

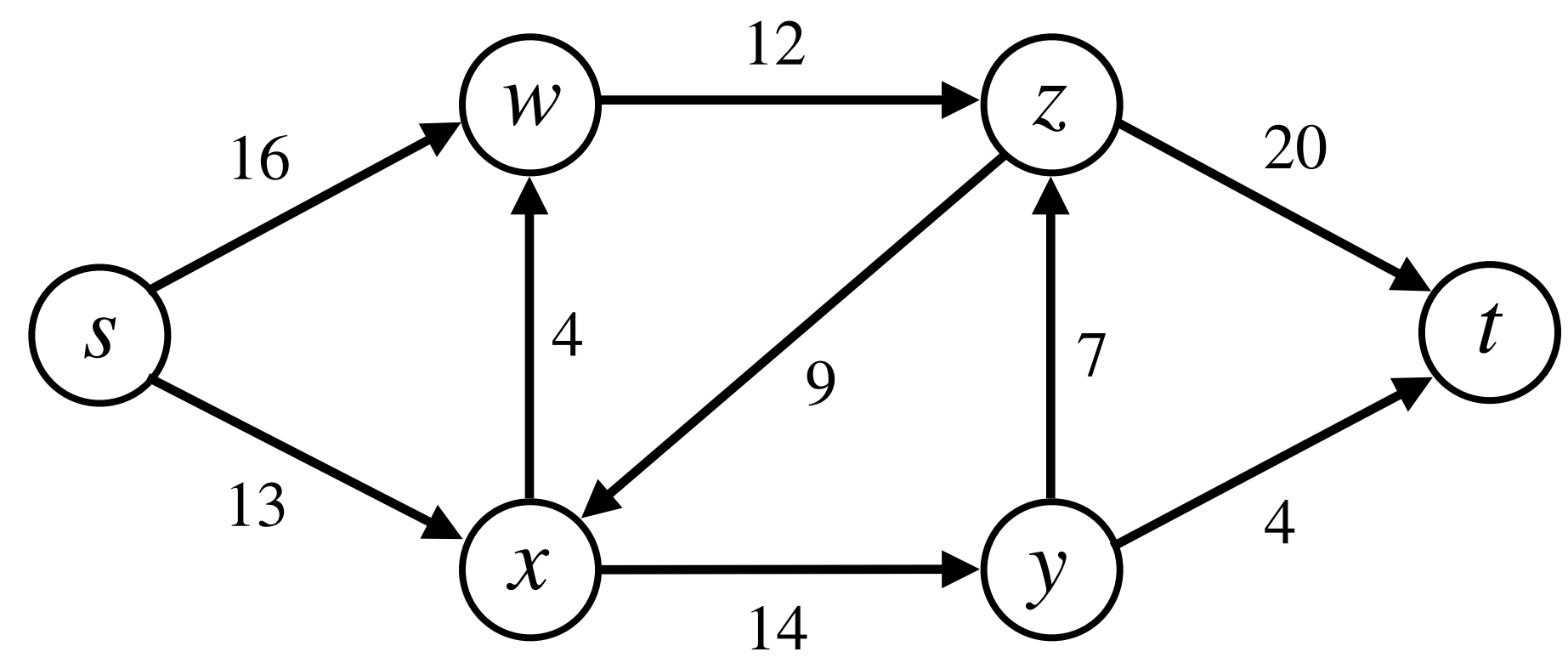
Lecture 17

Flow Networks, Ford-Fulkerson Method

Source: Introduction to Algorithms, CLRS and Kleinberg & Tardos

Flow Networks

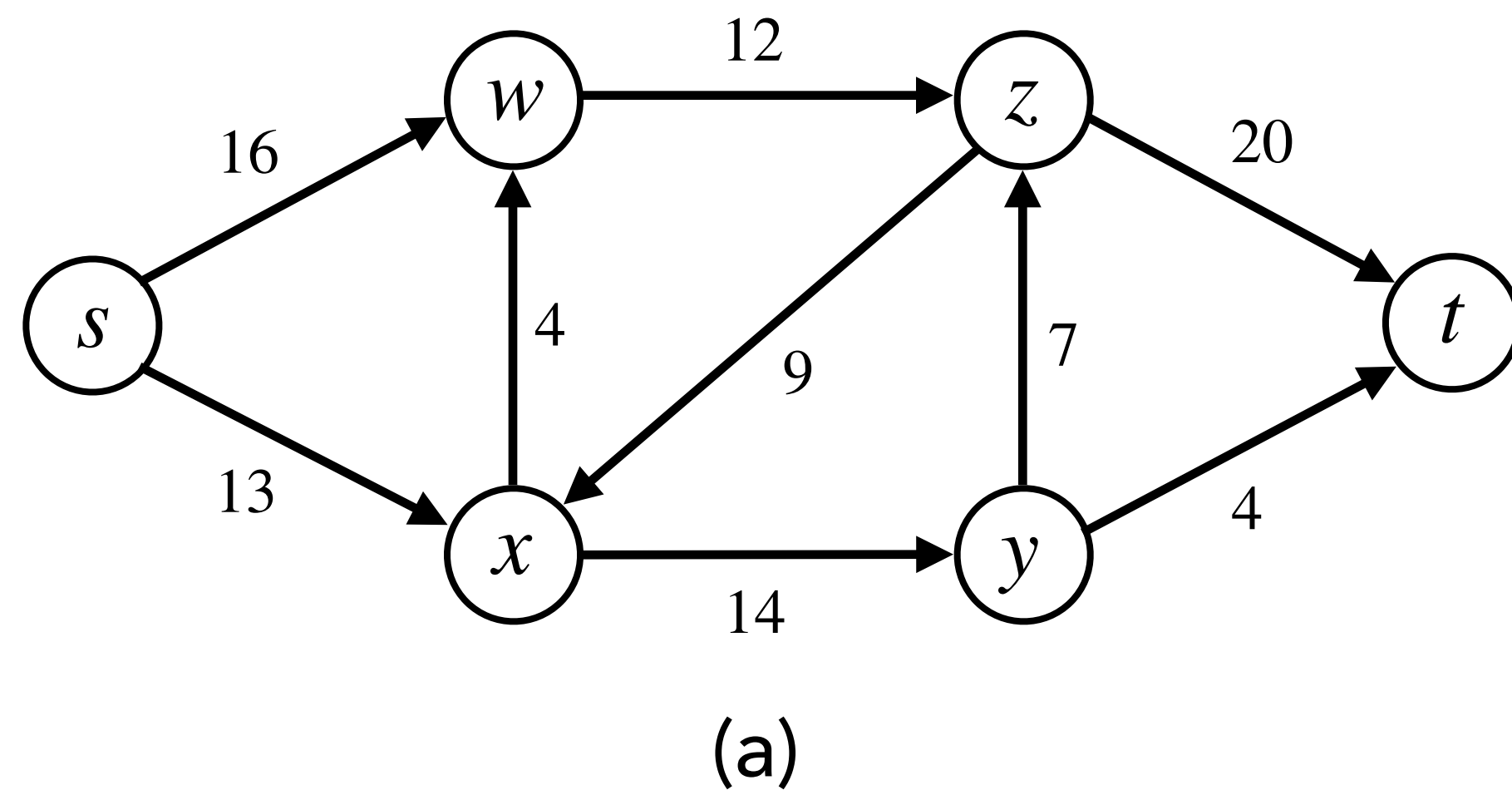
Flow Networks



(a)

Flow Networks

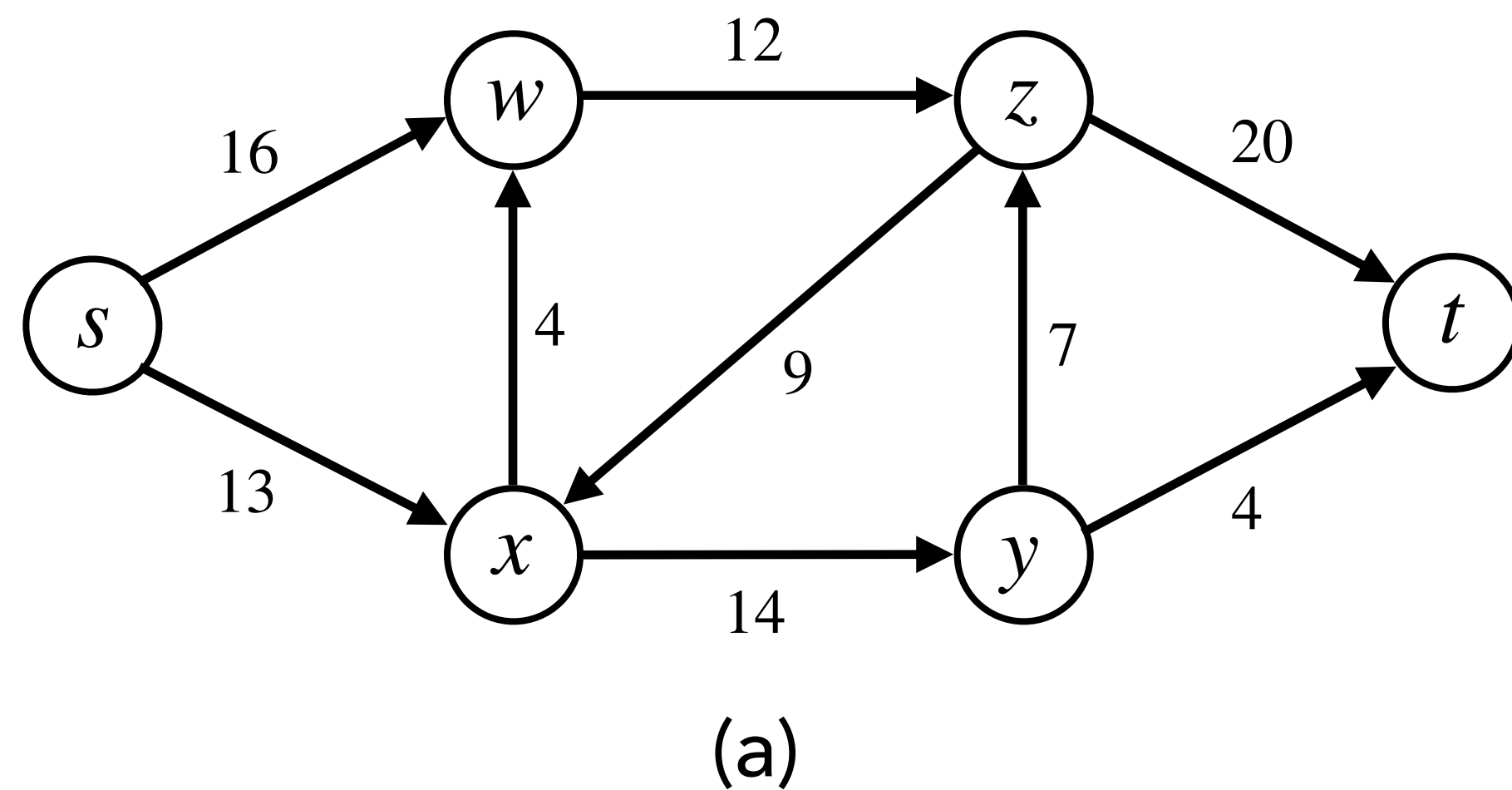
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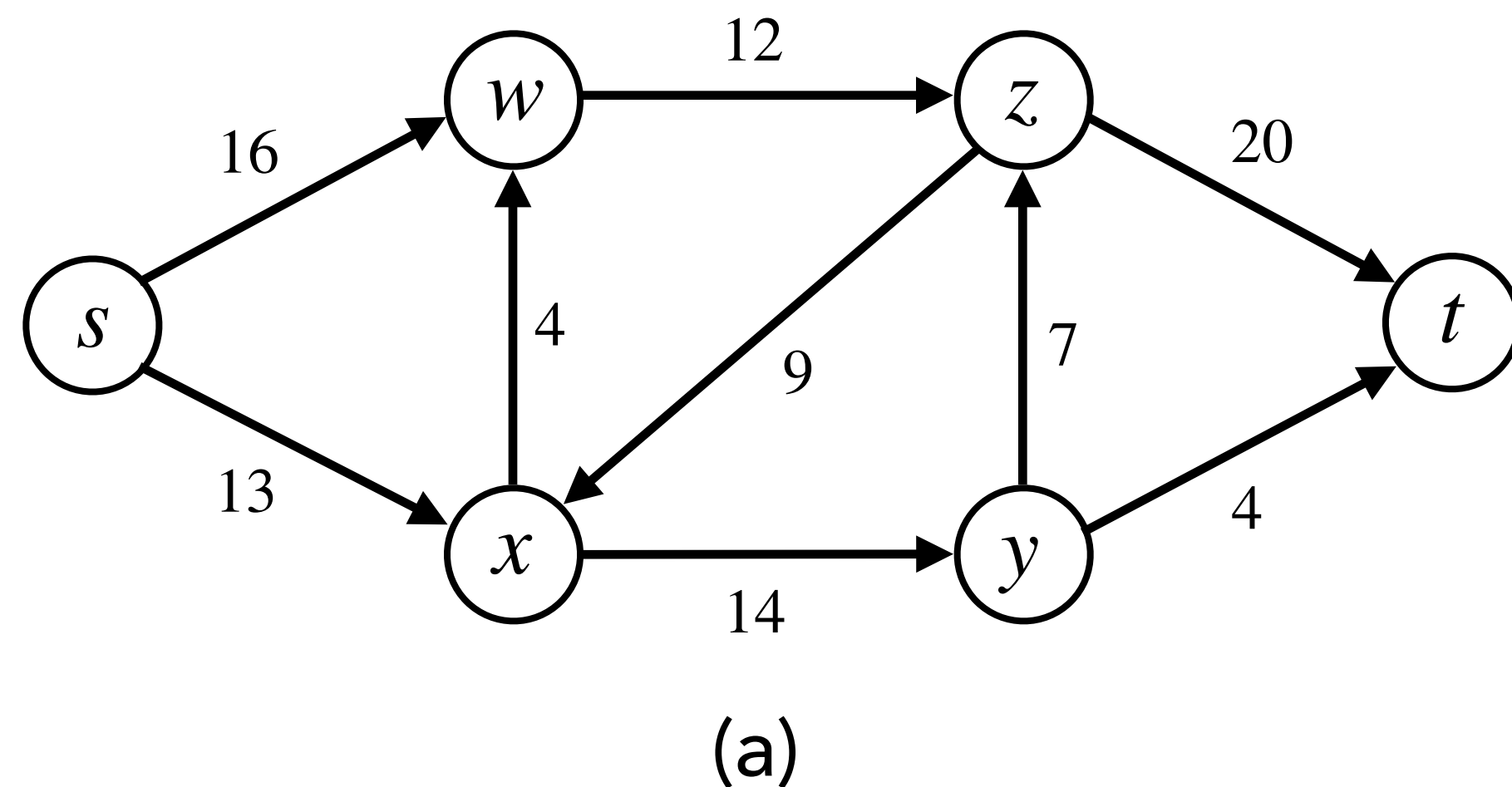
- Vertices represent **cities**. *s* & *t* are the **source** & **sink** cities.



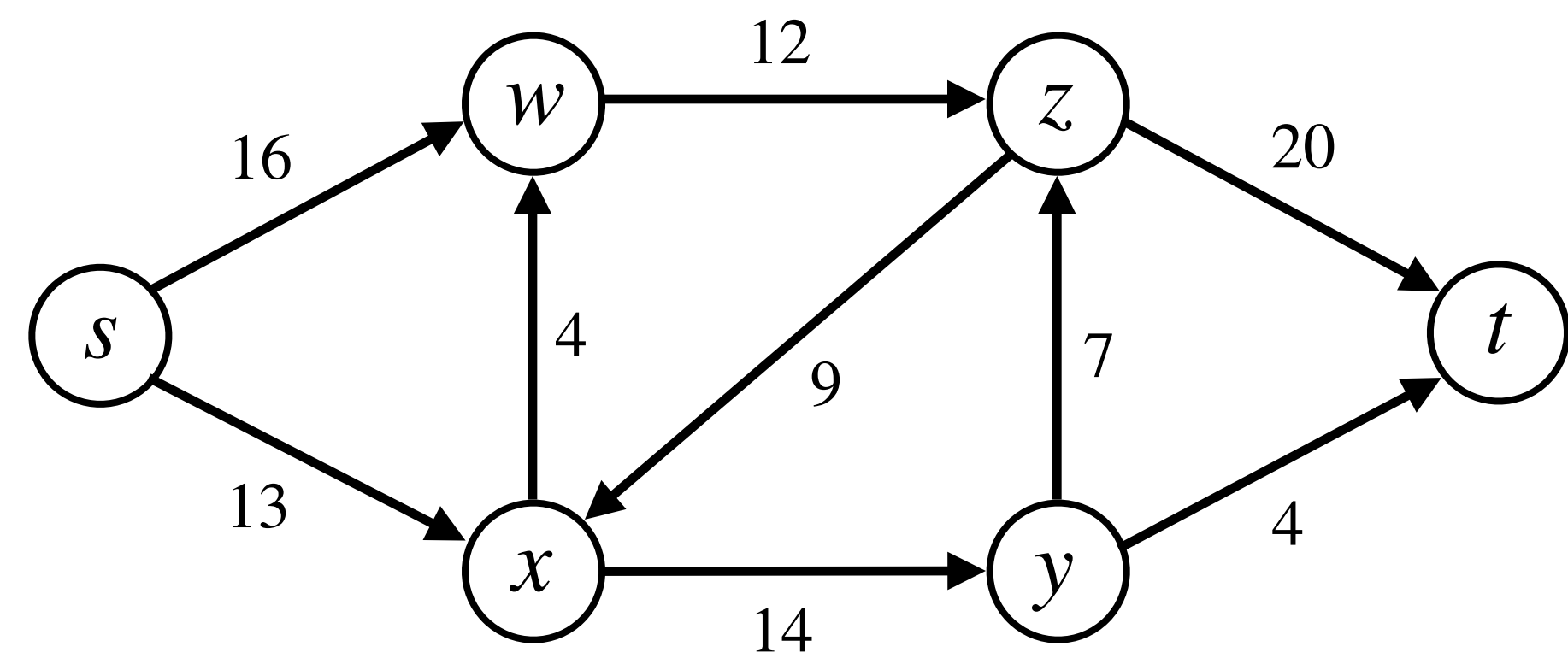
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- Vertices represent **cities**. s & t are the **source** & **sink** cities.
- The number on any (u, v) edge is the **maximum number of packets** that can go from u to v per day.



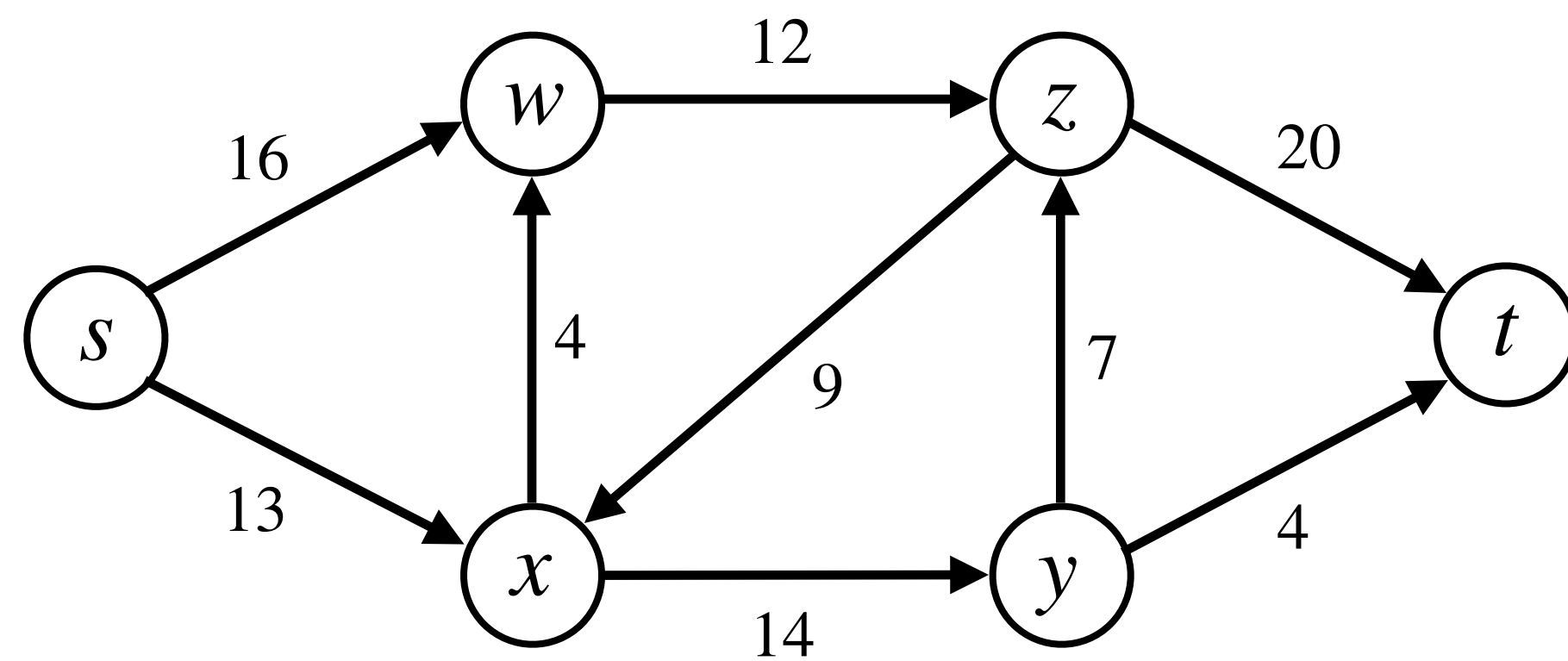
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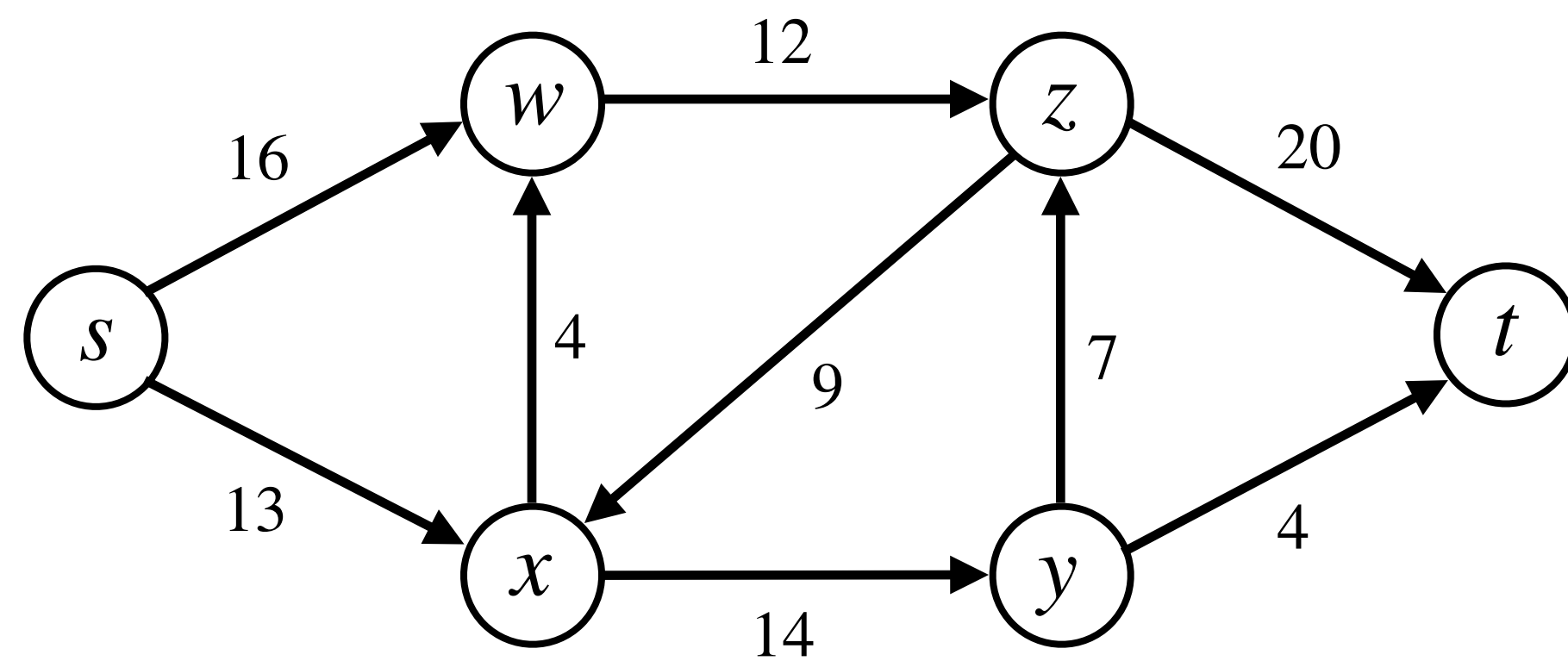
Goal: Find the maximum number of packets that can be shipped from s if the packets received and



