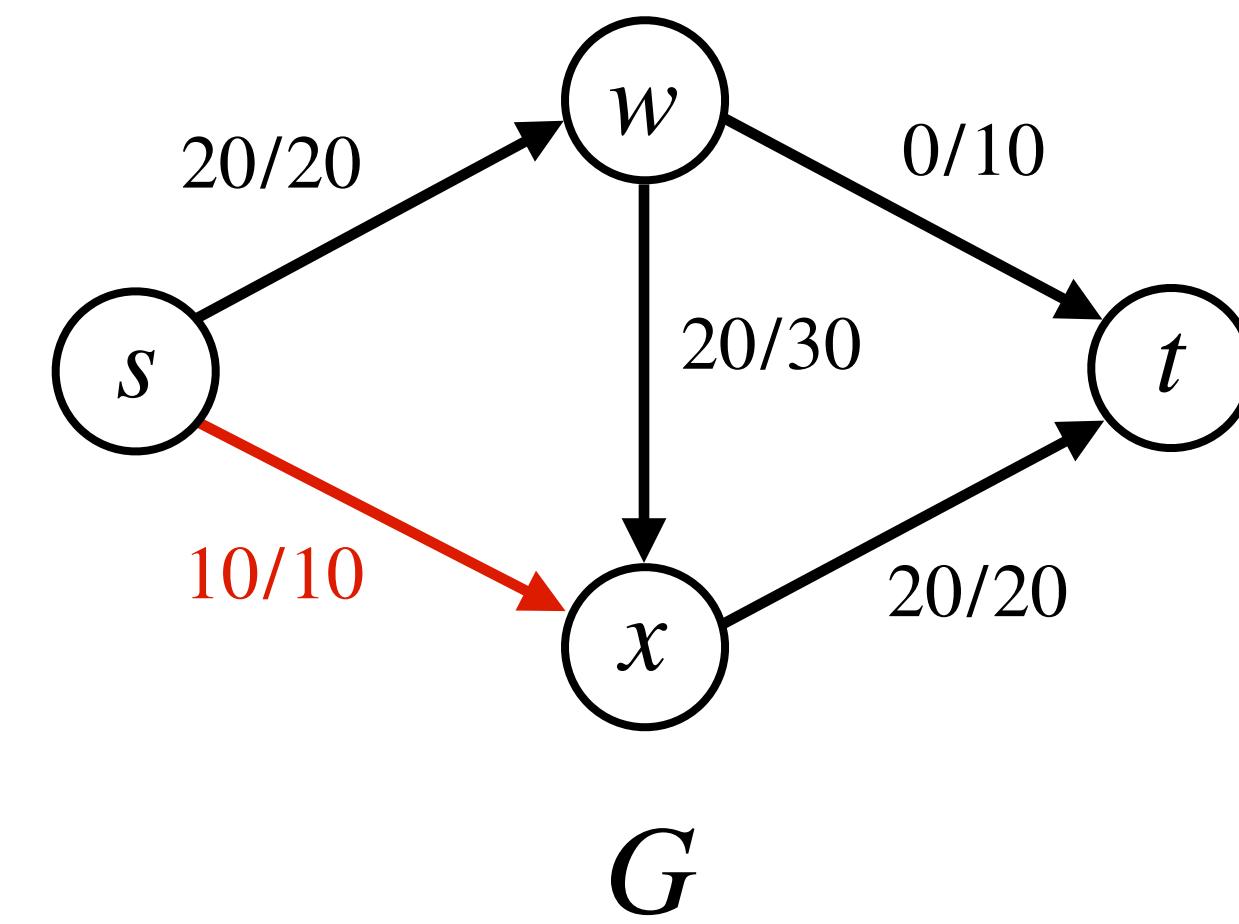
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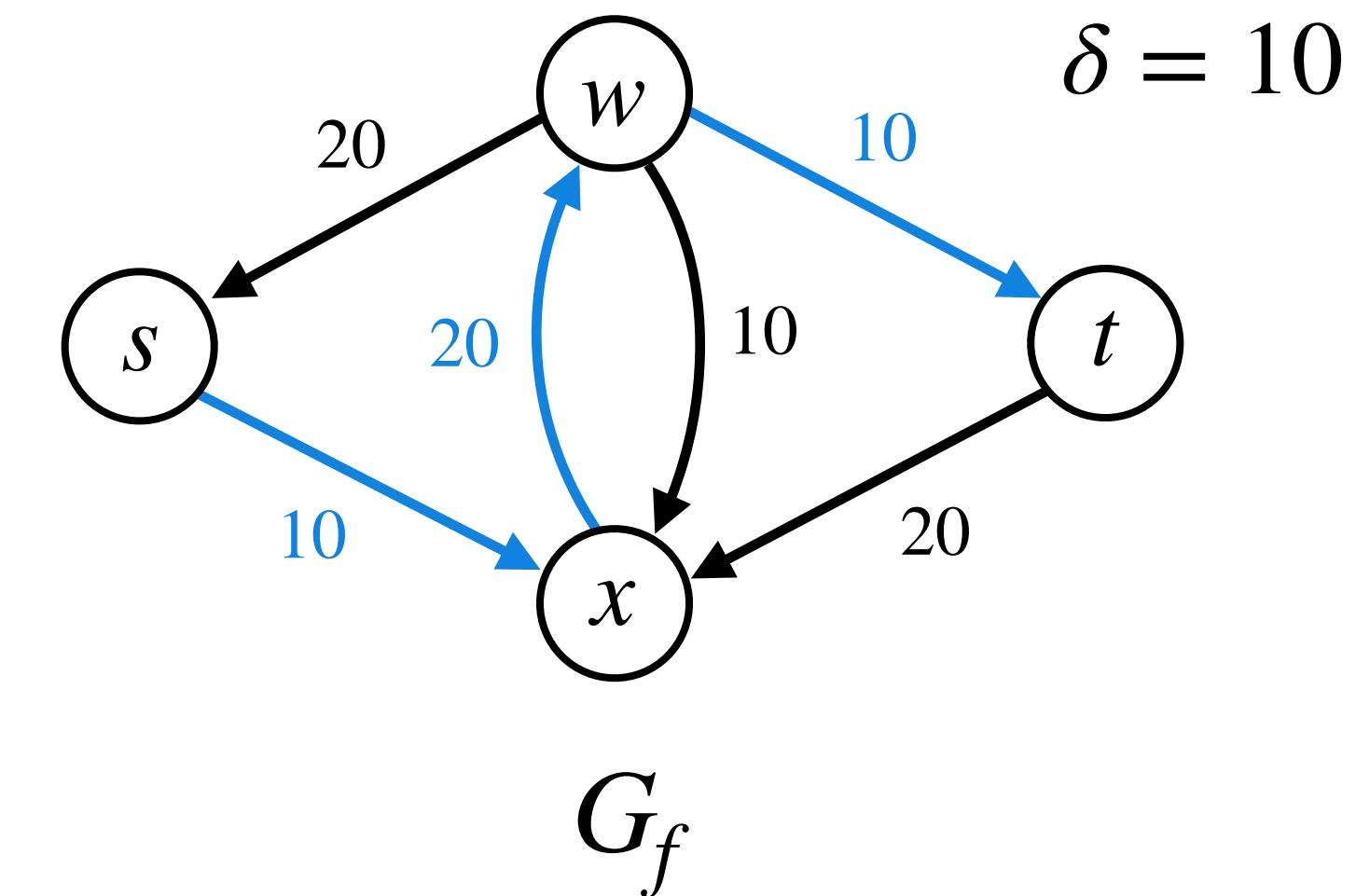
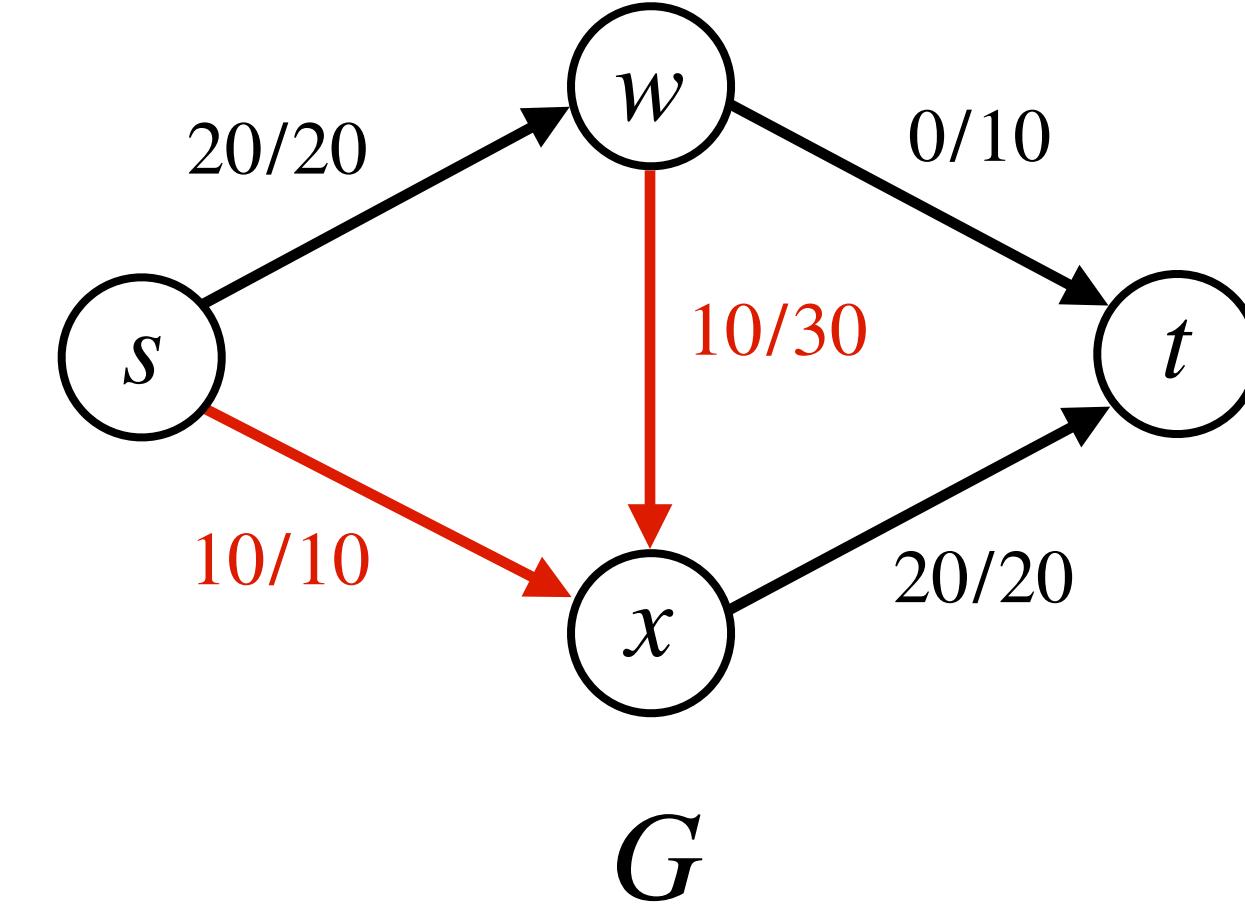
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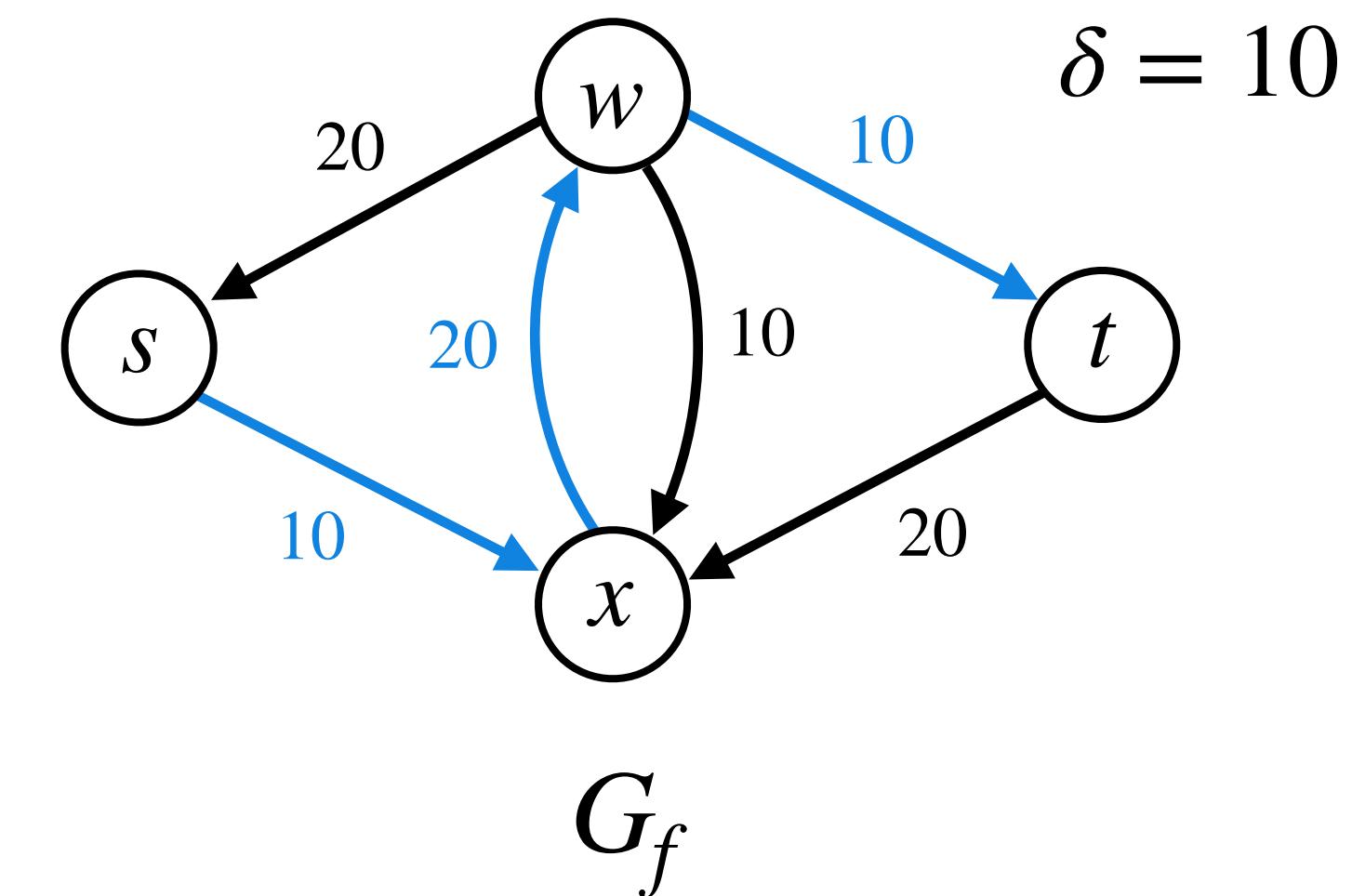