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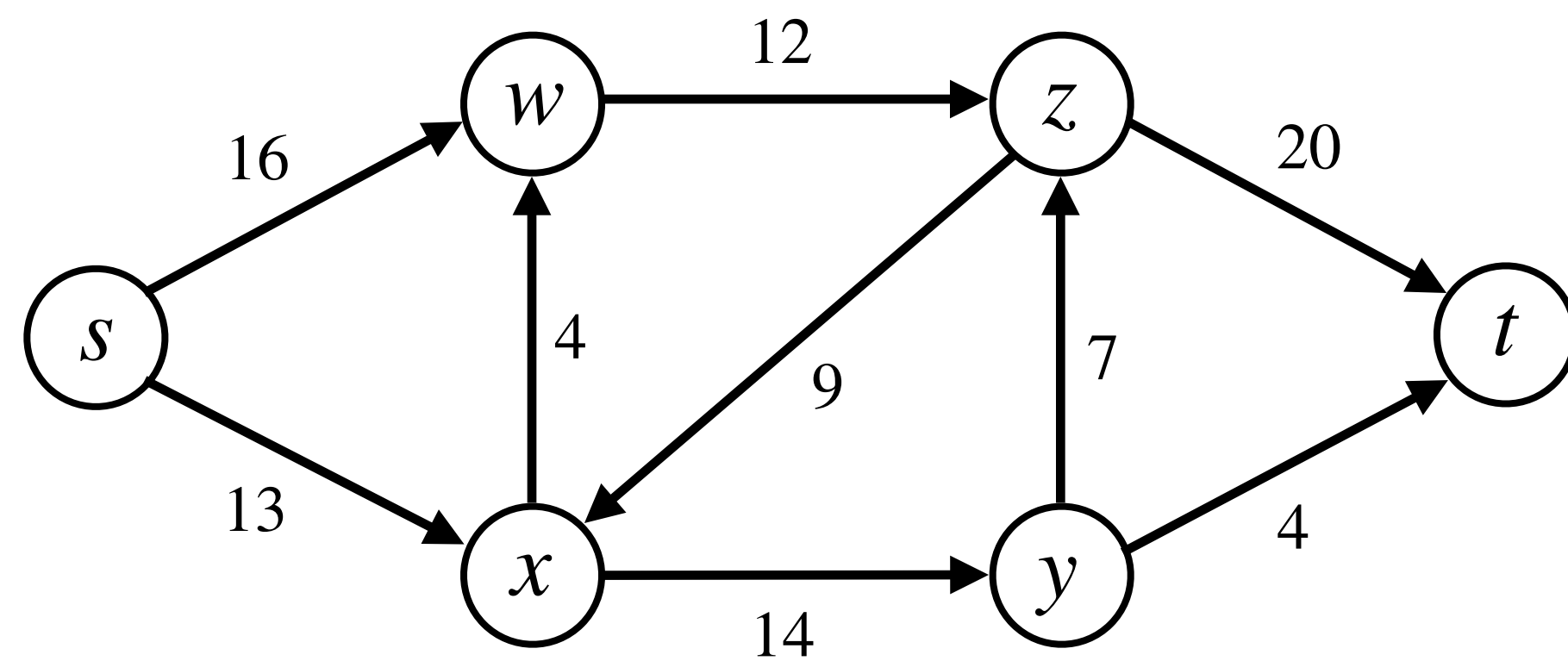
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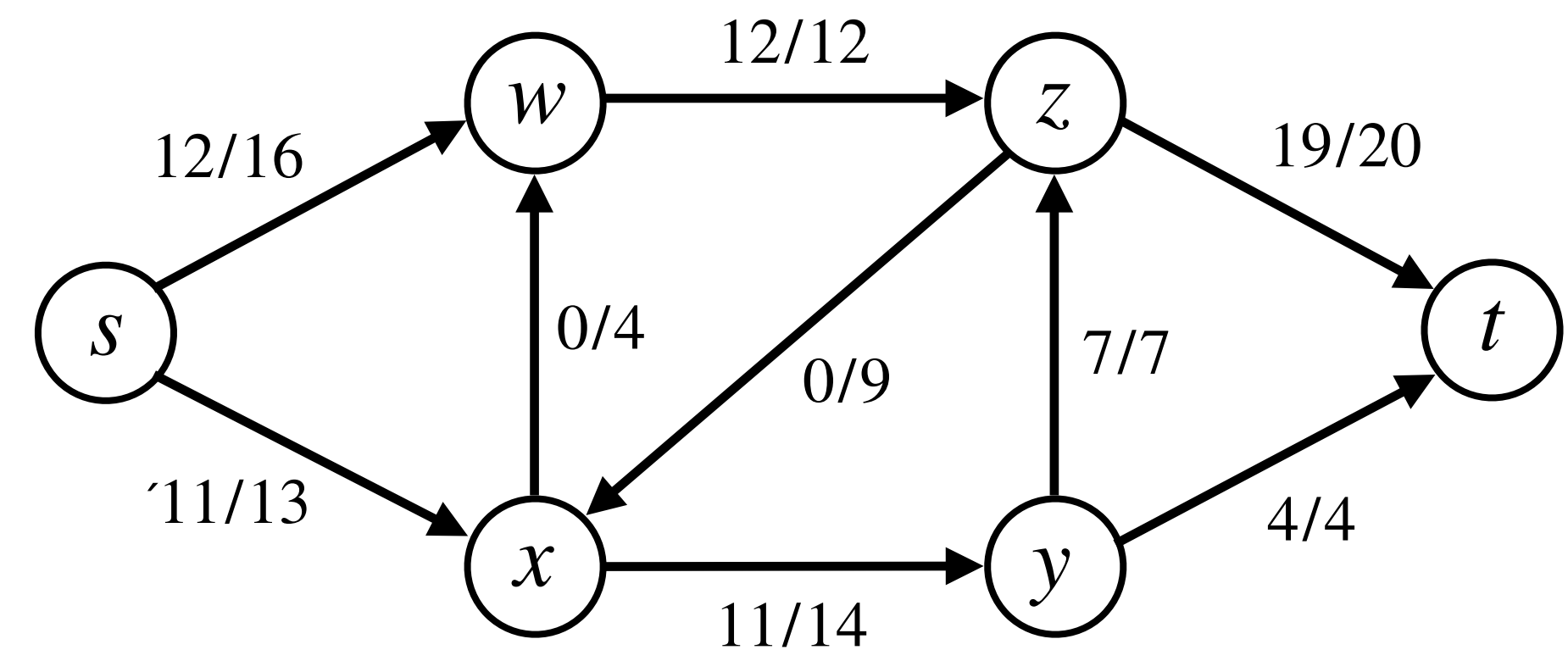
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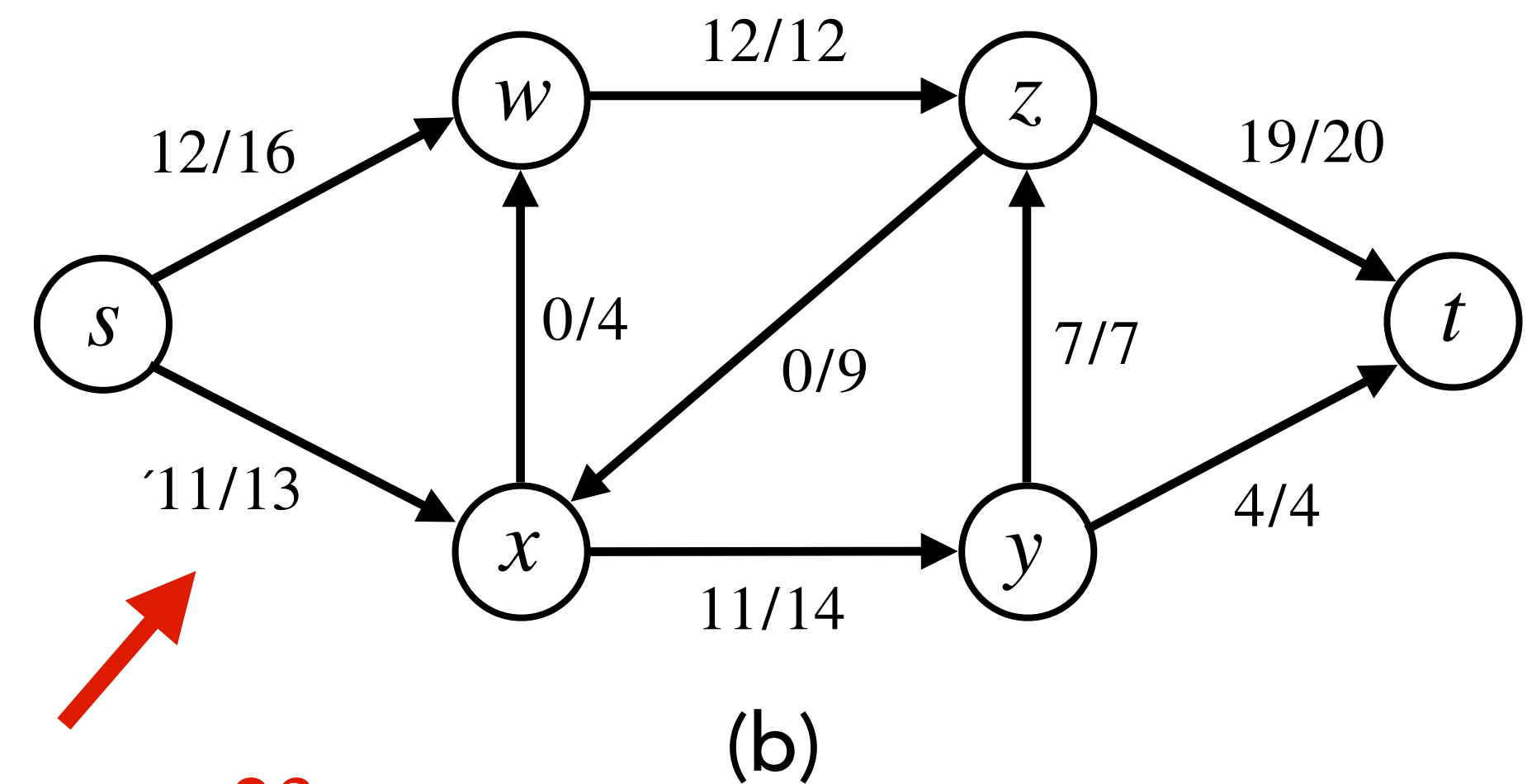
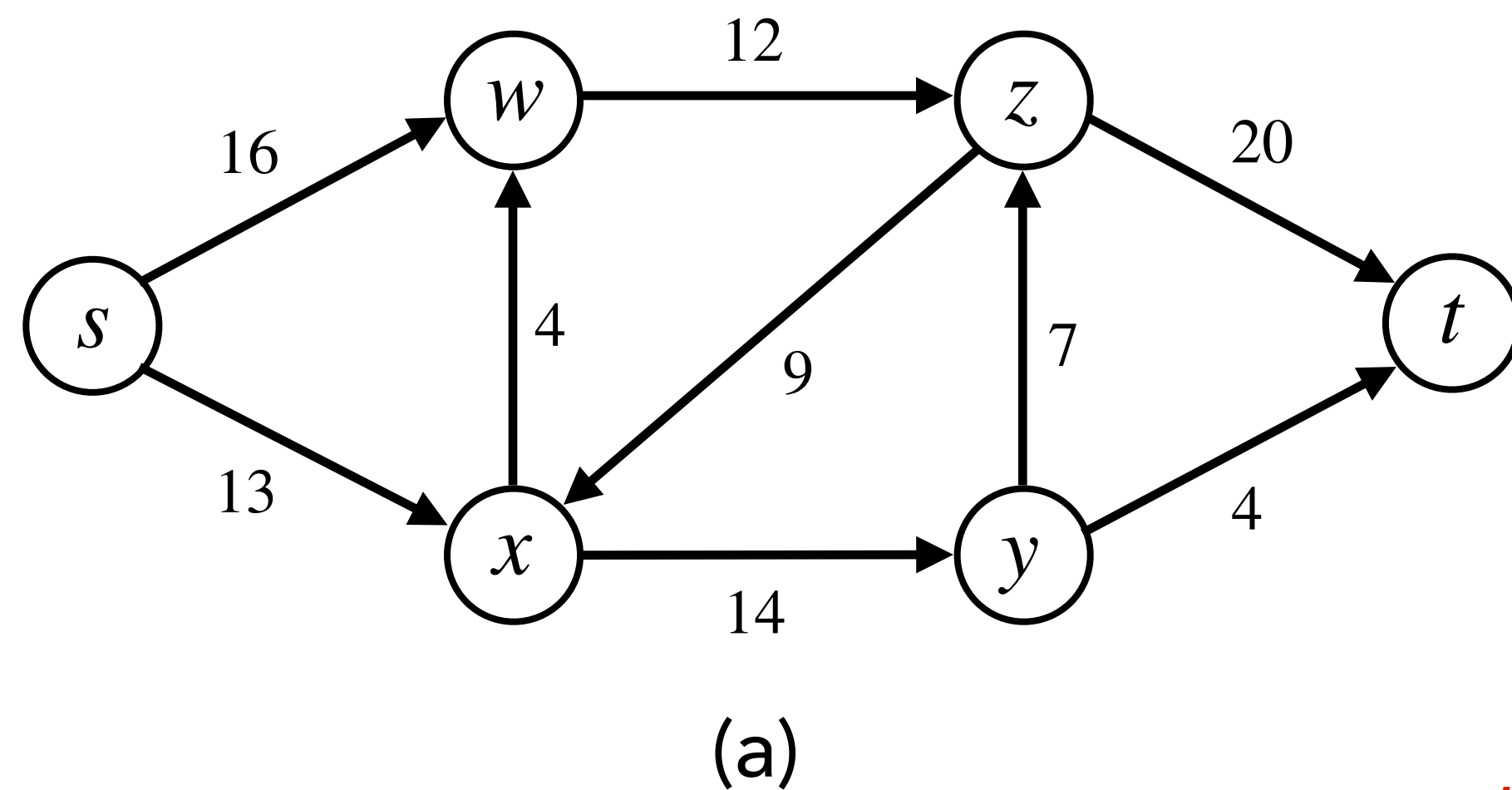
(a)



(b)

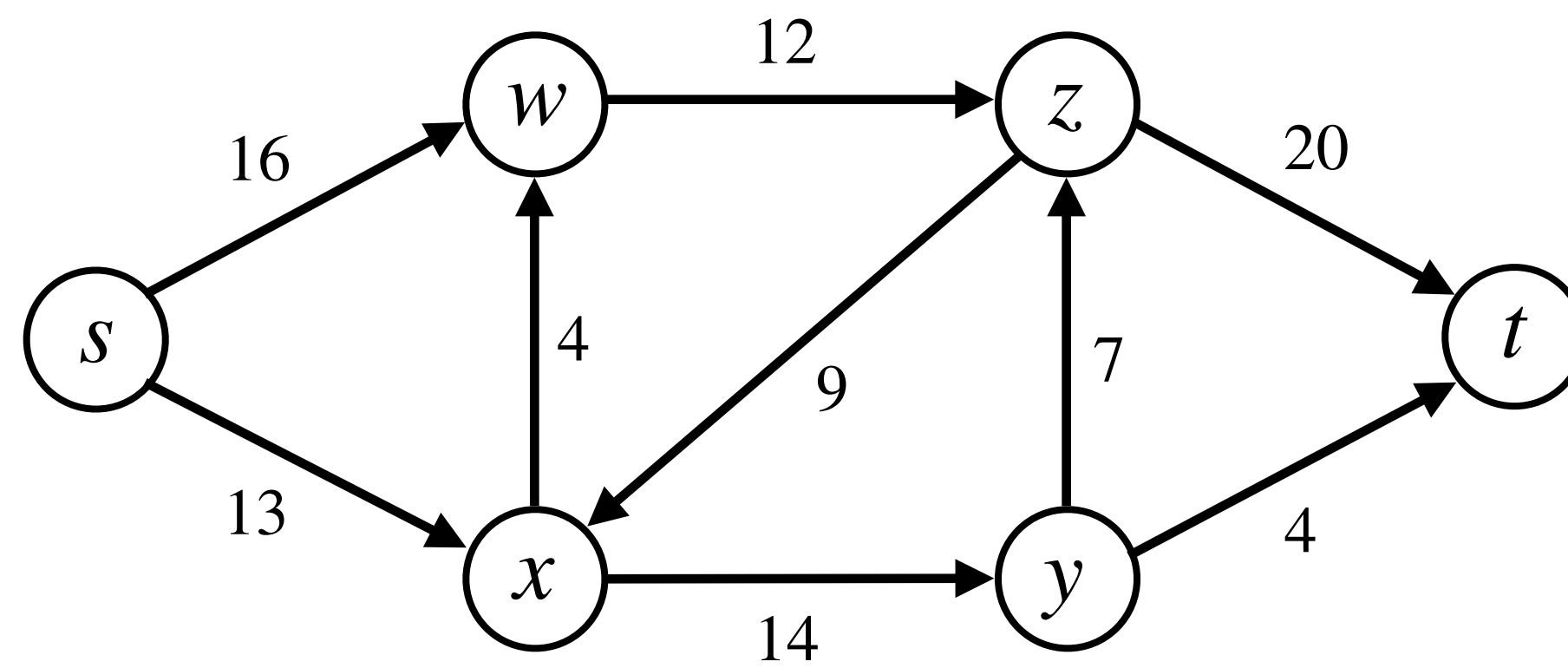
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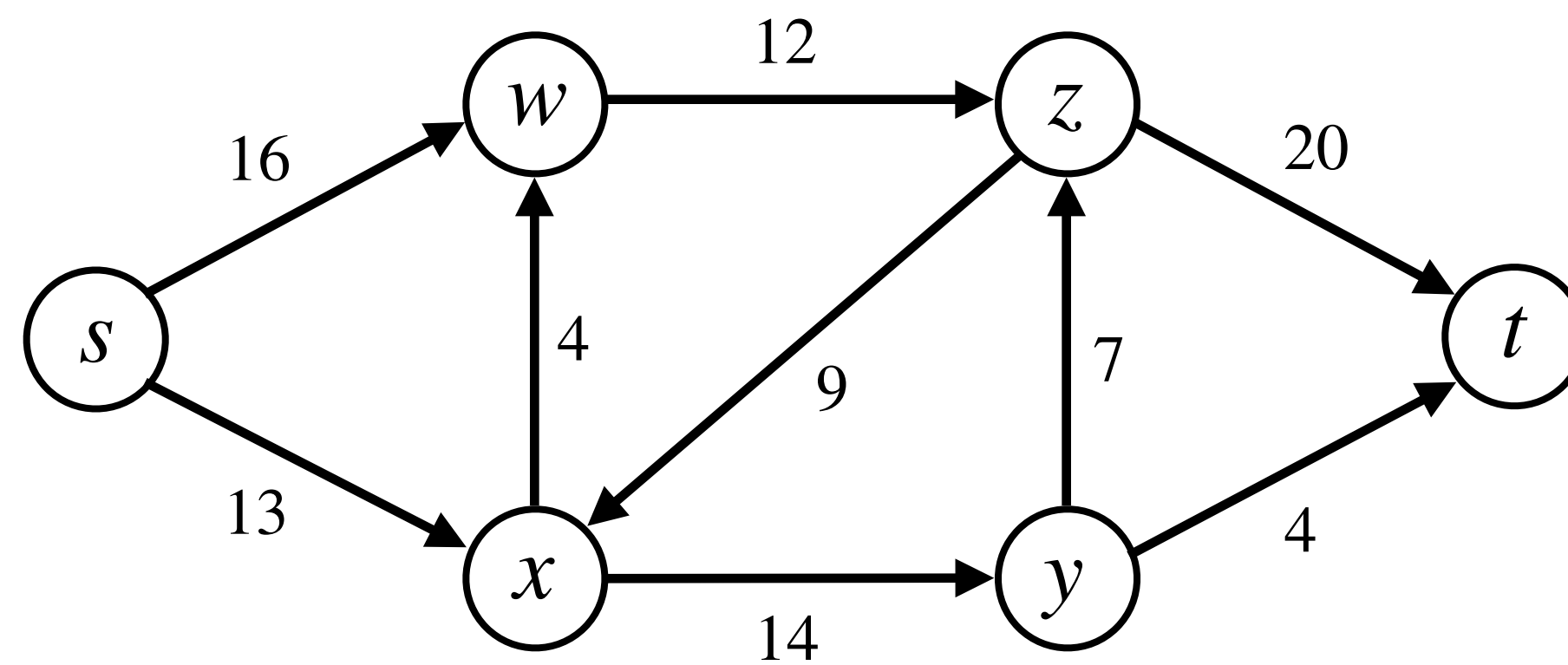
max packets = 23

Flow Networks



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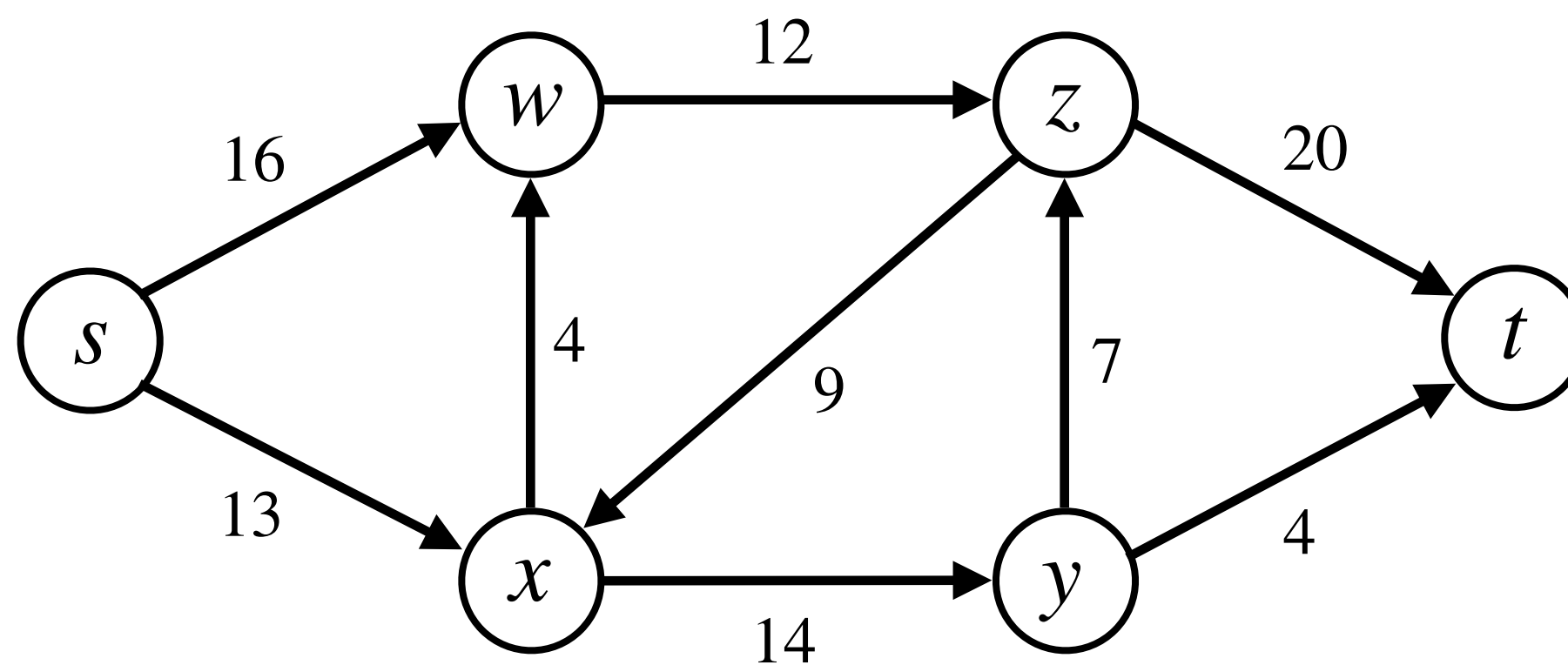
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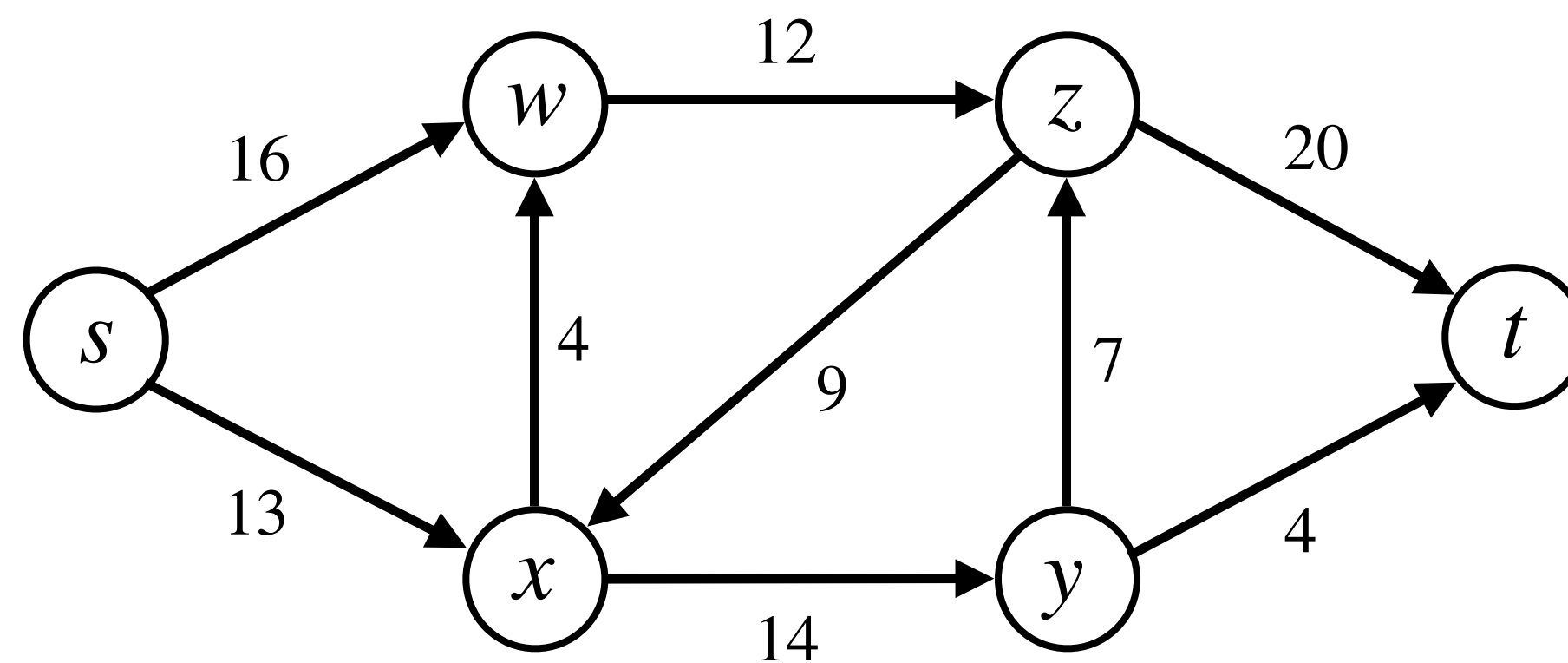
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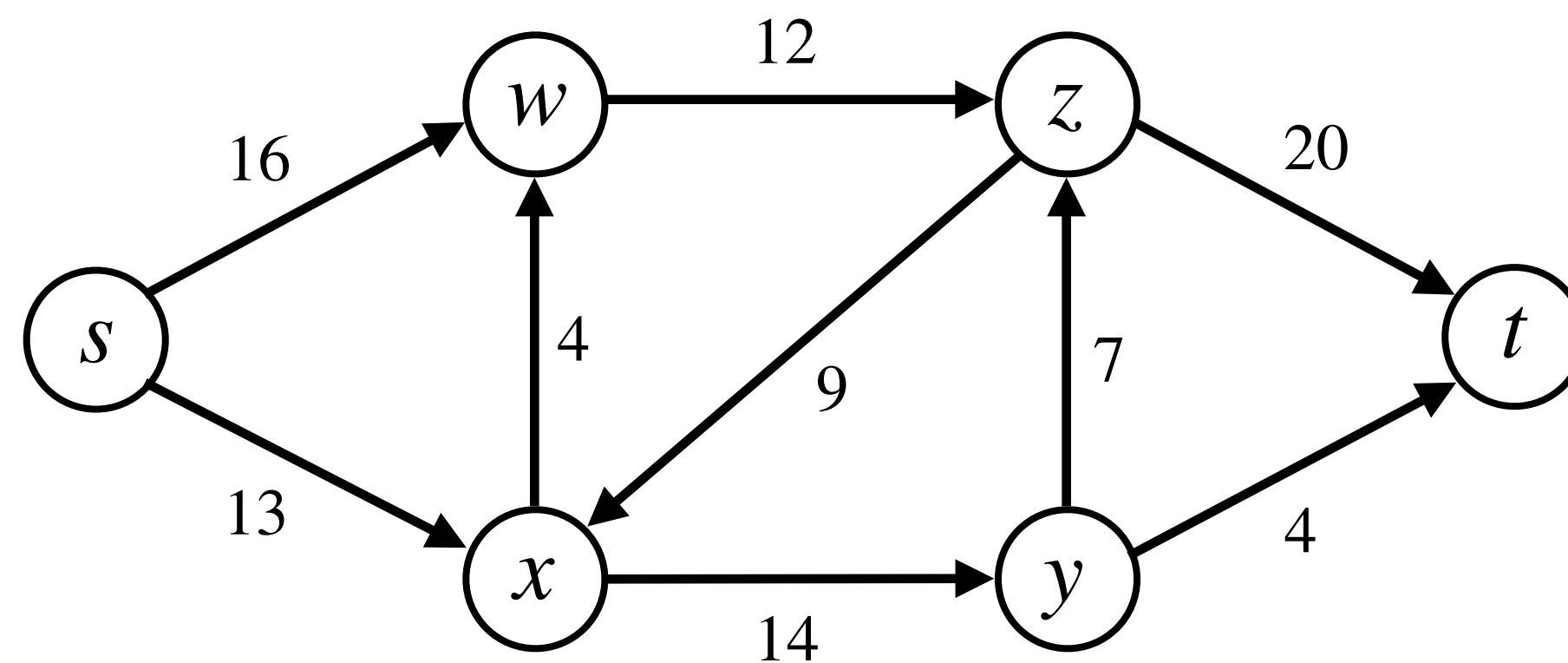
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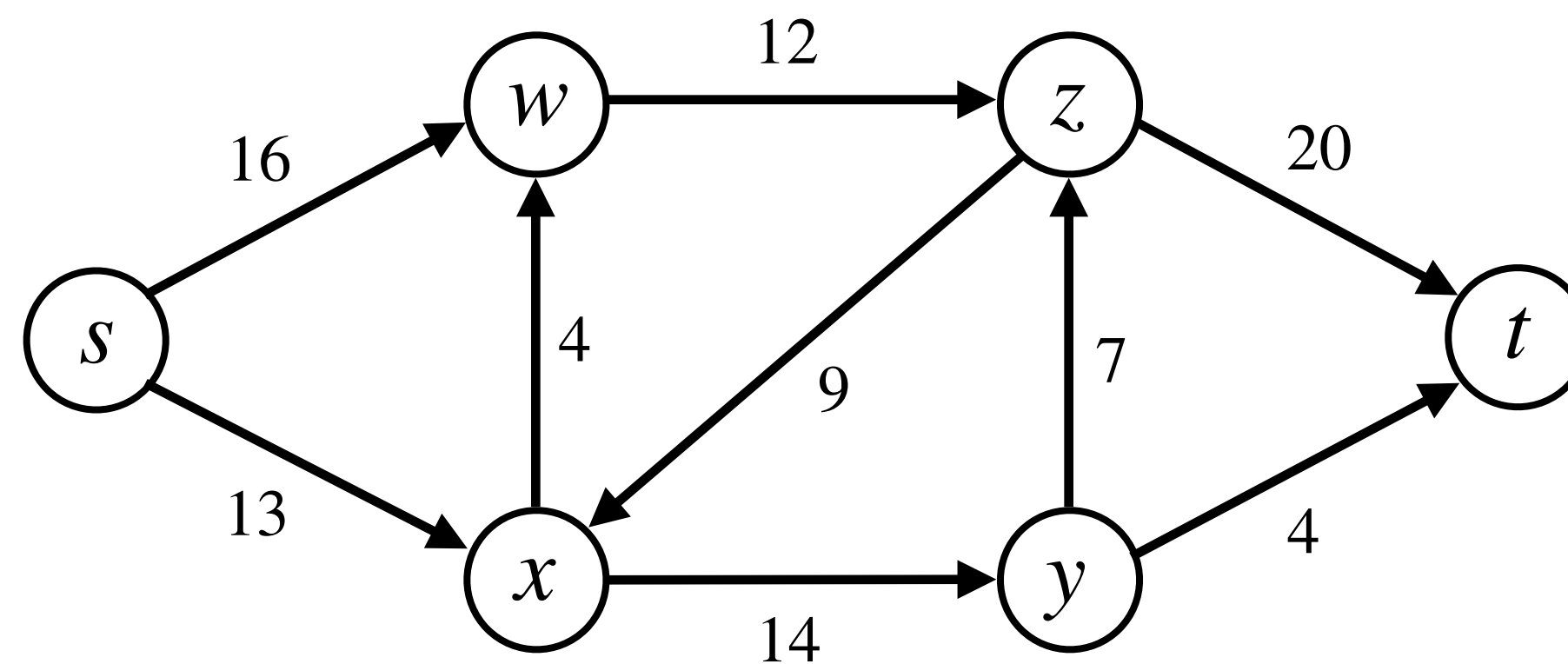
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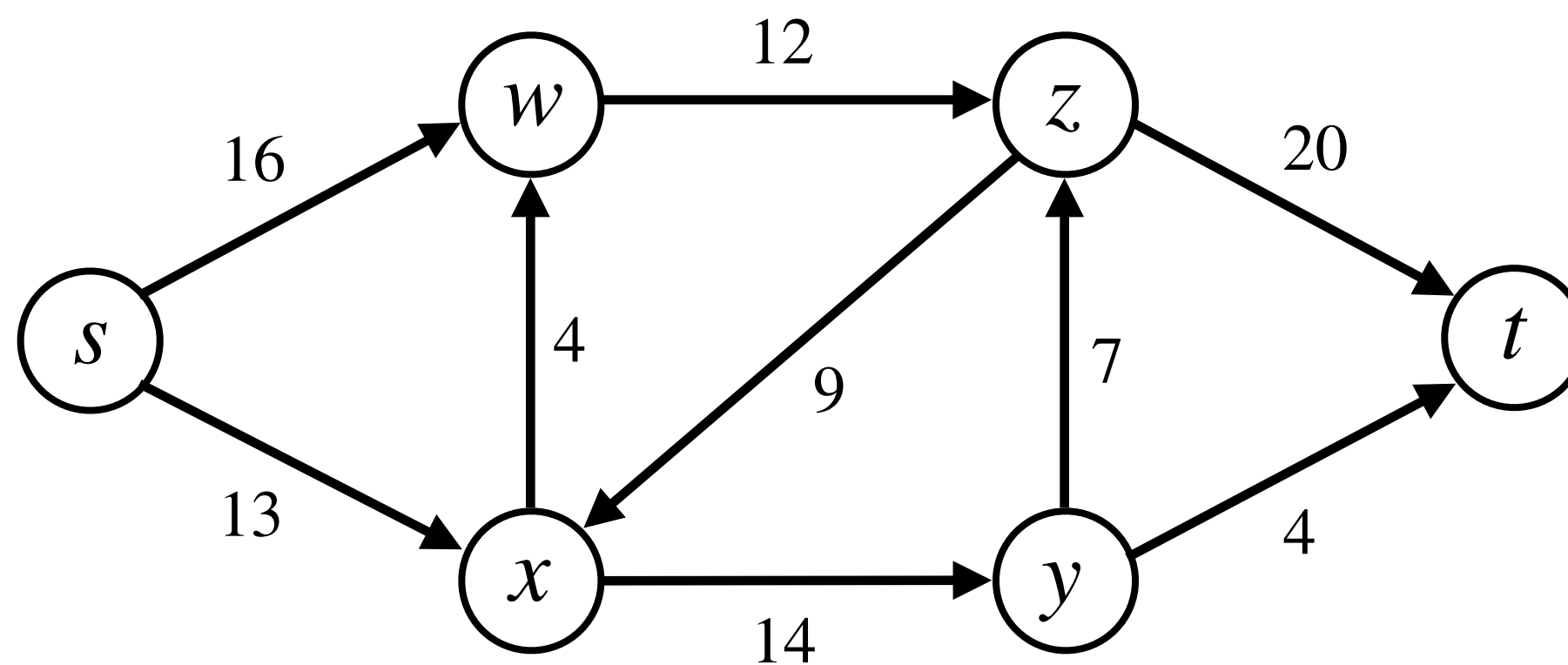
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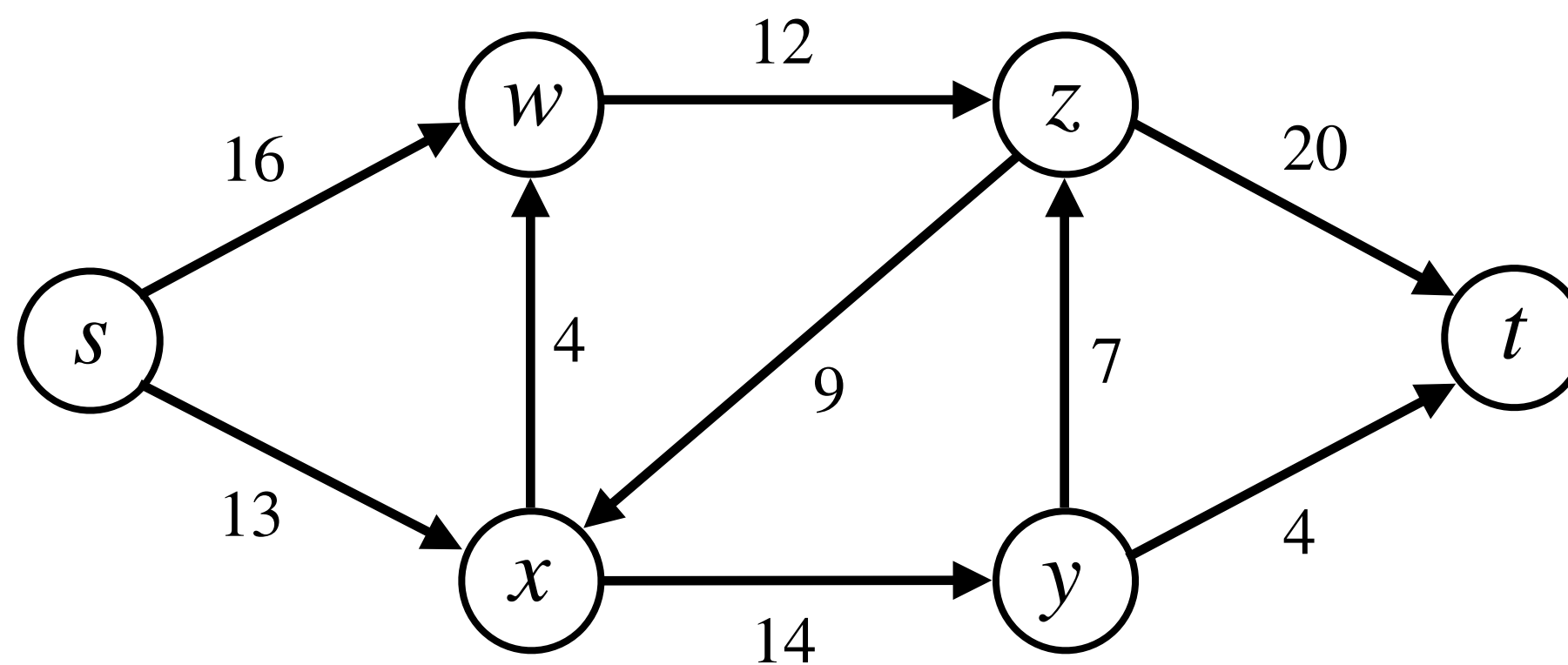
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- Two distinguished vertices: **source** s (no incoming edges) and **sink** t (no outgoing edges).
- For every $v \in V$, some $s \rightsquigarrow v \rightsquigarrow t$ path exists. Hence, $|E| \geq |V| - 1$.



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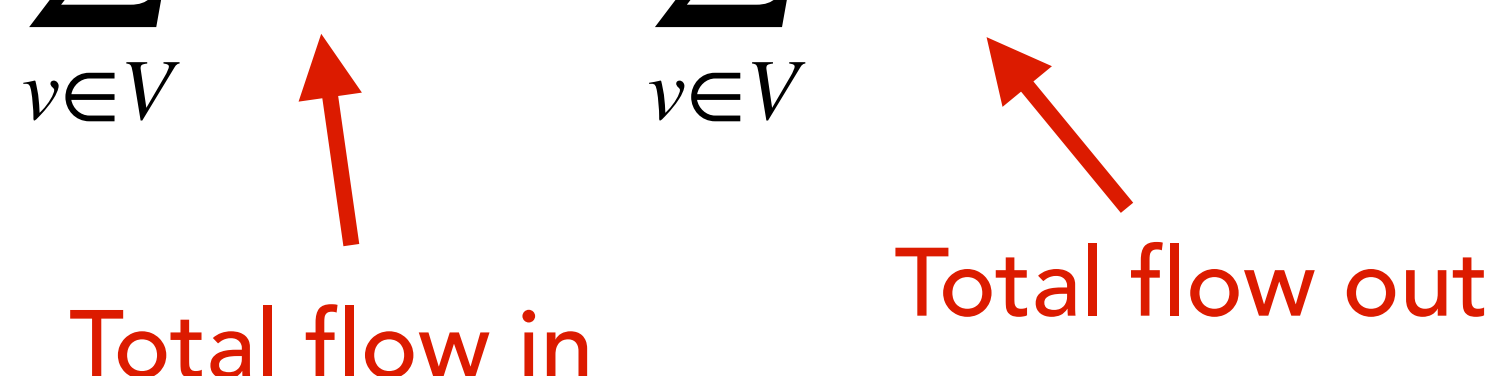
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The diagram shows the flow conservation equation $\sum_{v \in V} f(v, u) = \sum_{v \in V} f(u, v)$. A red arrow points from the text "Total flow in" to the left-hand side of the equation, $\sum_{v \in V} f(v, u)$. Another red arrow points from the text "Total flow out" to the right-hand side of the equation, $\sum_{v \in V} f(u, v)$.

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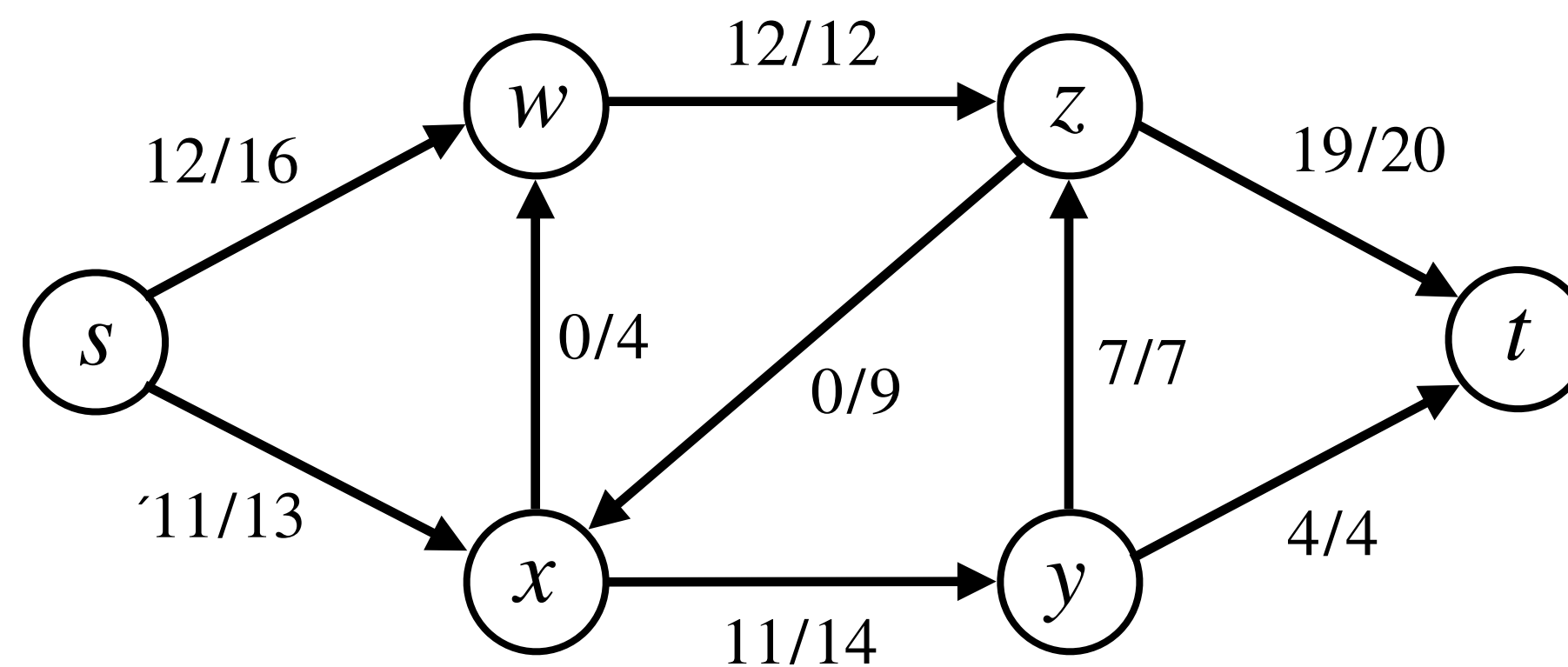
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Example:



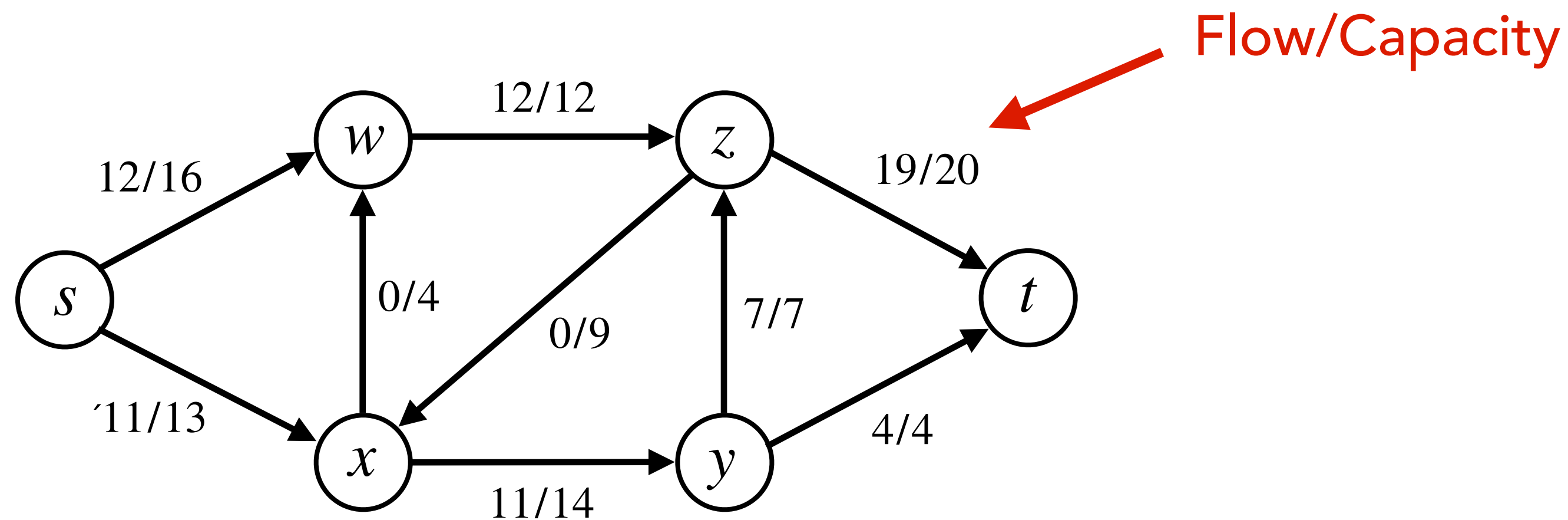
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Output: Flow of maximum value.

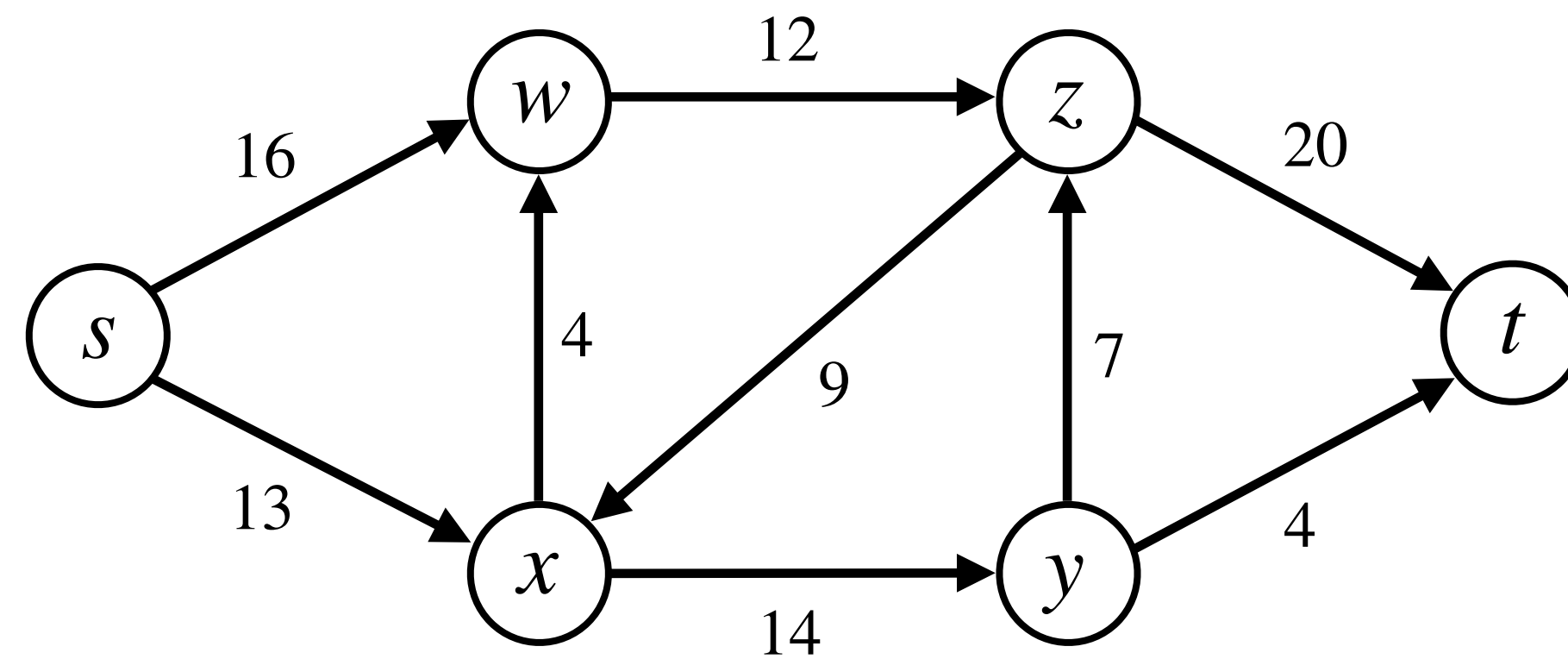
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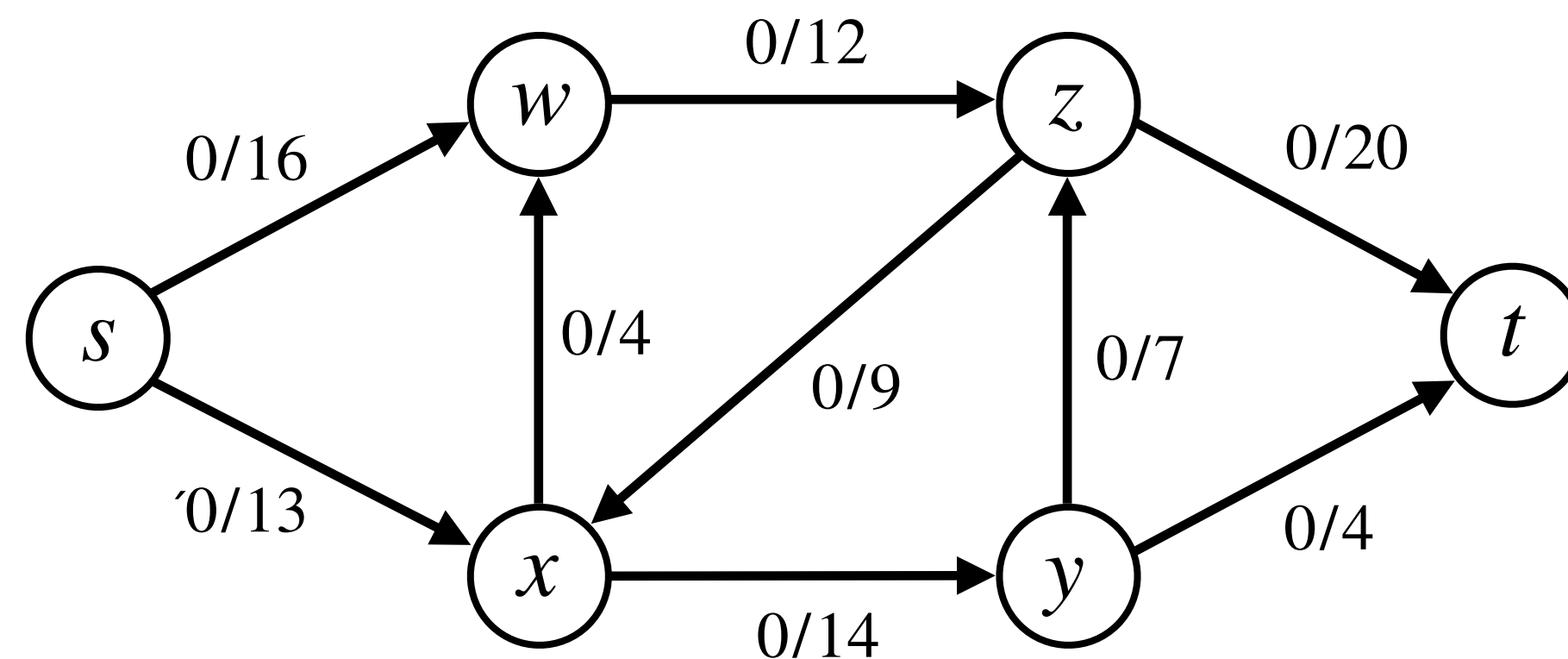


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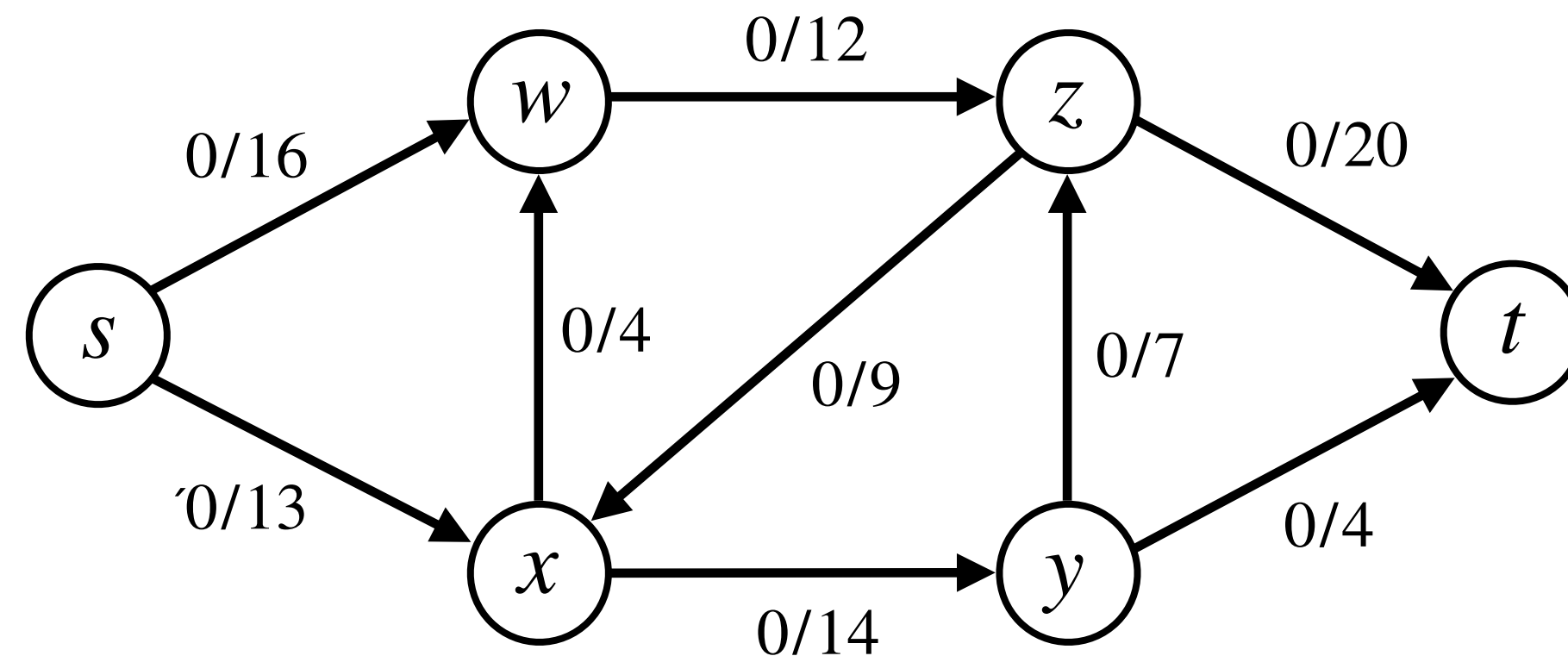


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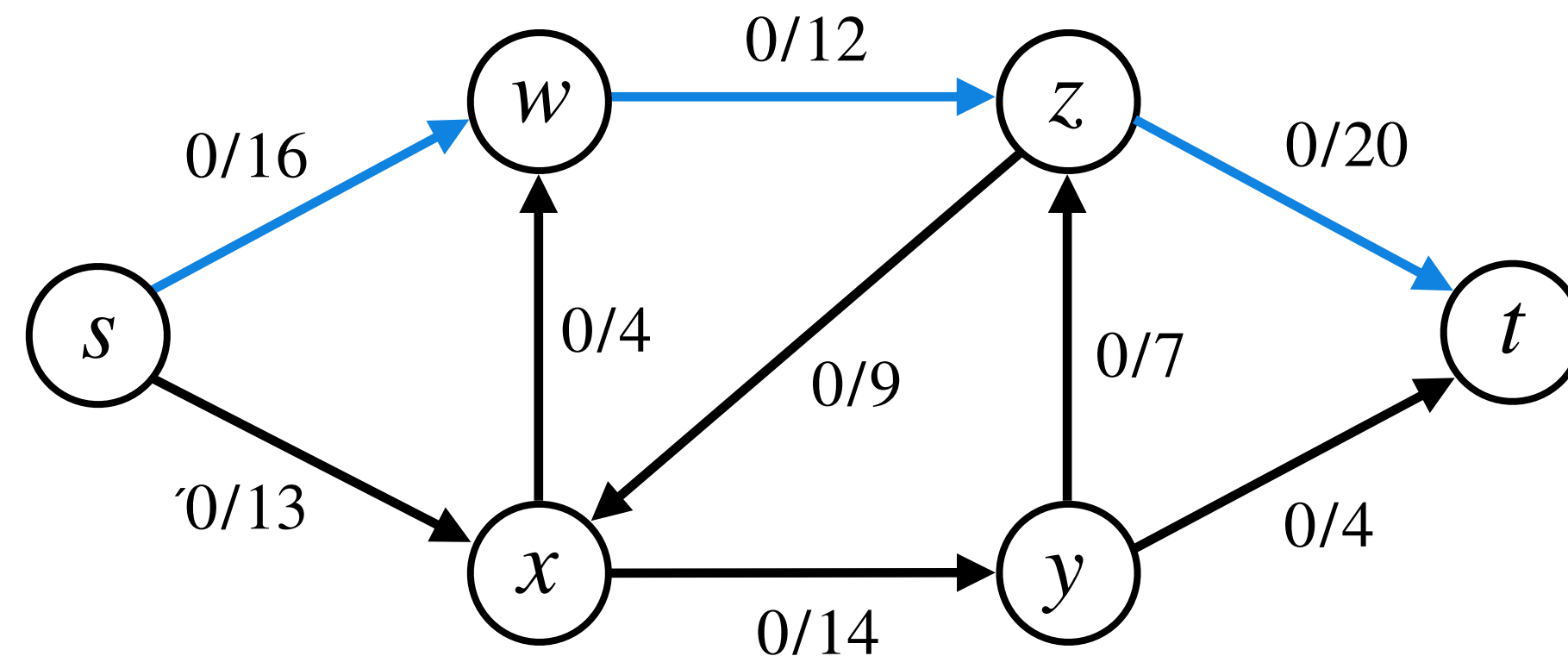


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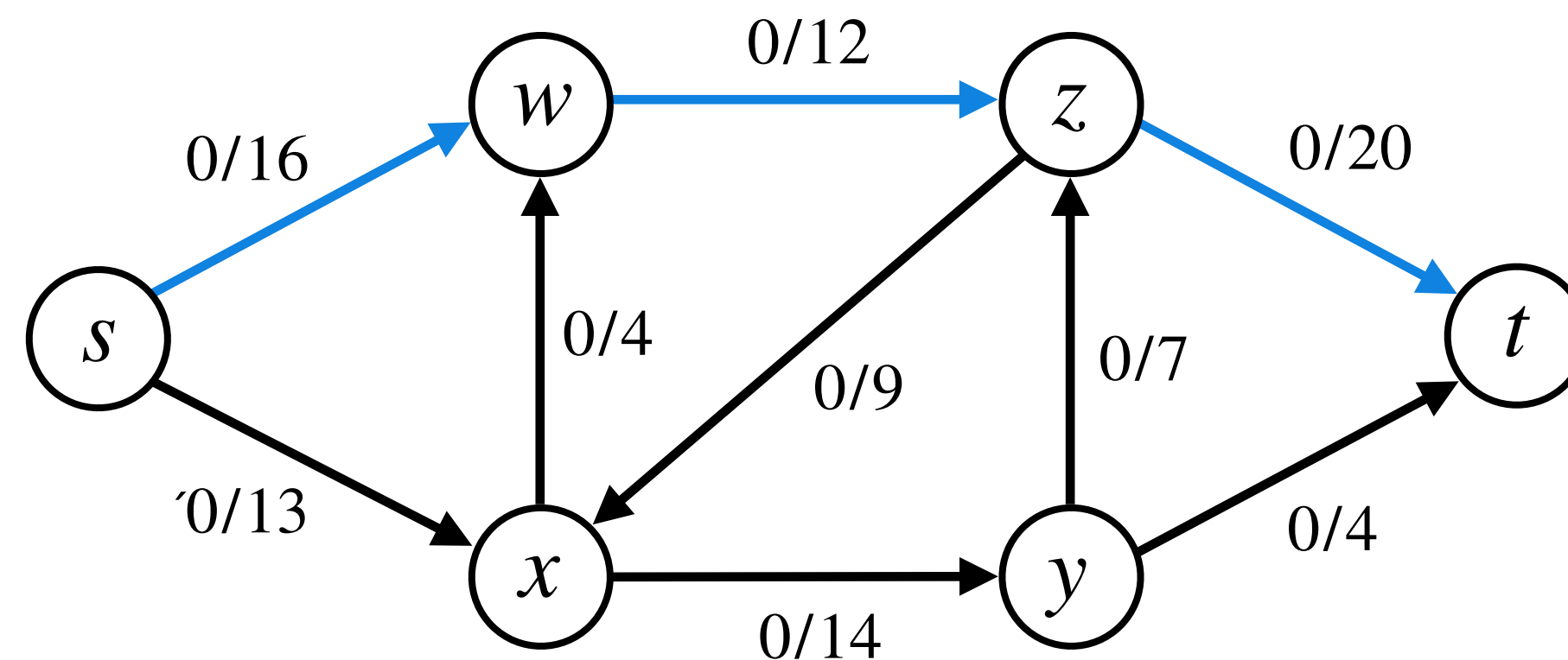


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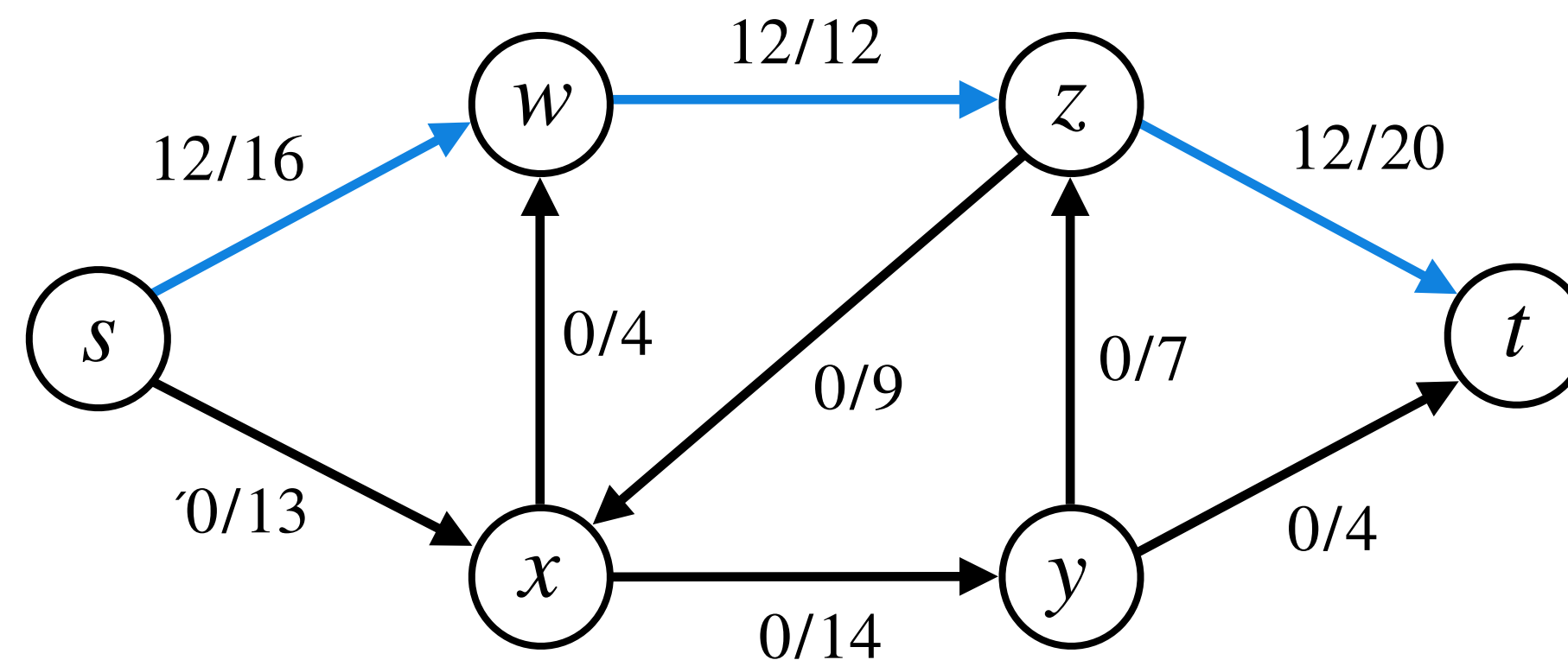


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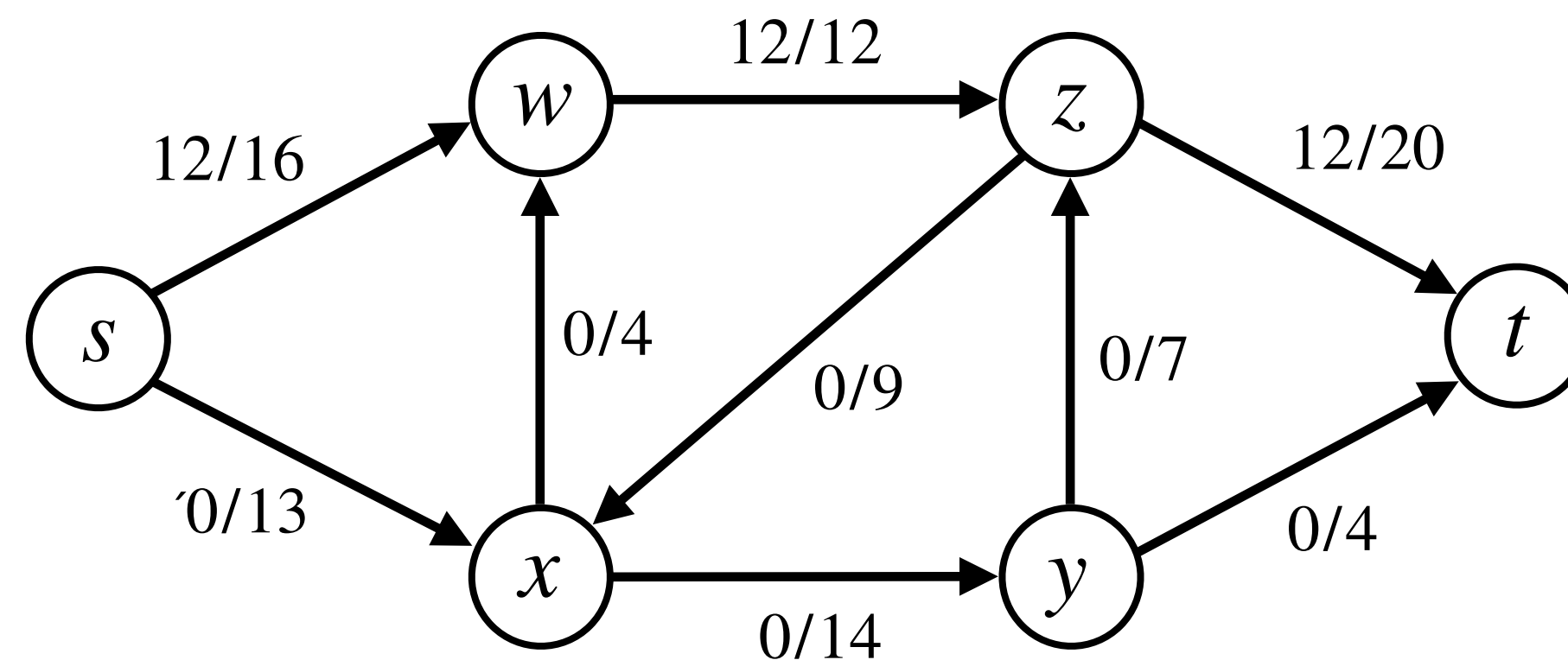


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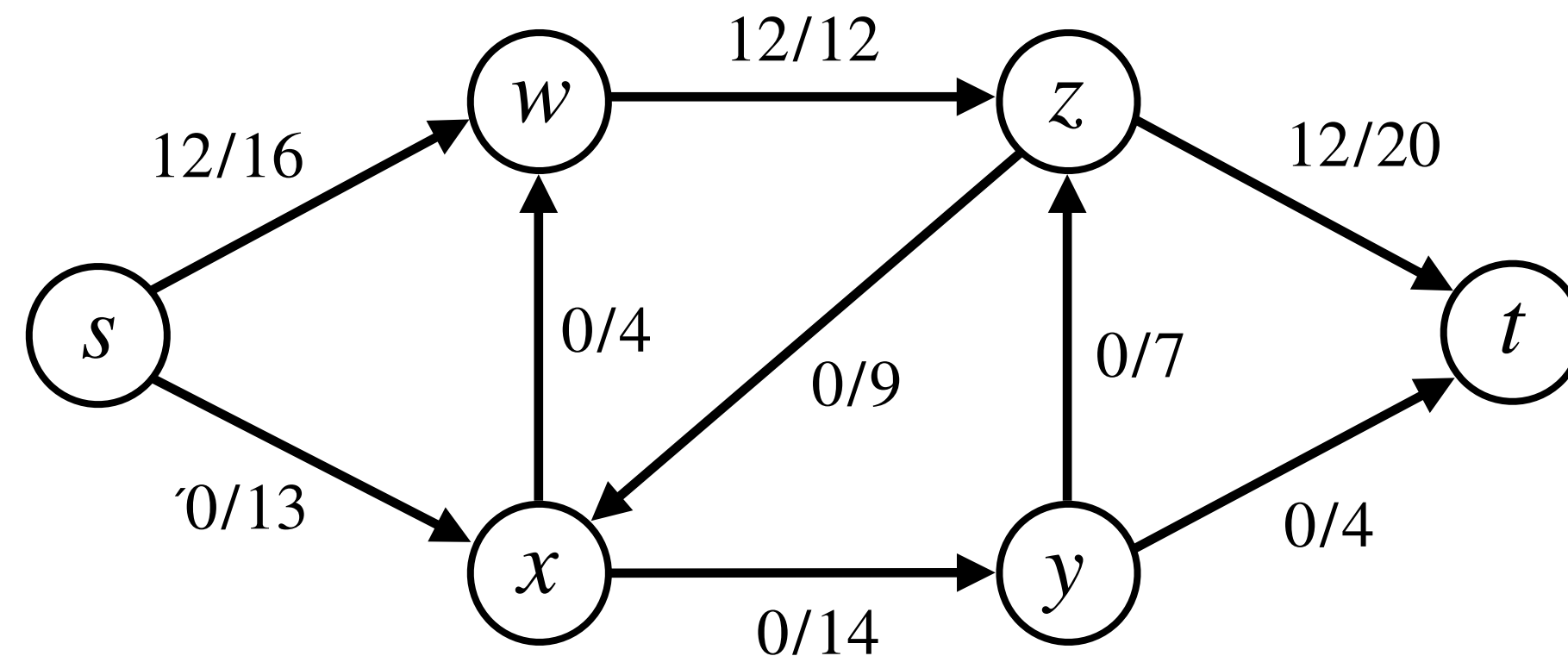


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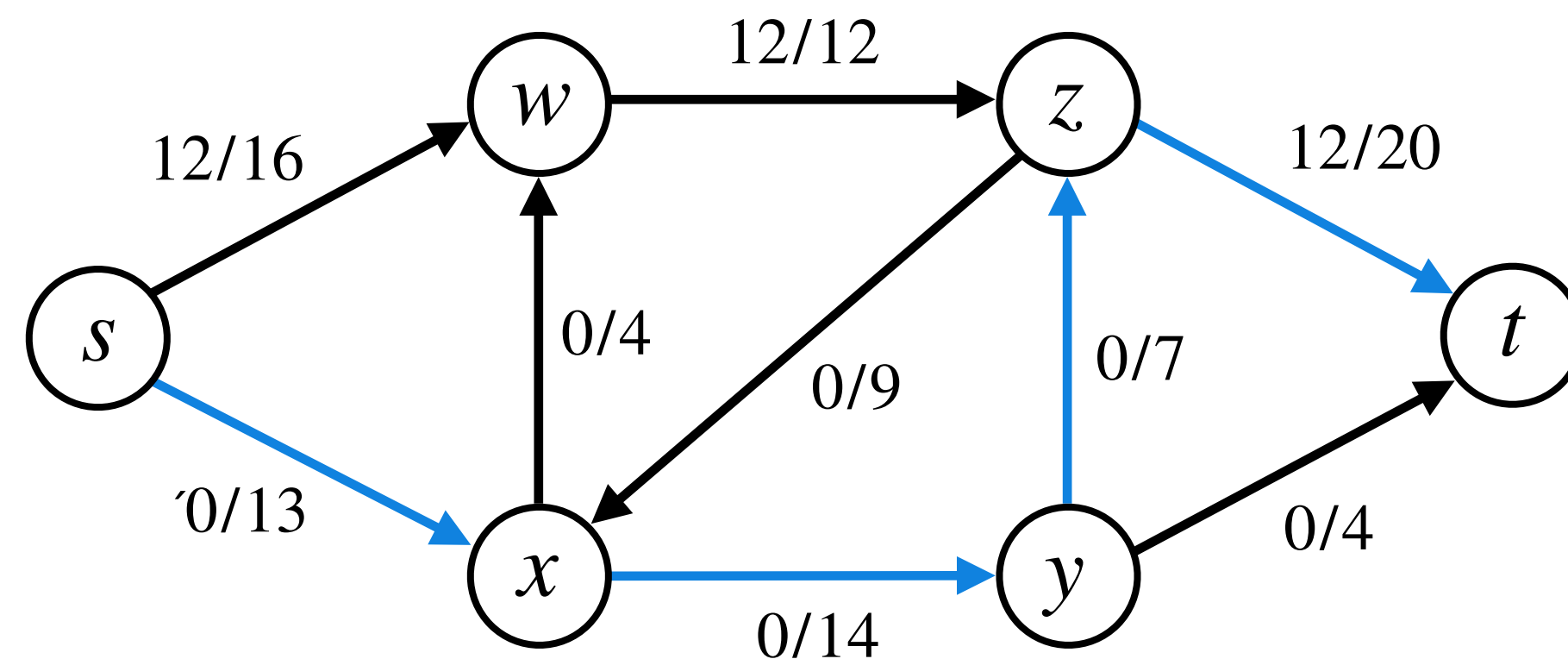


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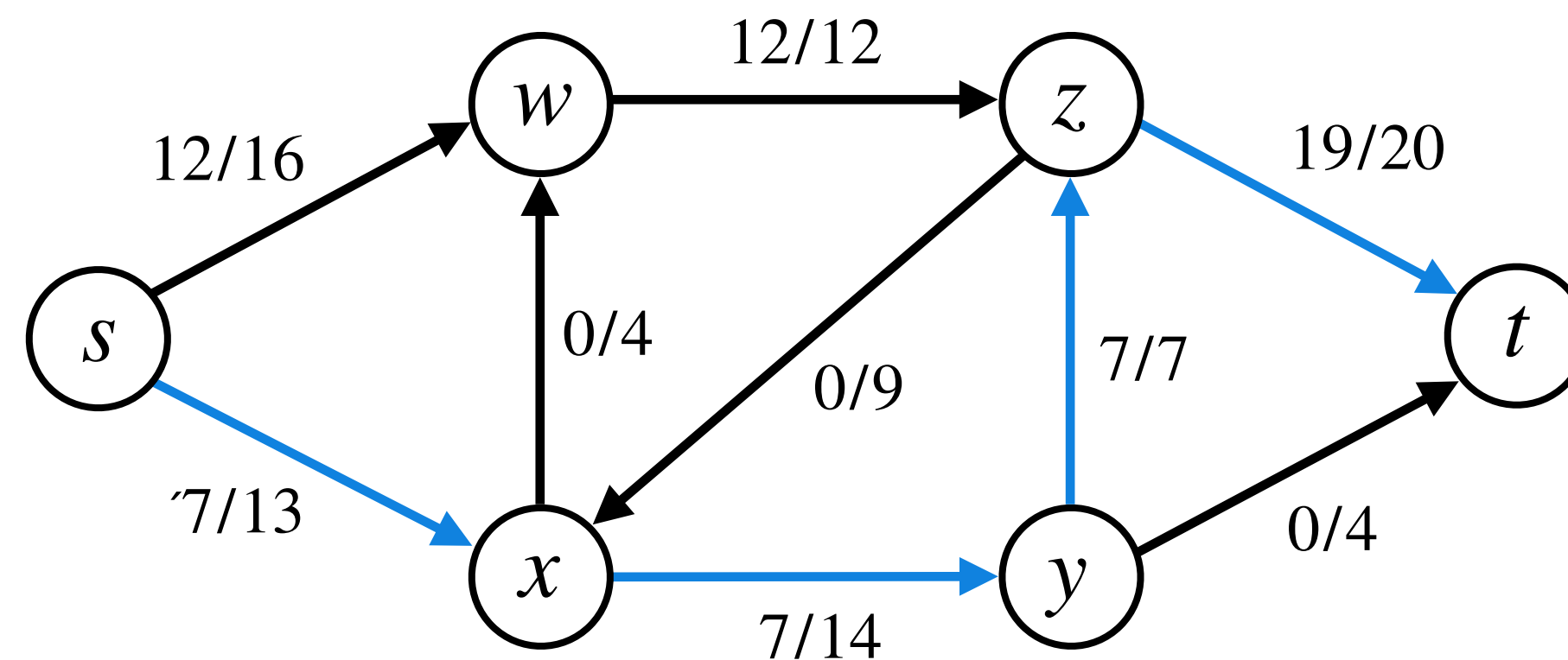


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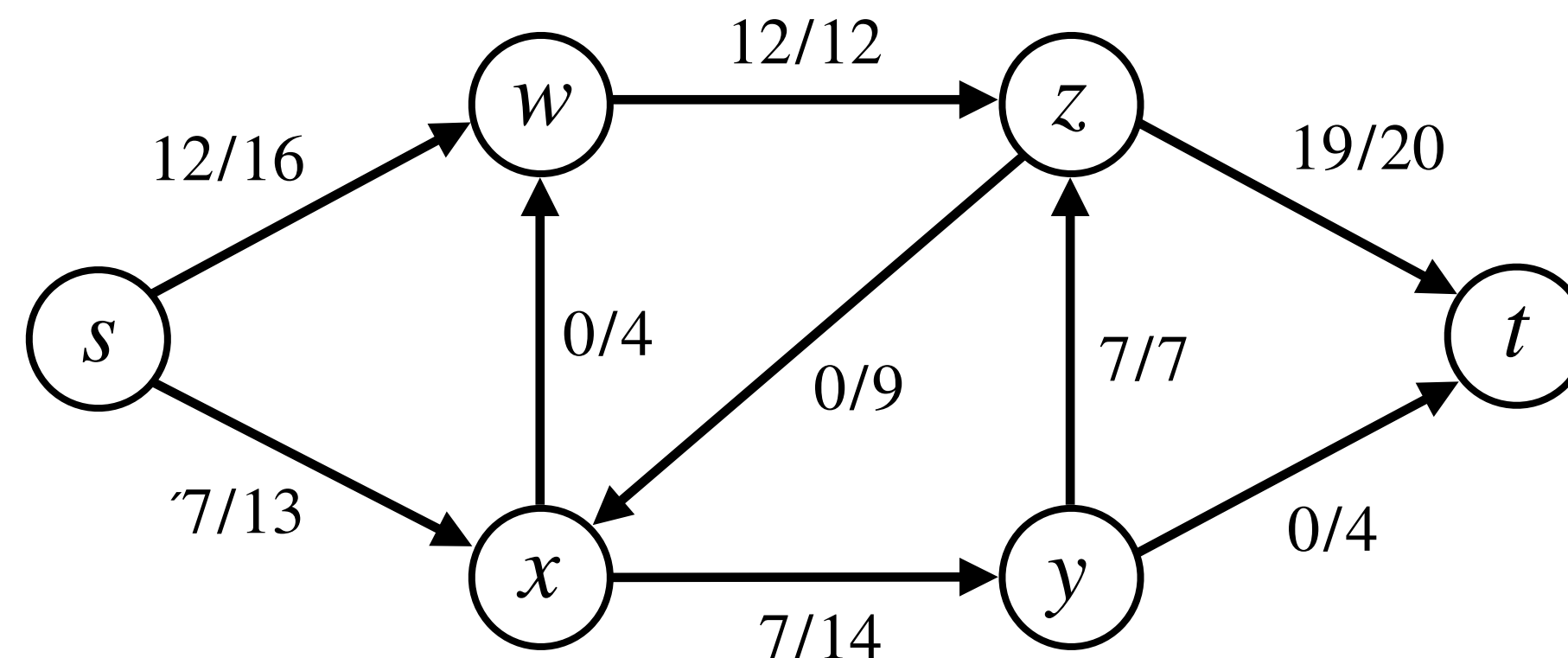


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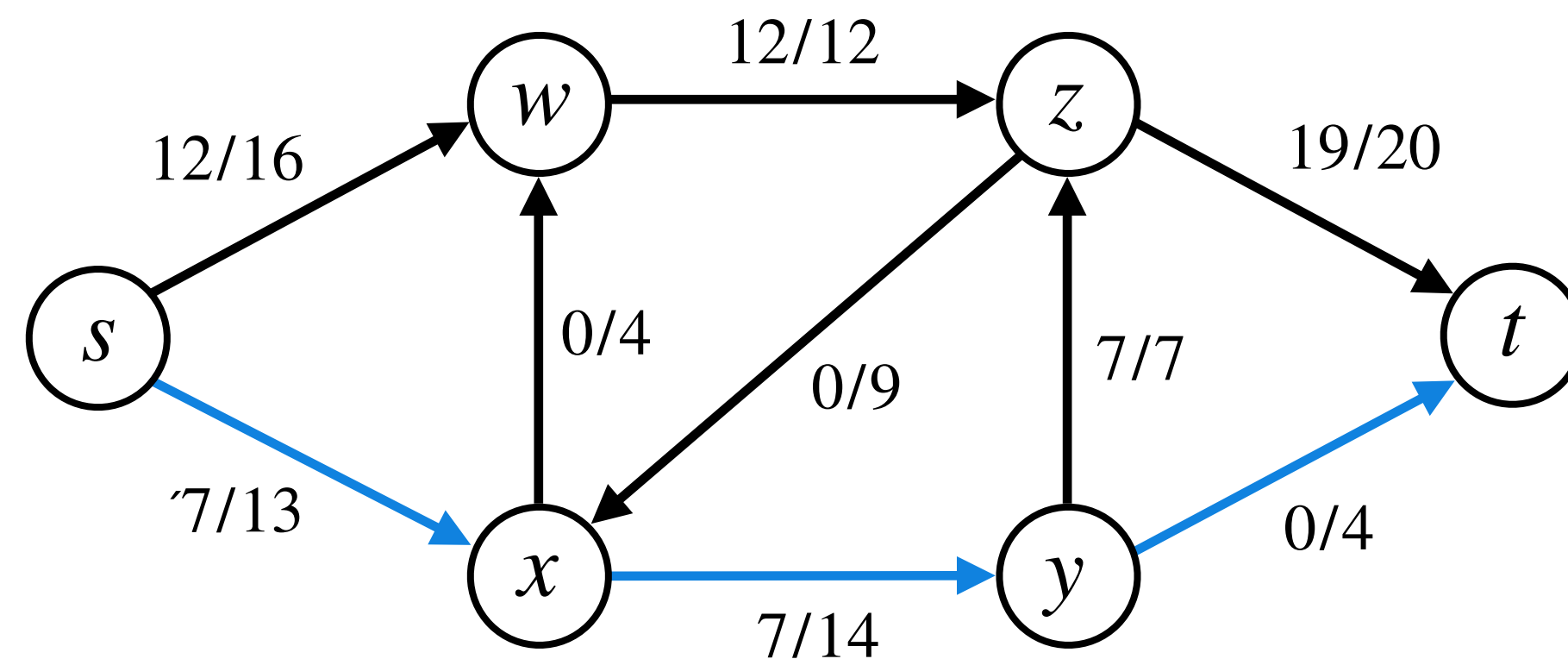


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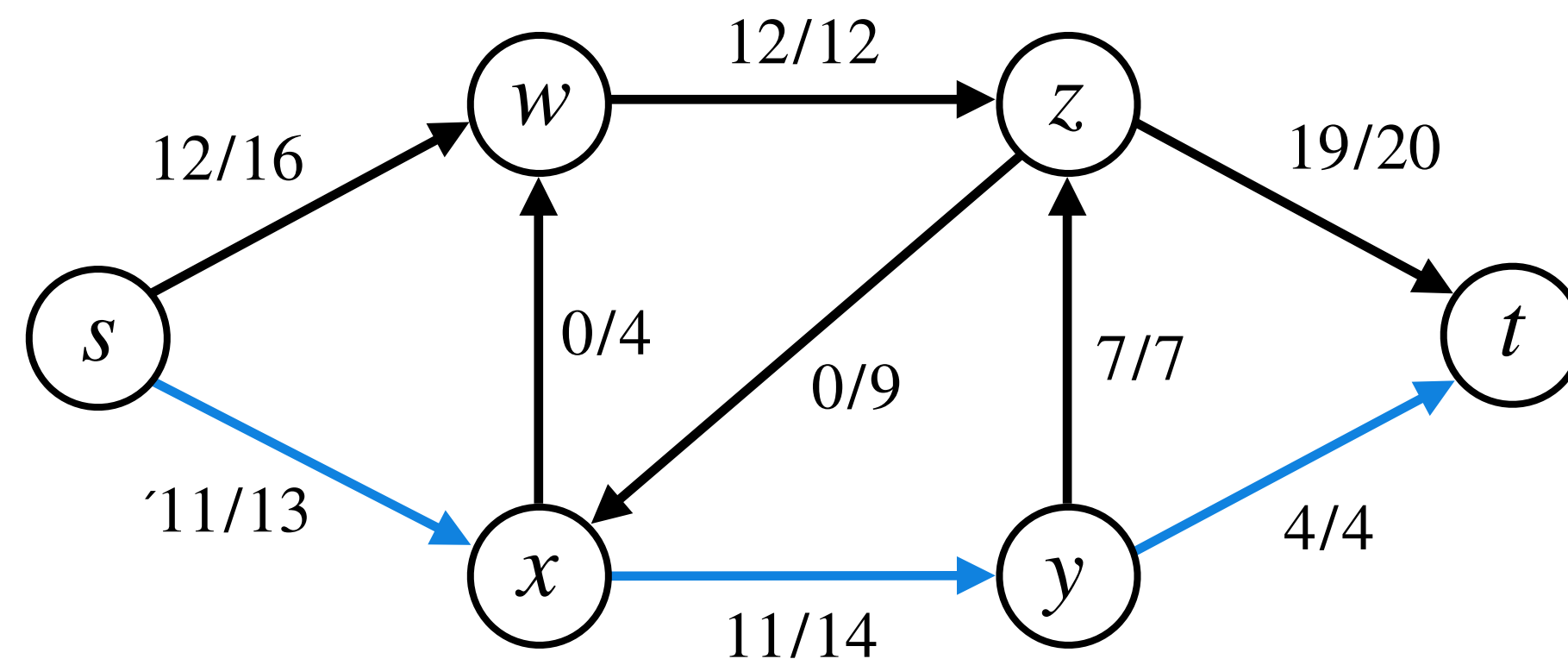


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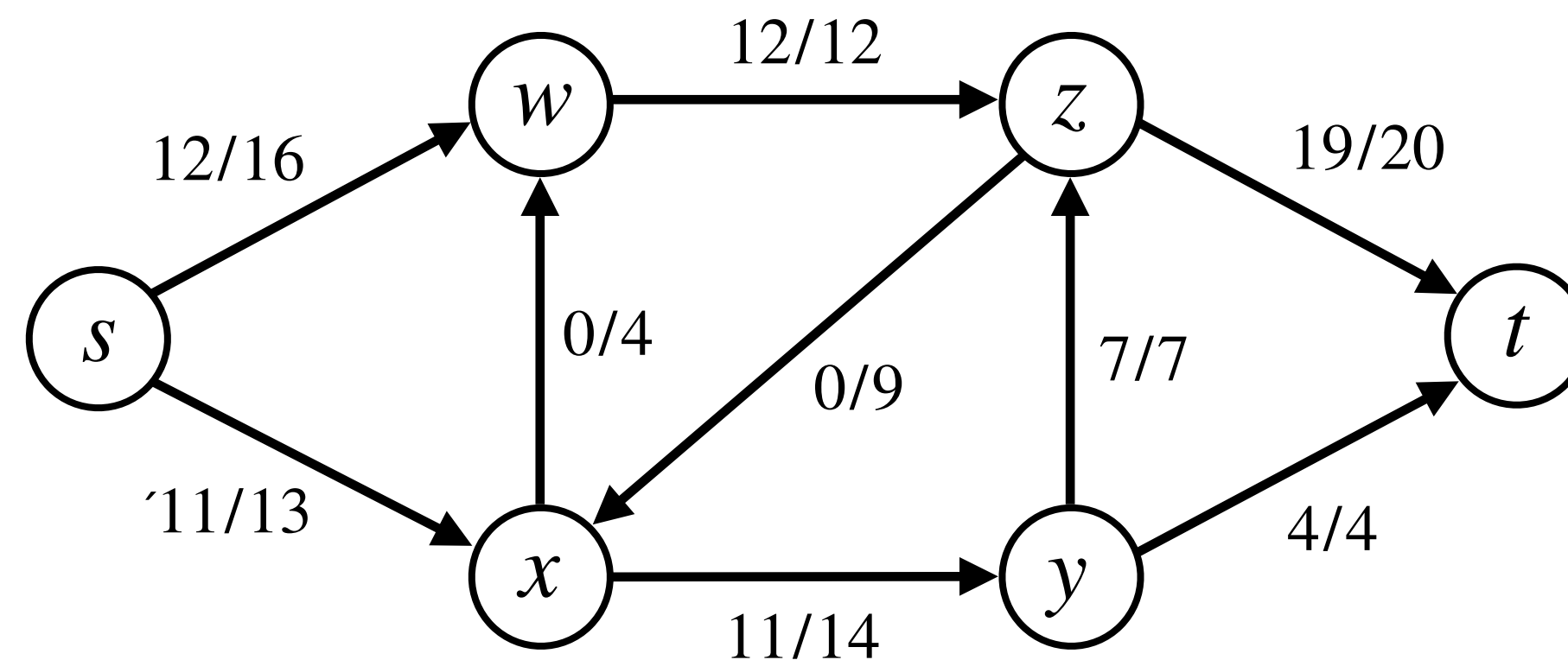


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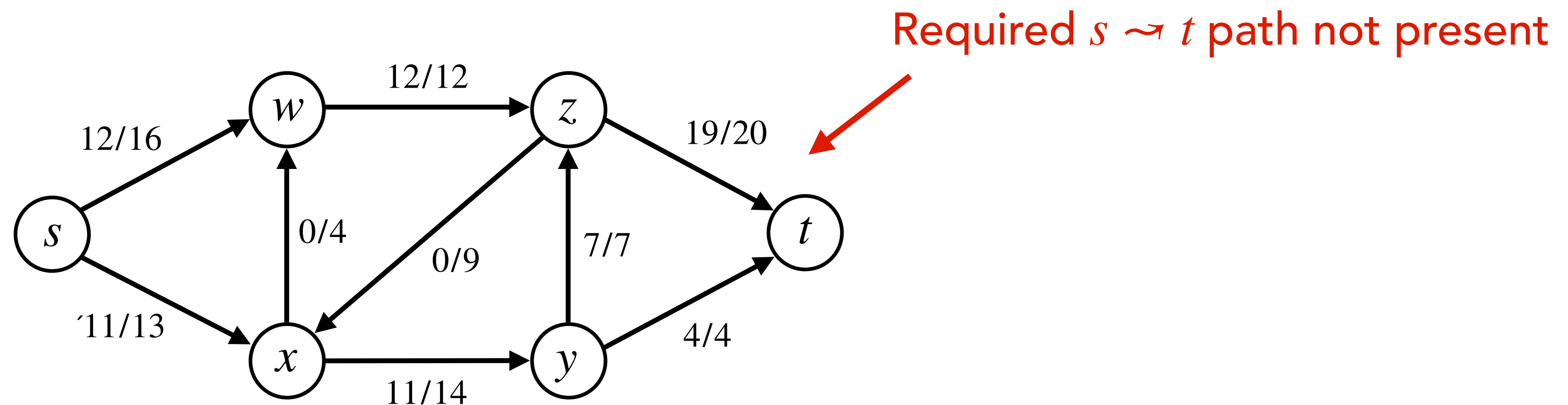


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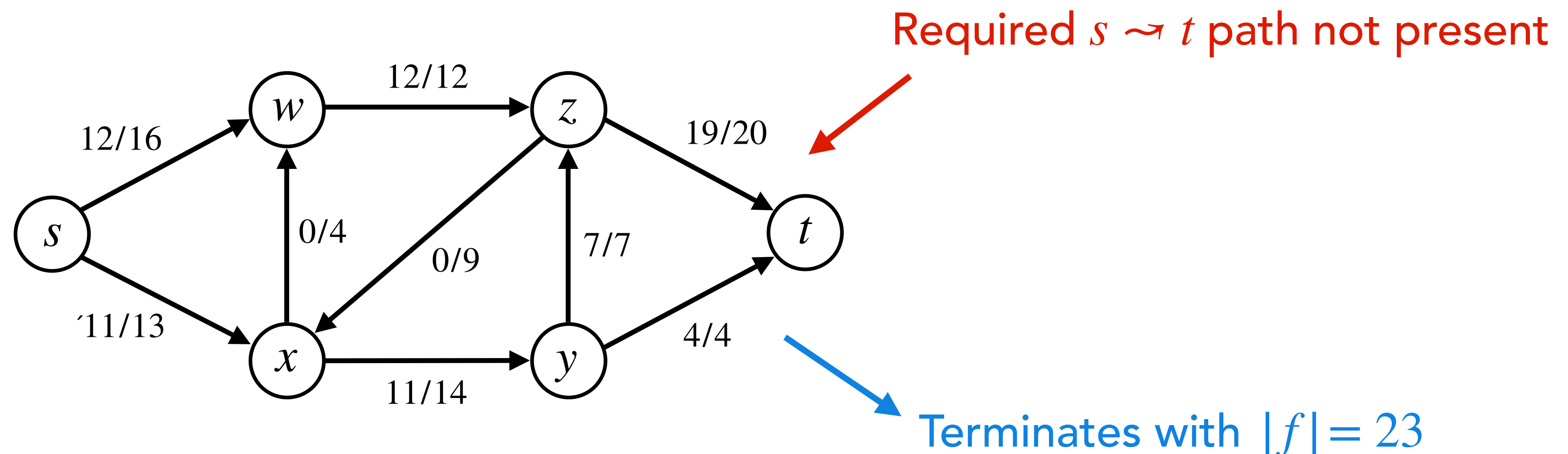


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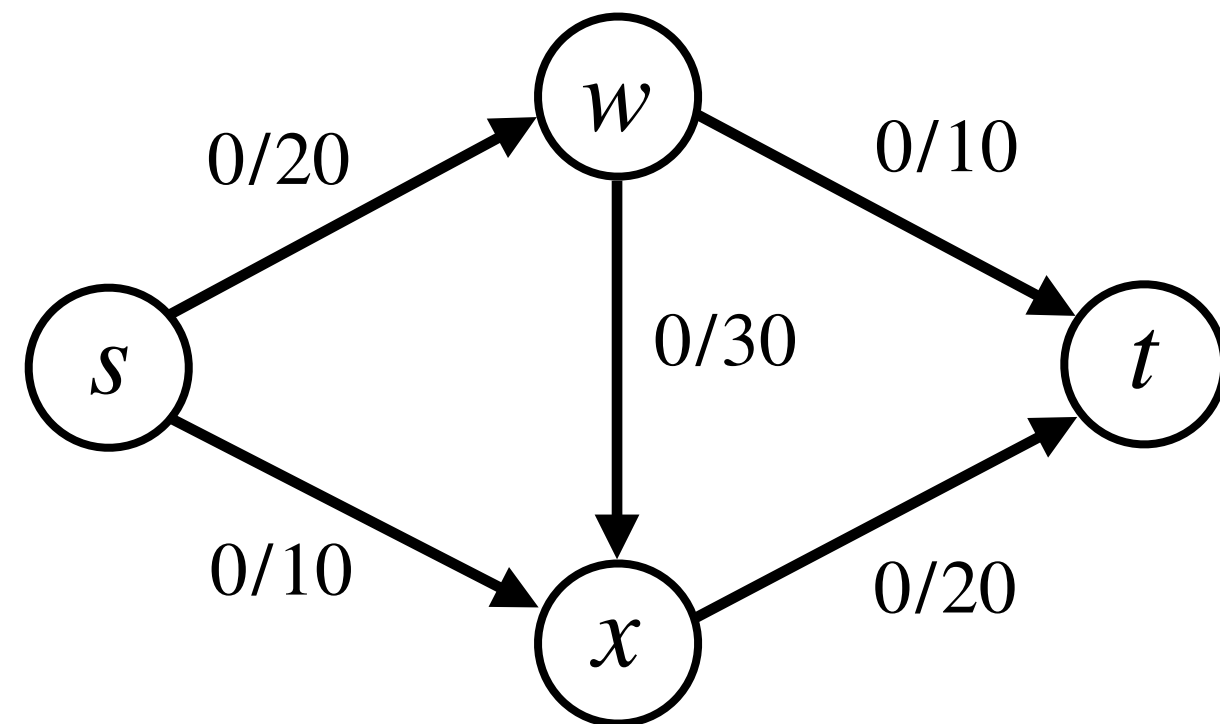
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Counter-Example:

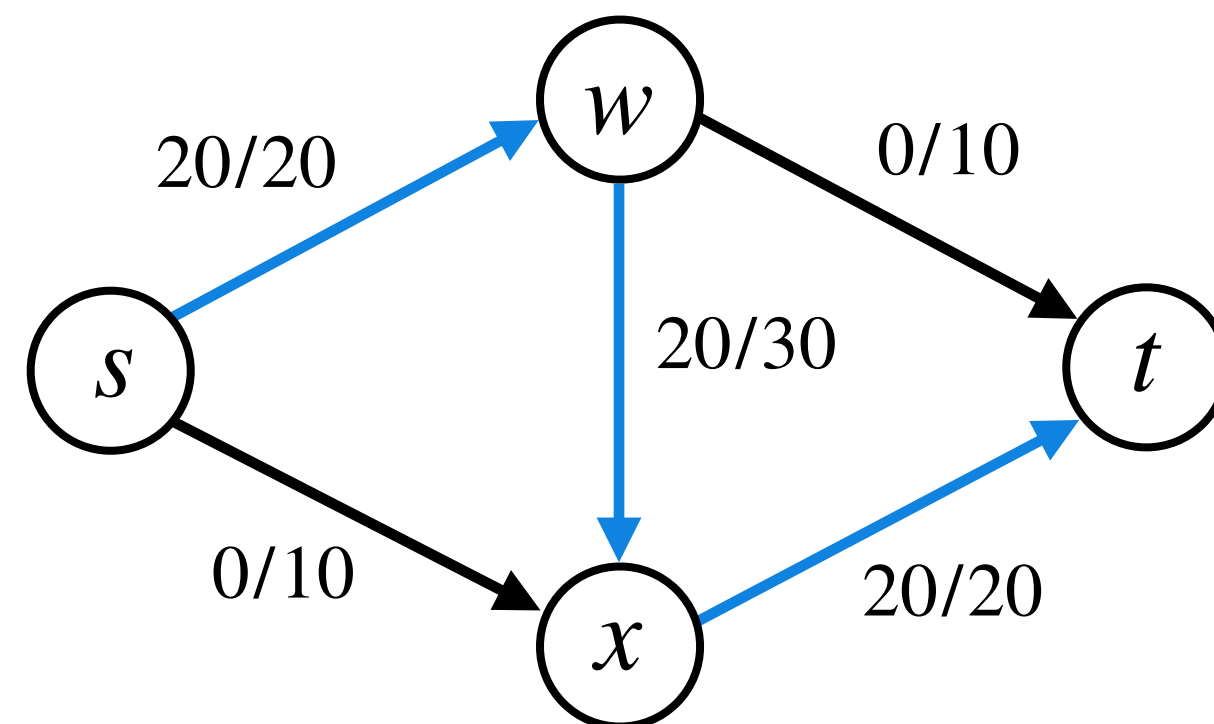


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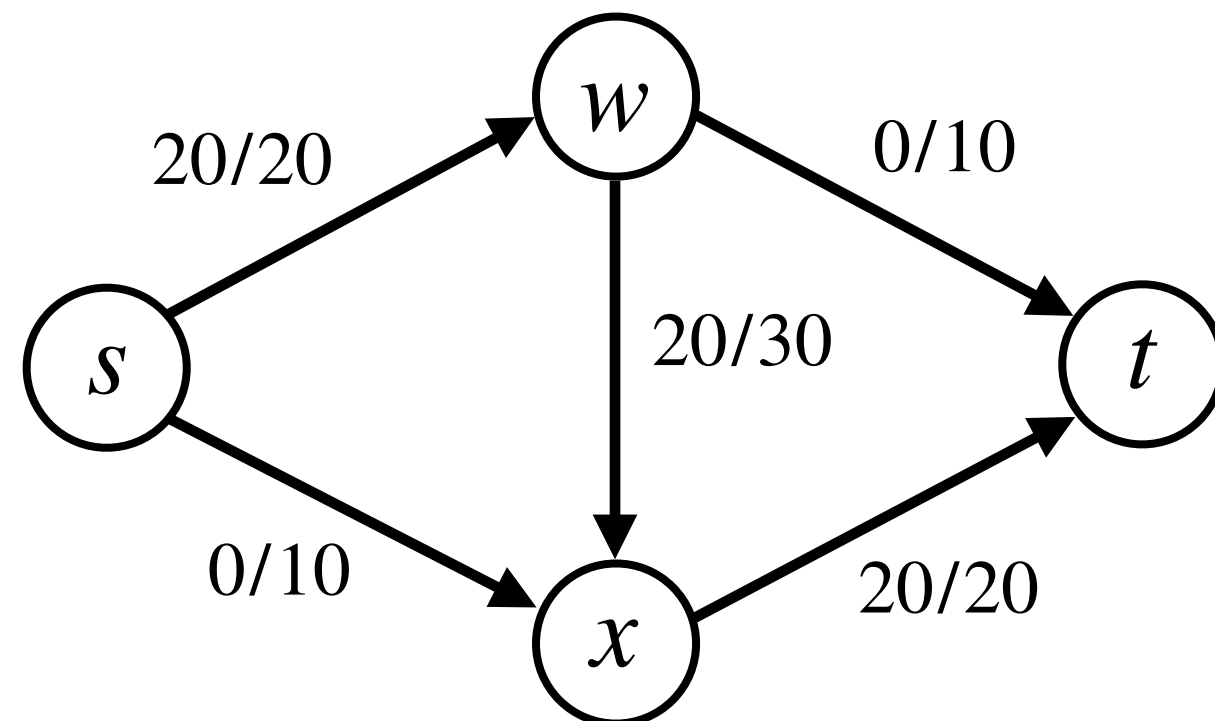


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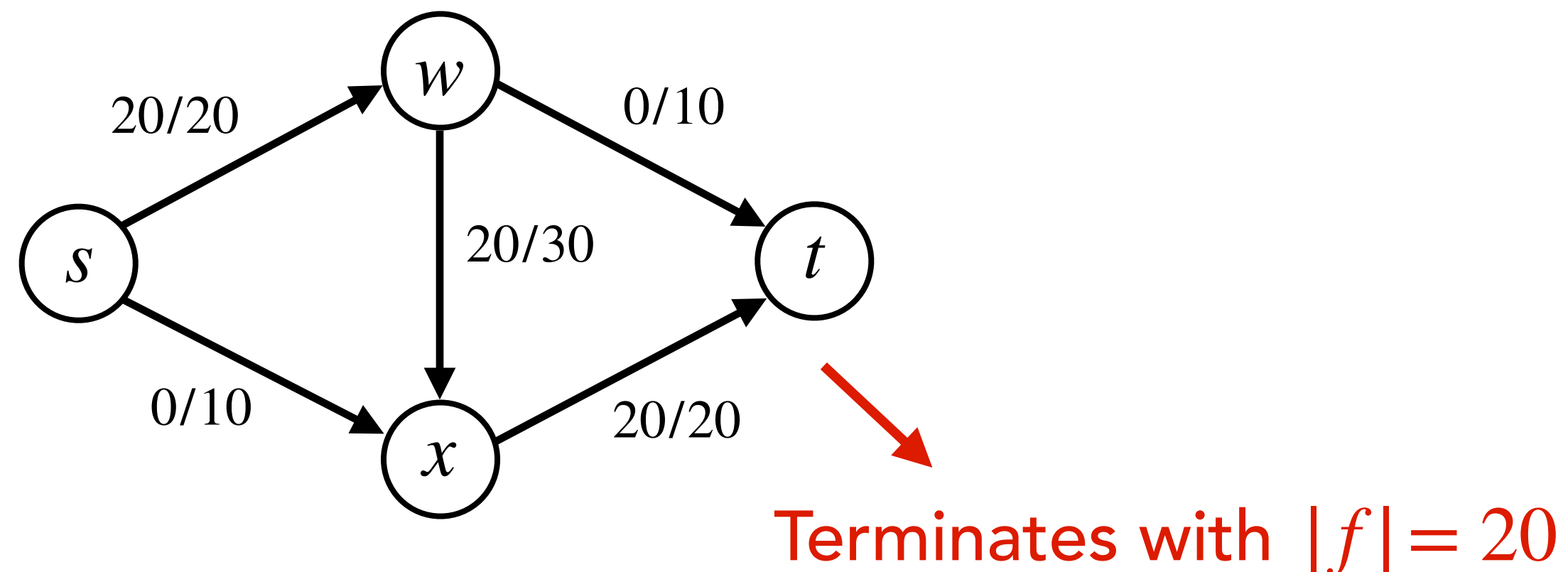


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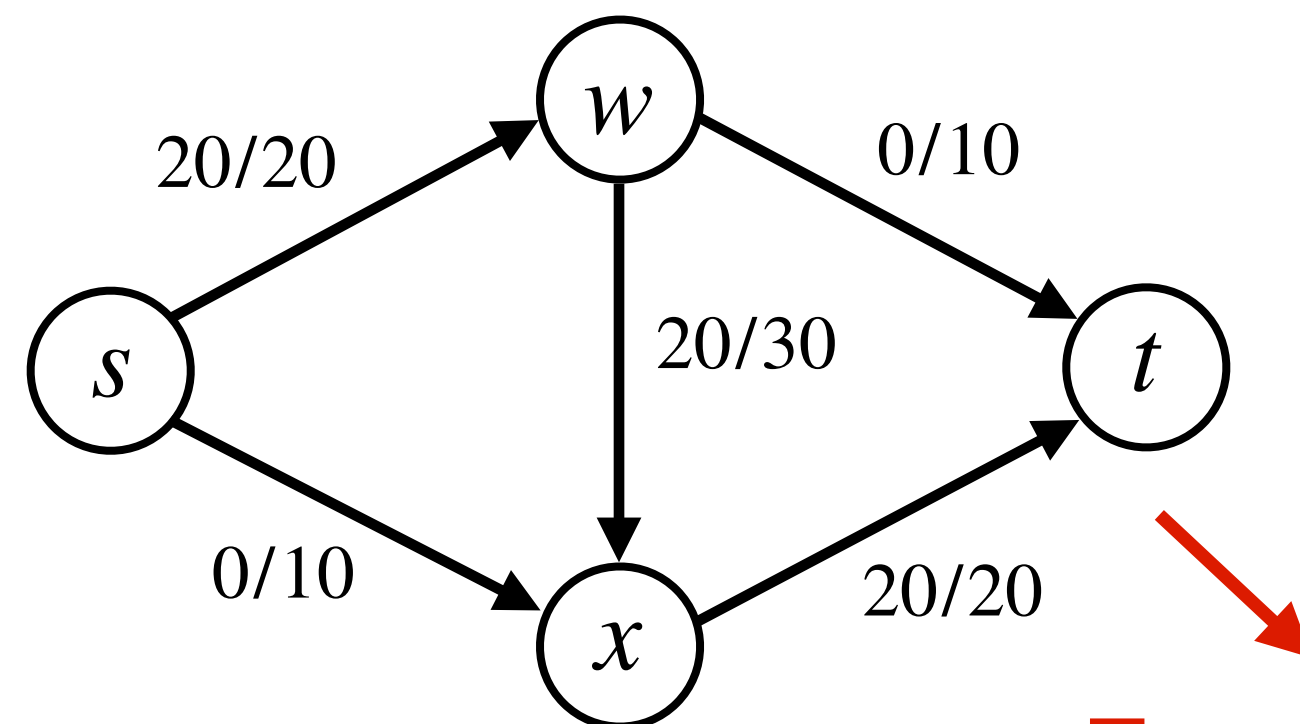


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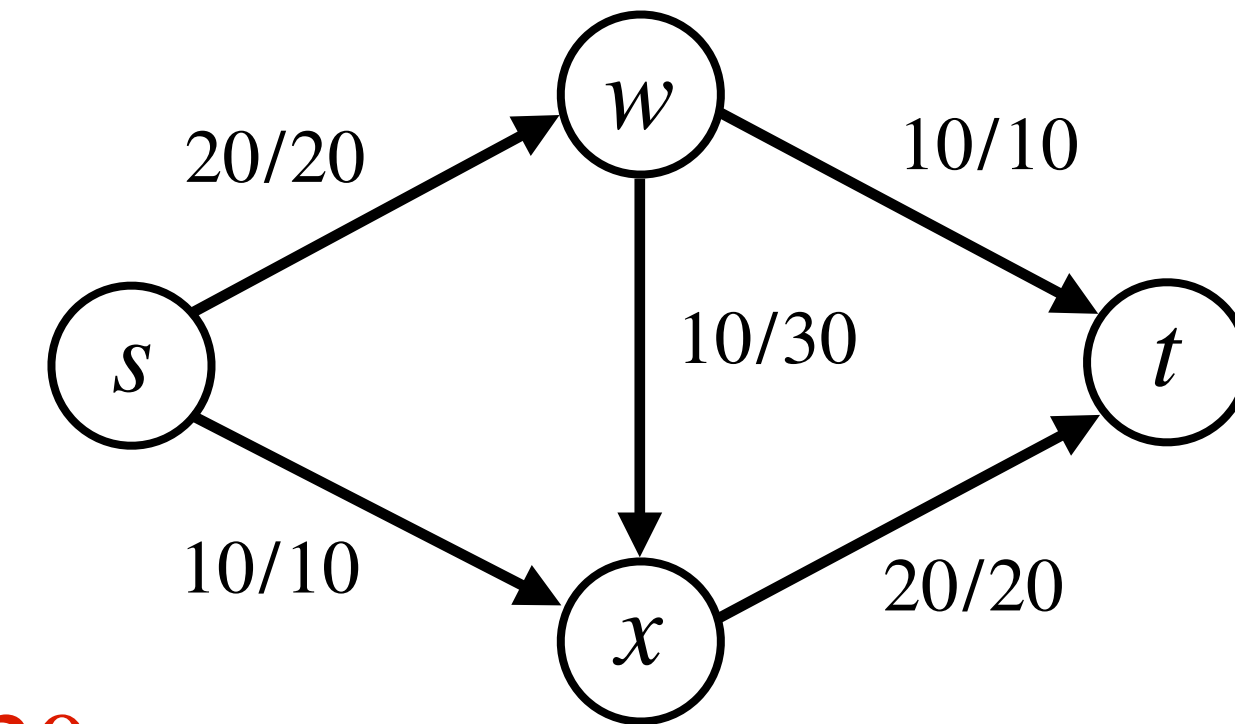
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Terminates with $|f| = 20$

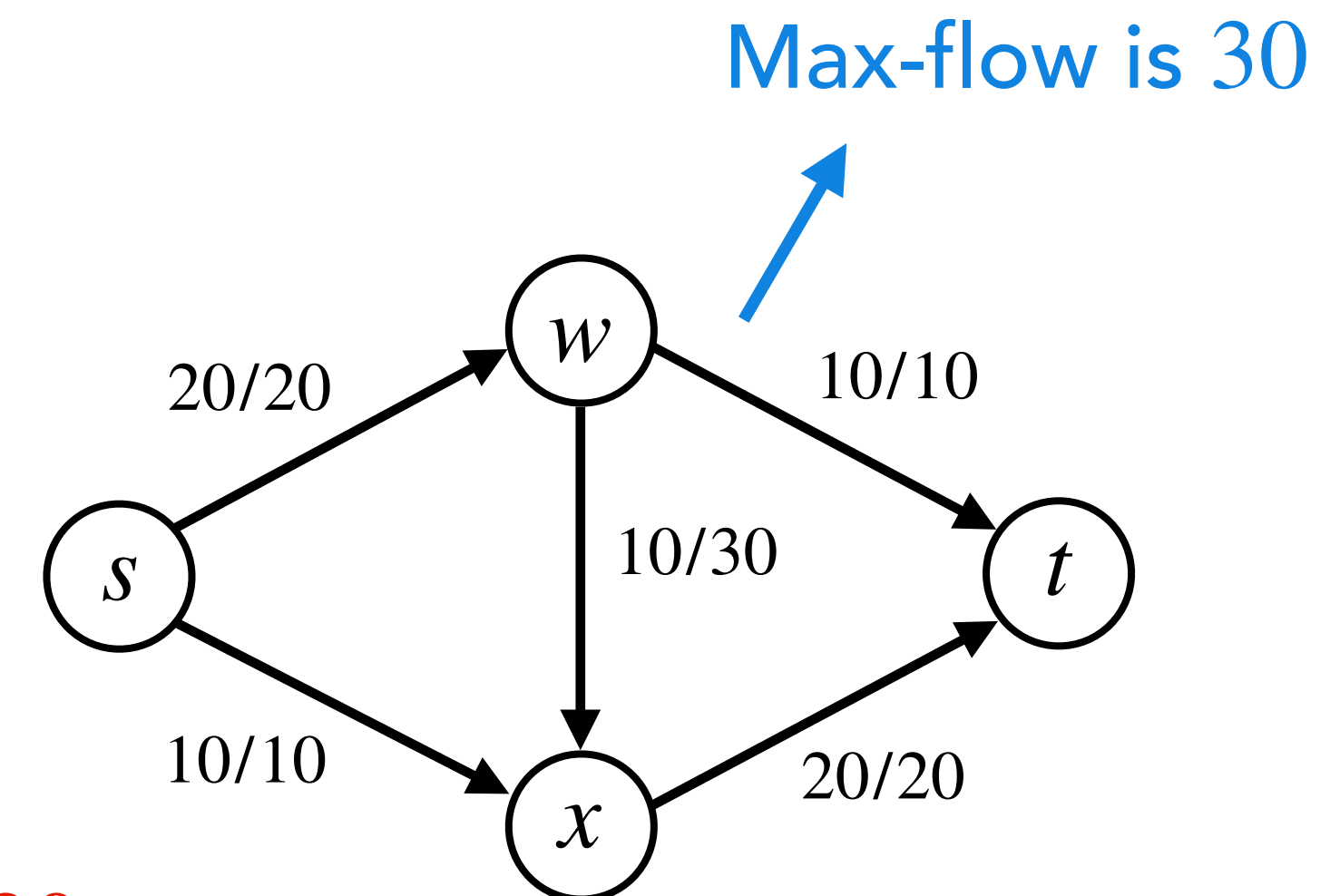
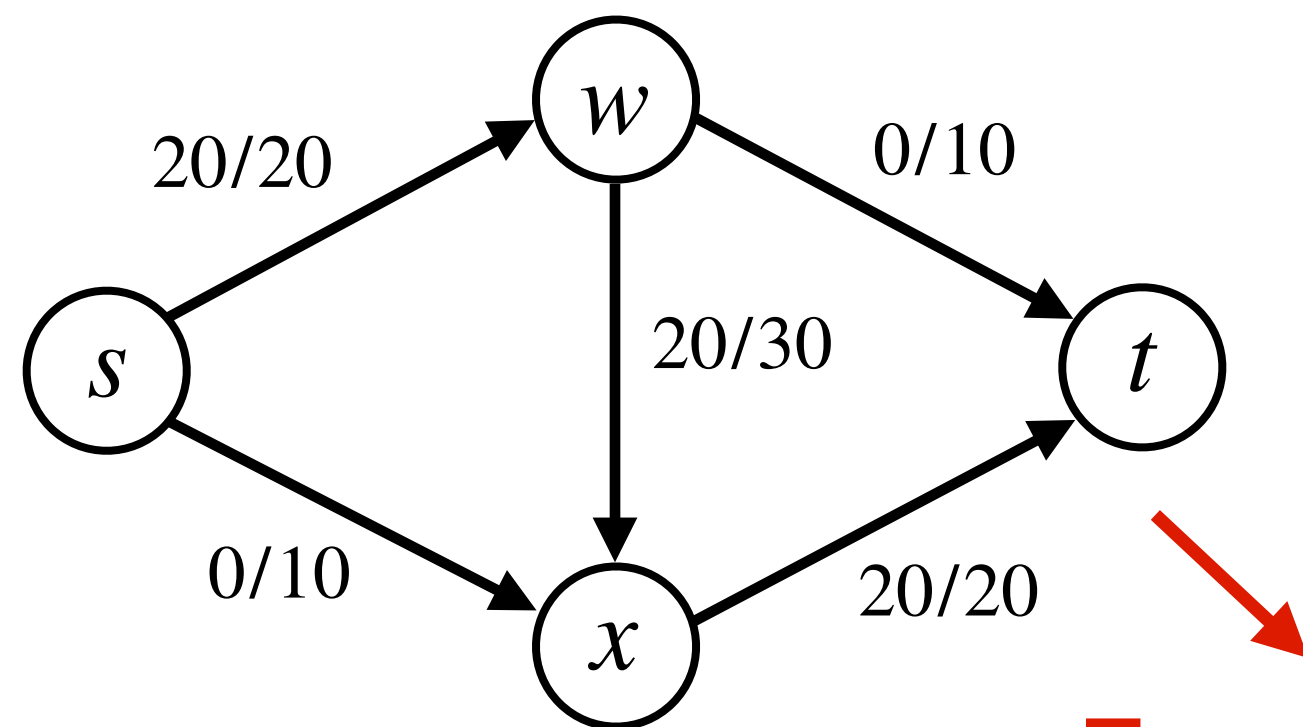


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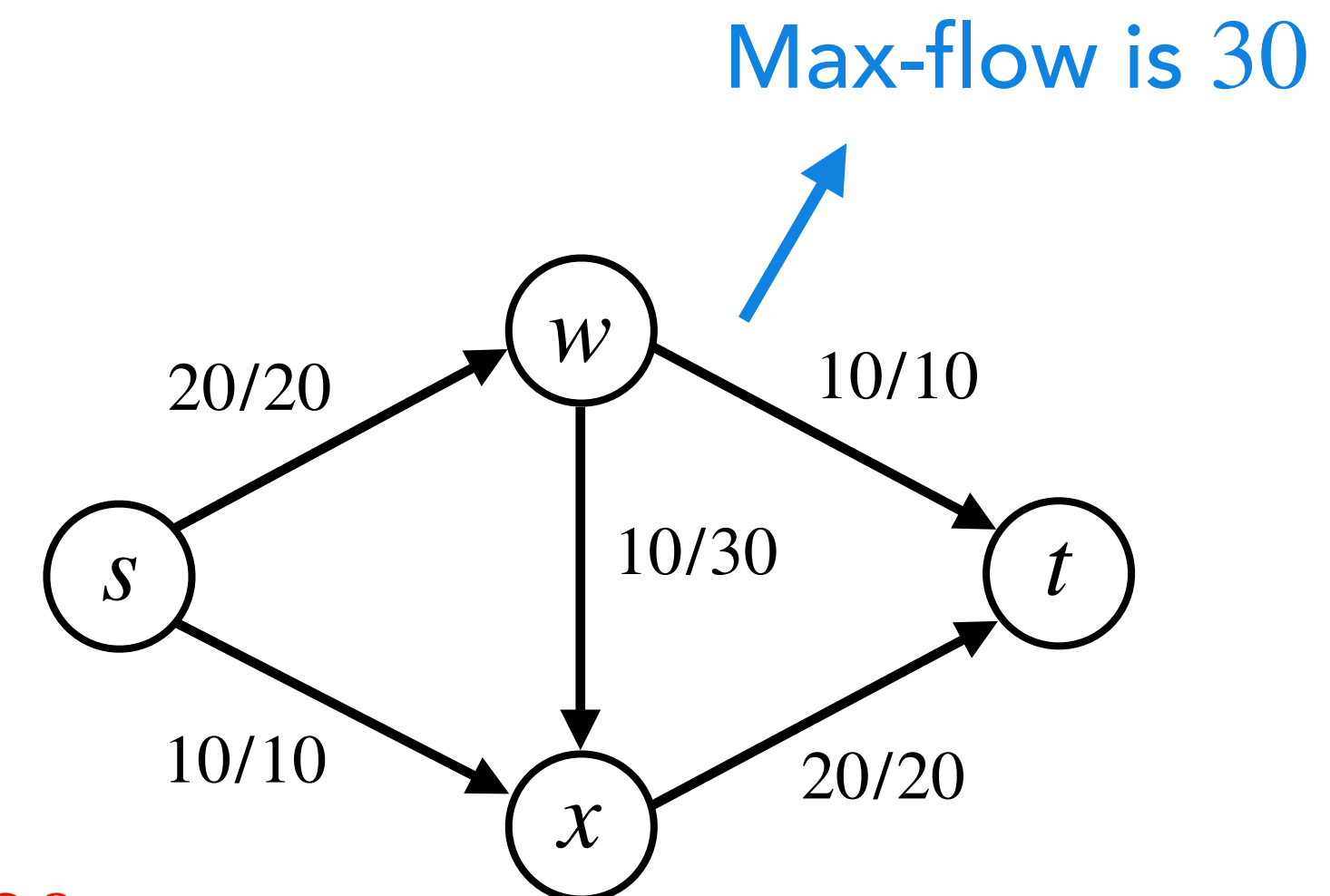
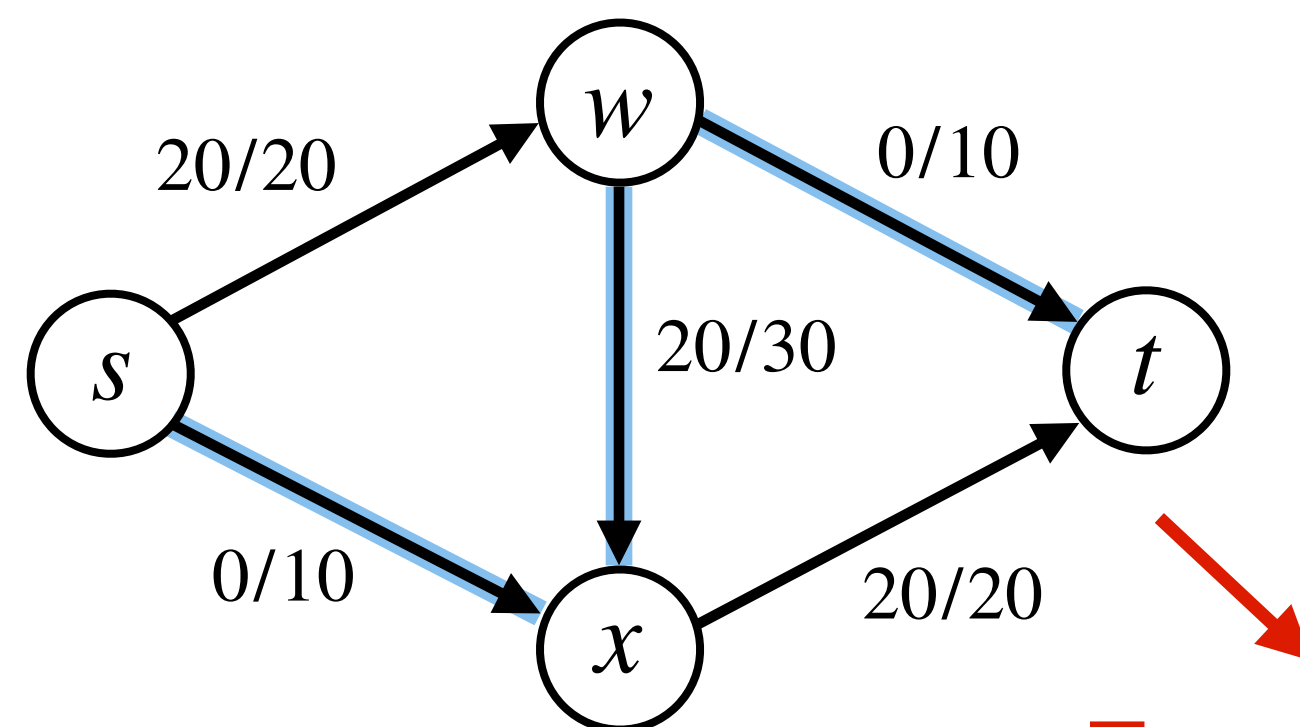


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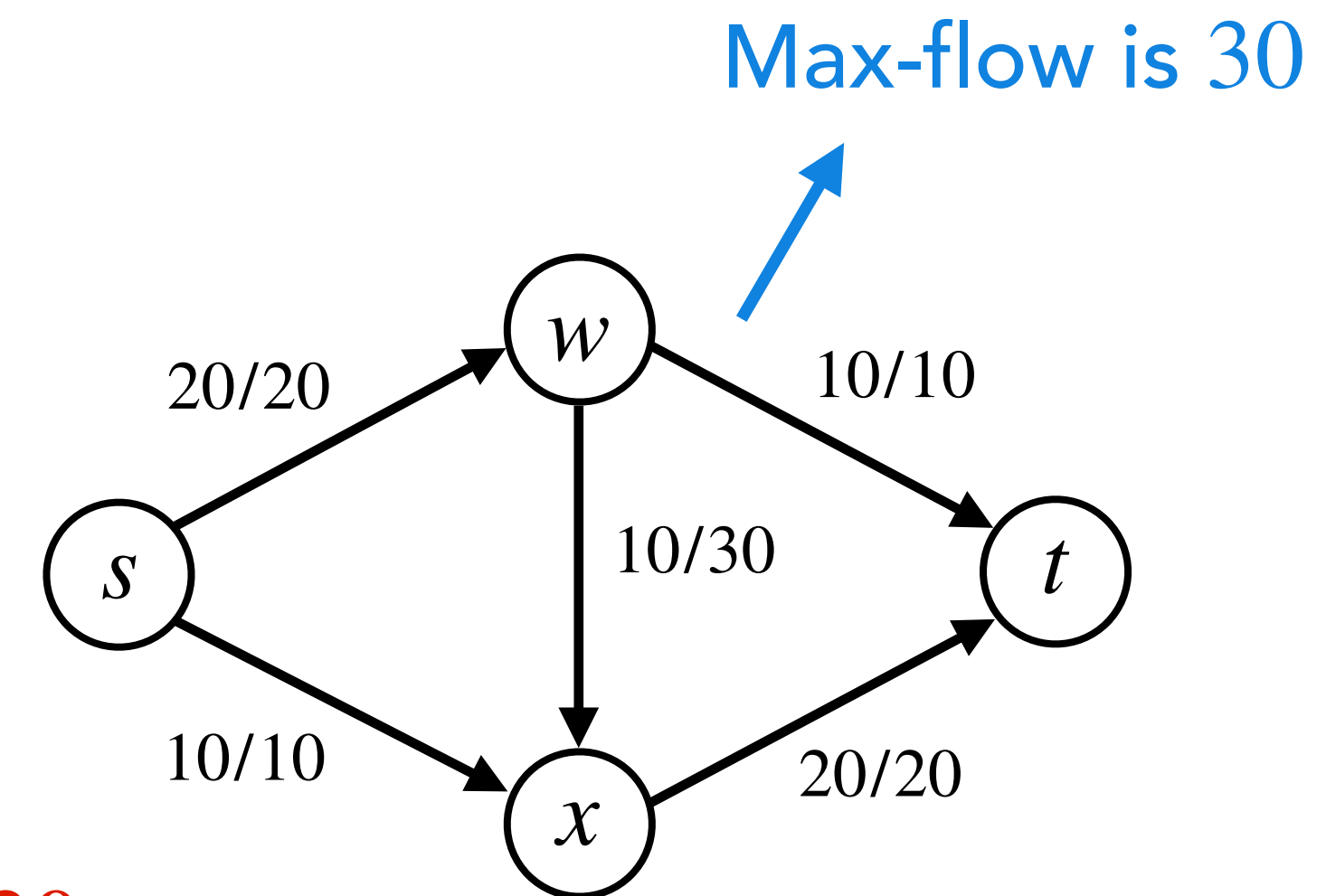
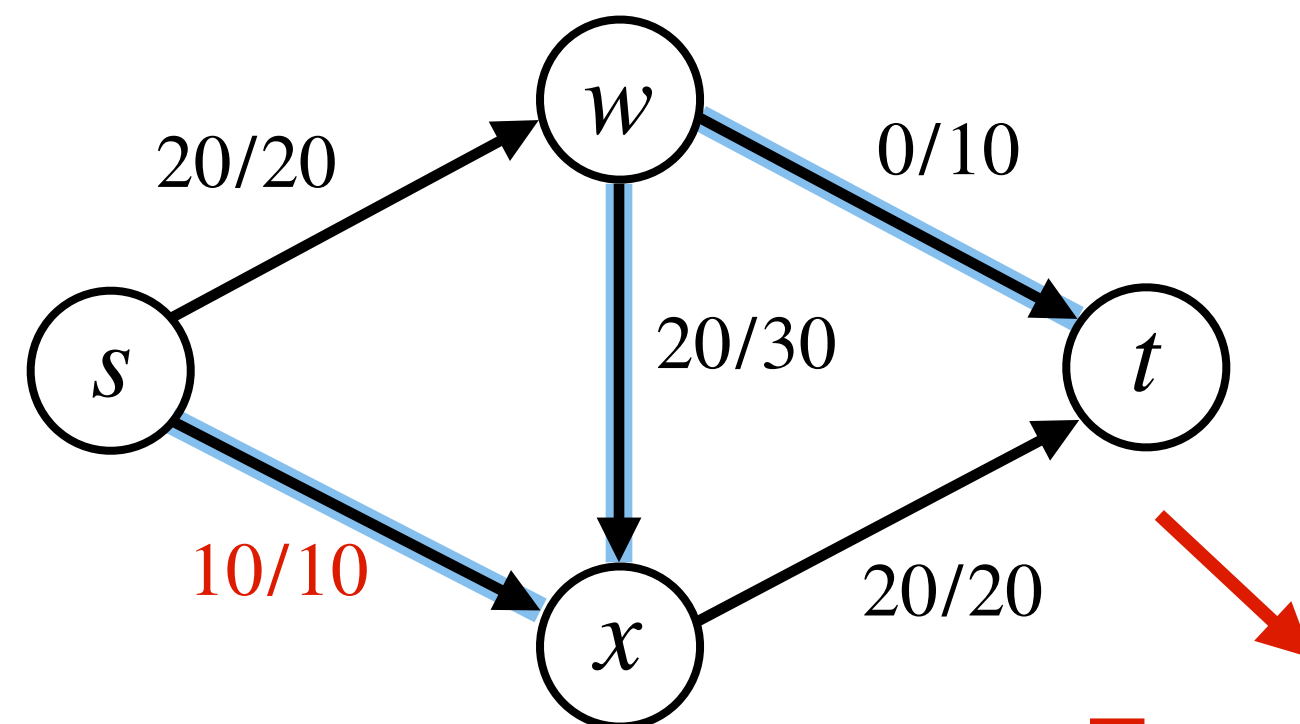


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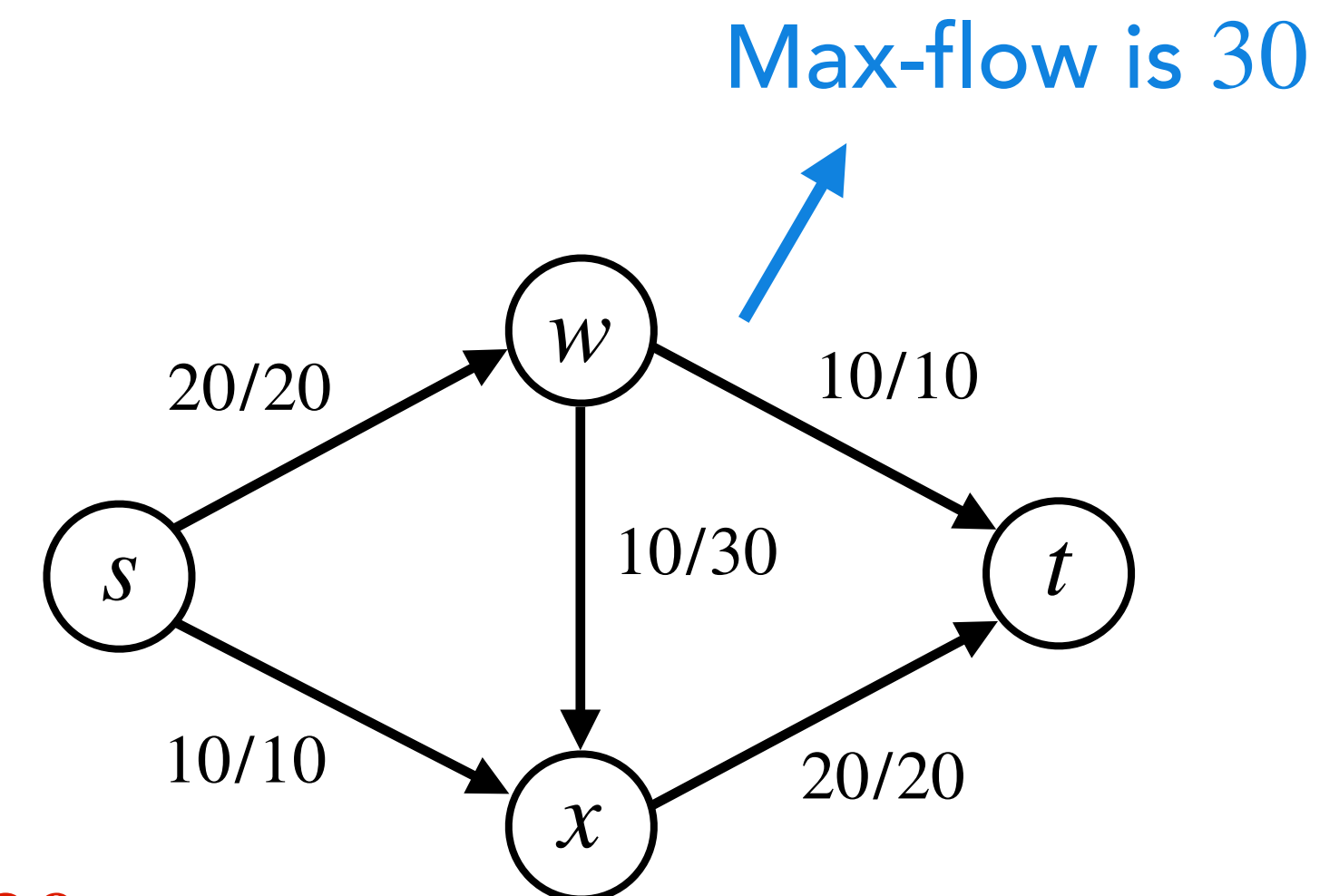
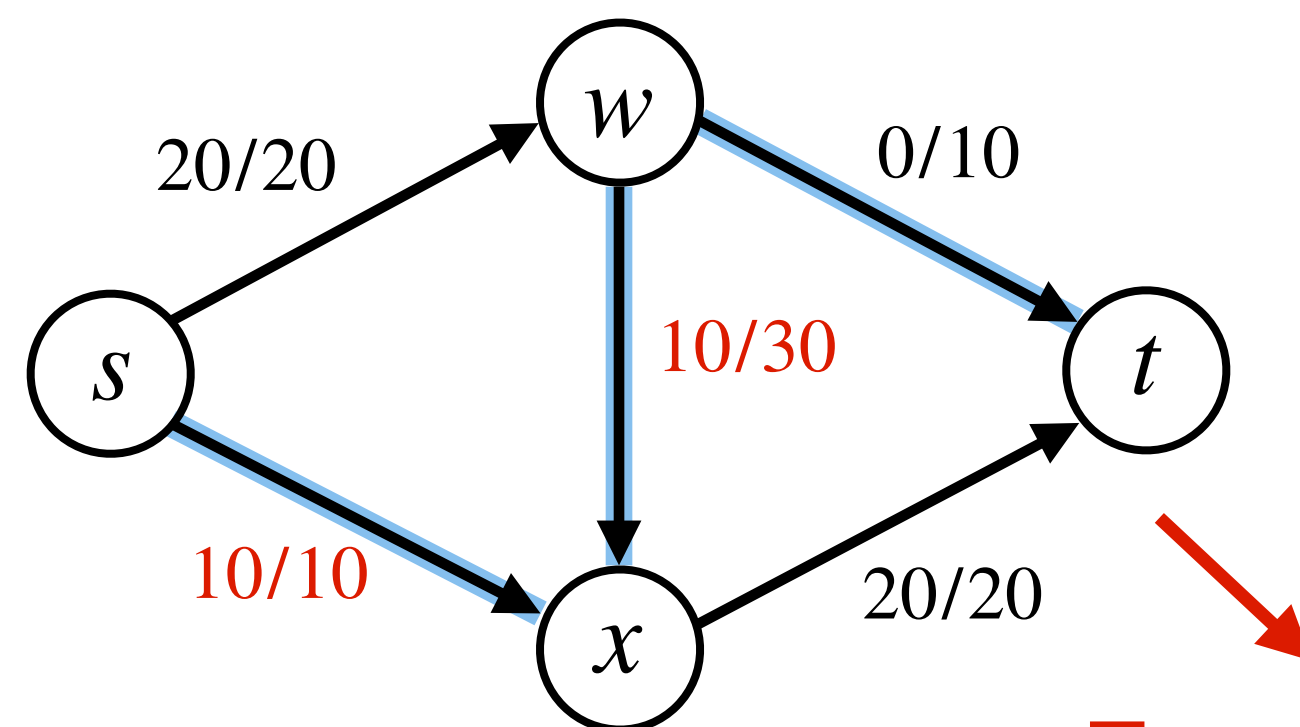


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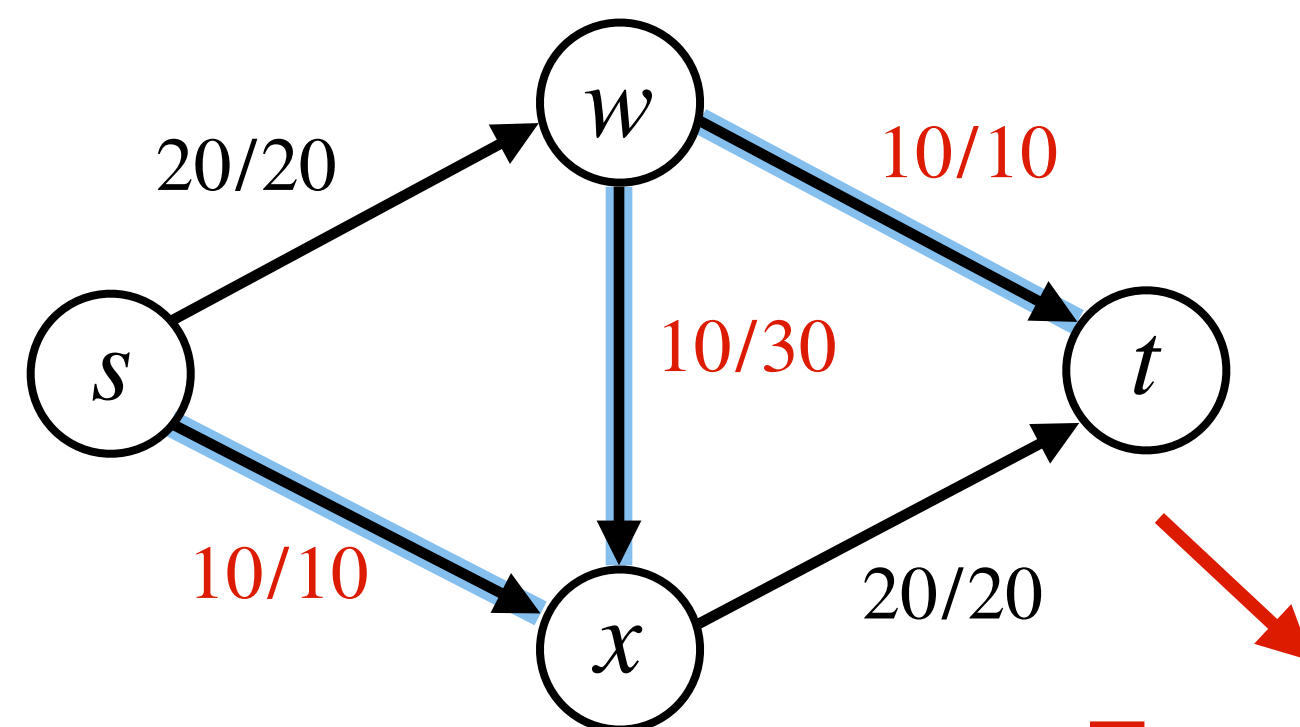


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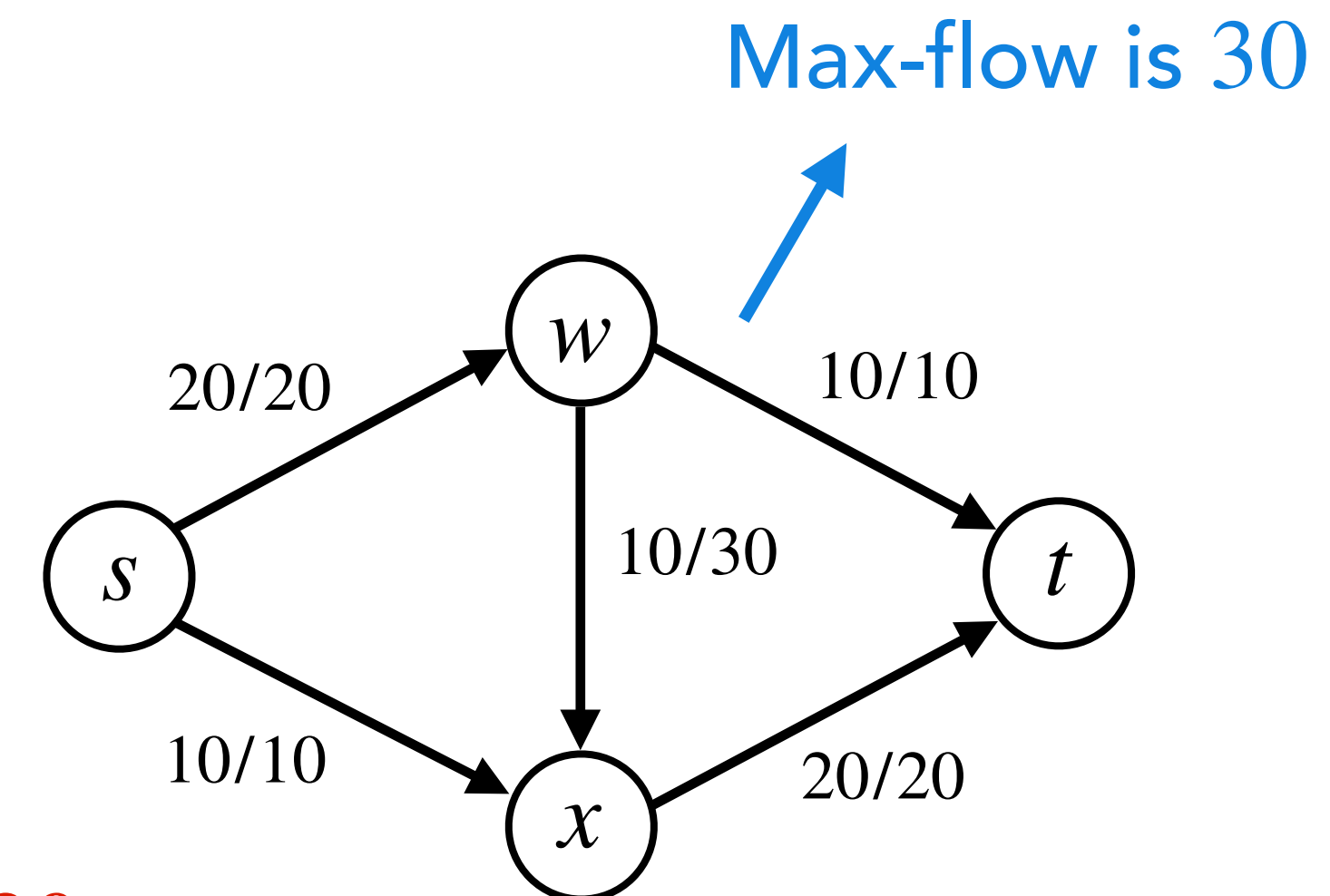
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Need to learn a new structure for that!



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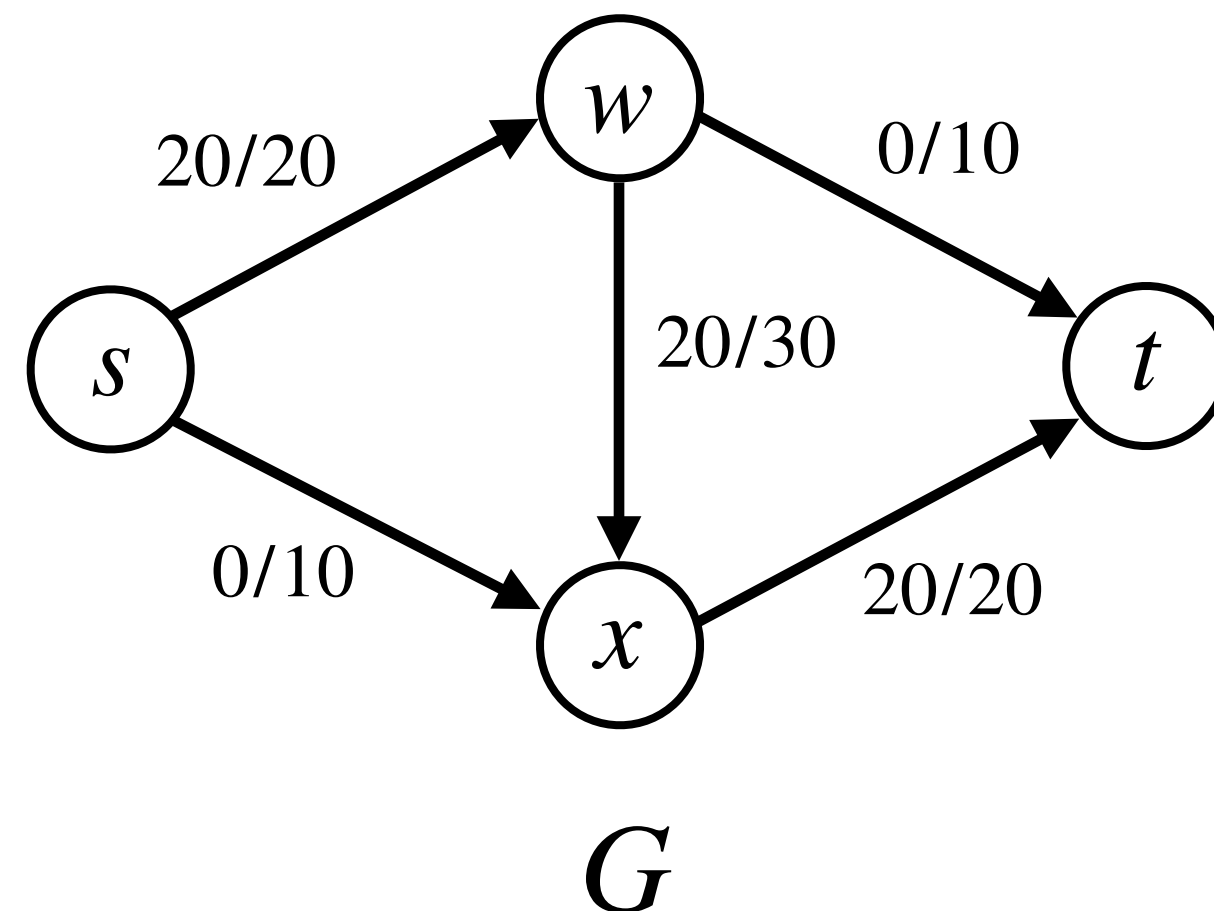
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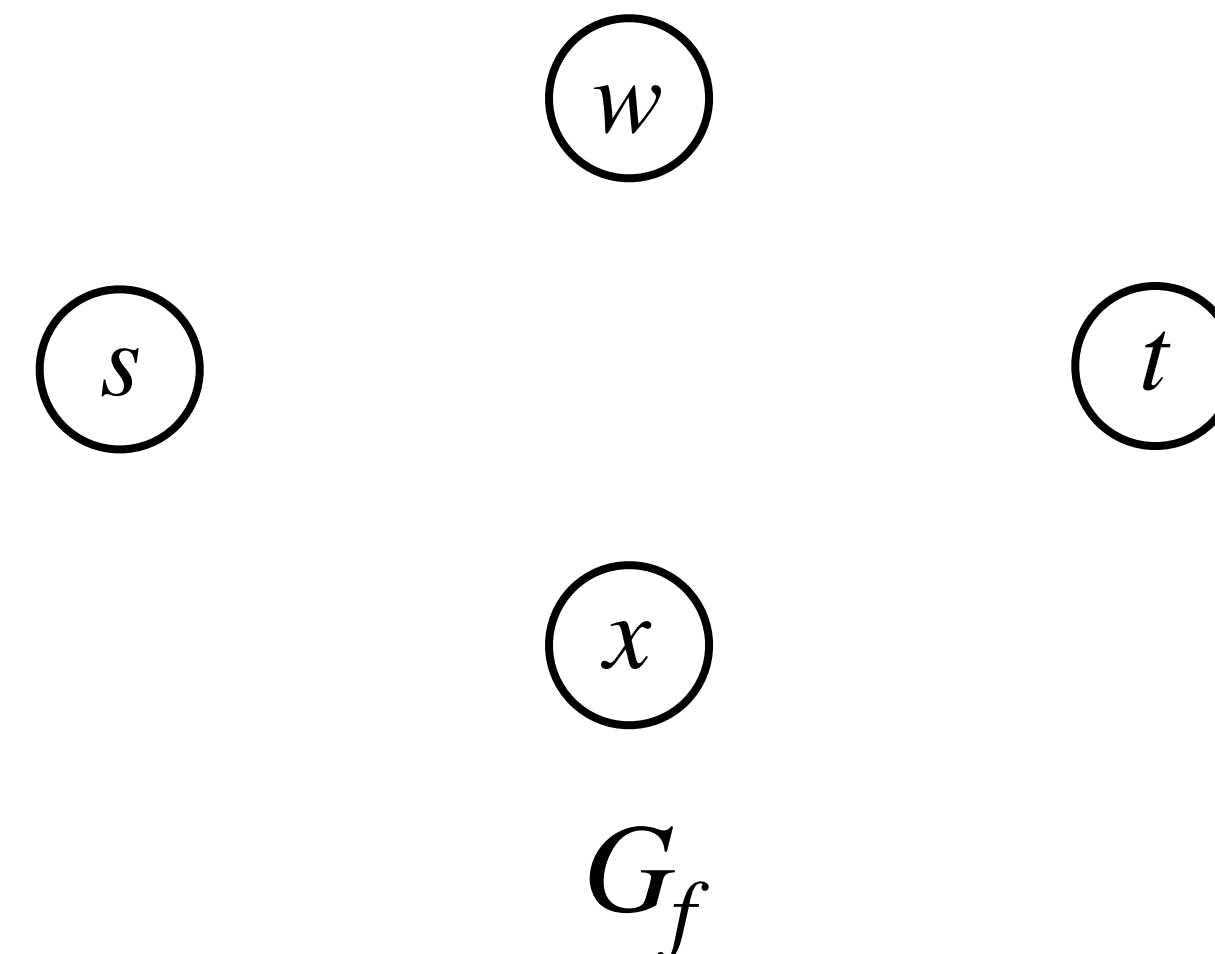
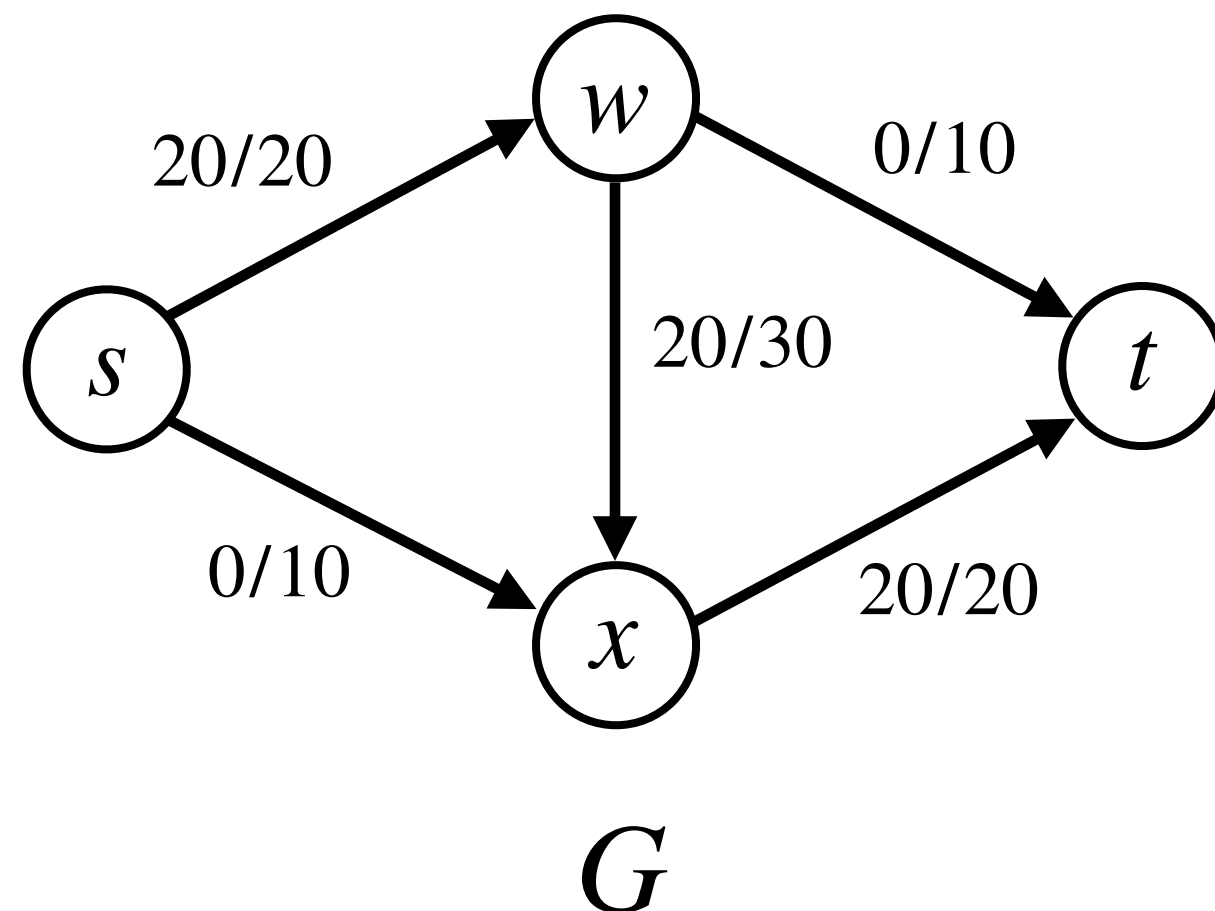


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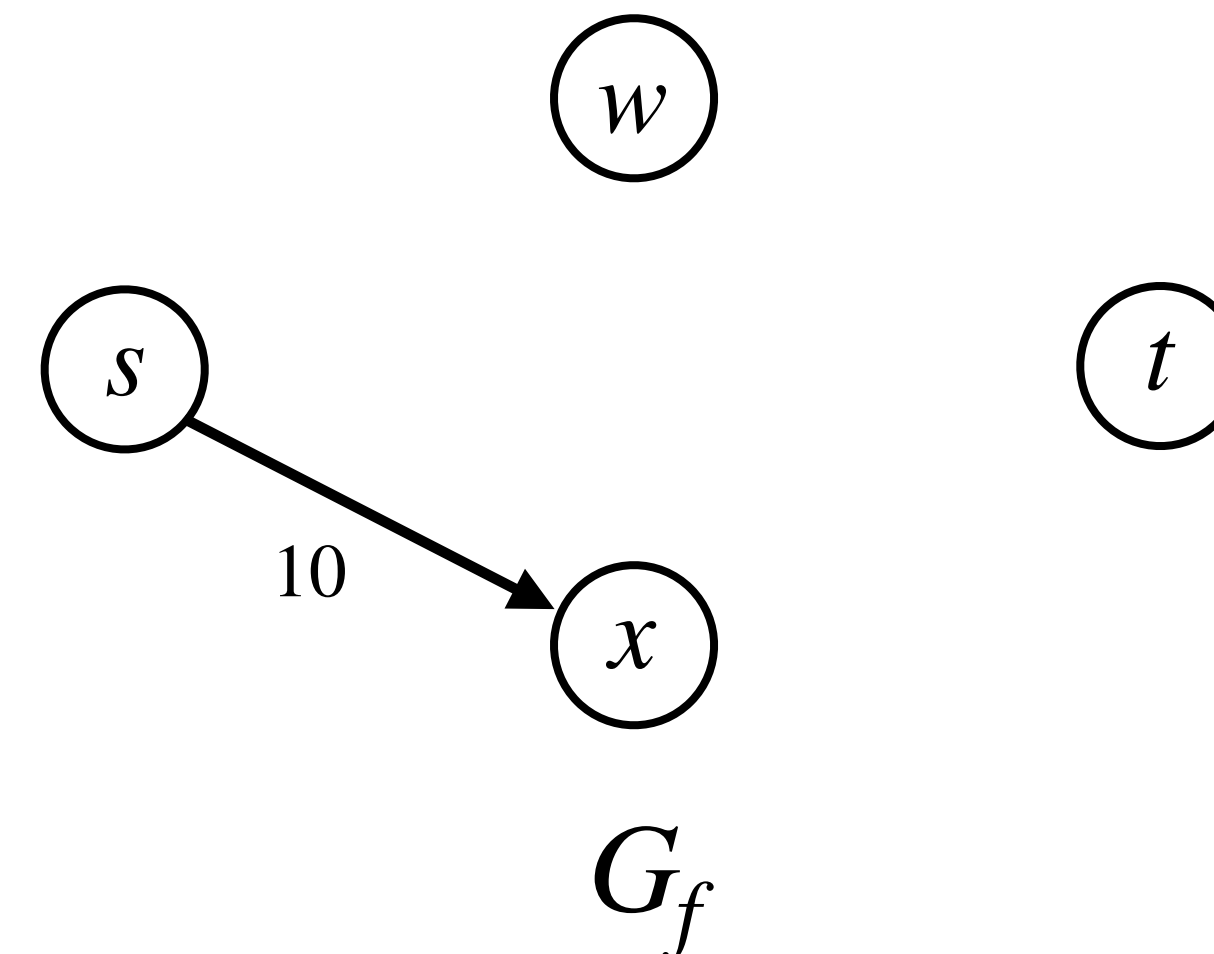