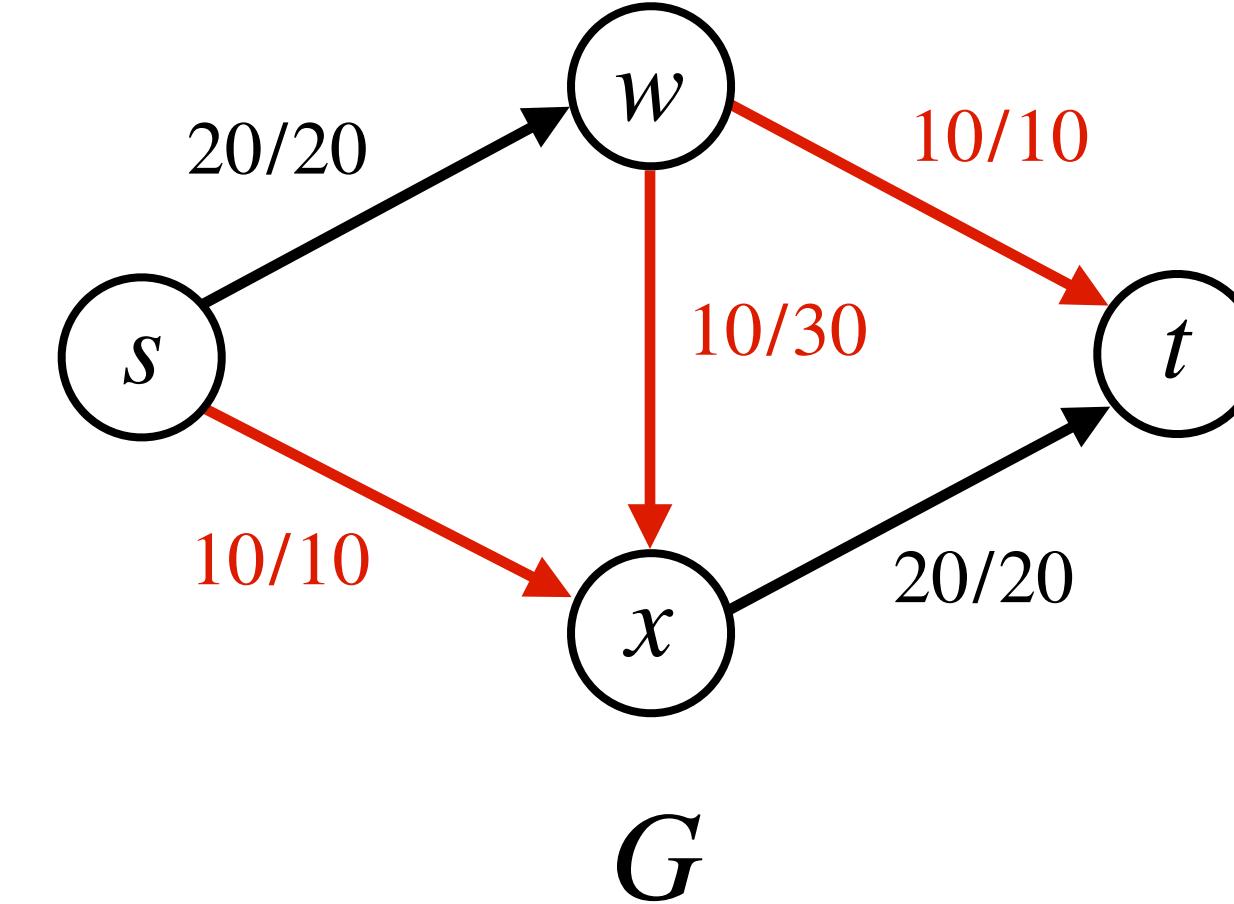
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- Find $s \rightsquigarrow t$ path P in the residual network G_f and its **bottleneck capacity δ** .
- For every $(u, v) \in P$:
 - If $(u, v) \in E(G)$, add δ flow to (u, v) in f .
 - If $(v, u) \in E(G)$, subtract δ flow from (v, u) in f .

What about capacity, conservation constraints?

Example:

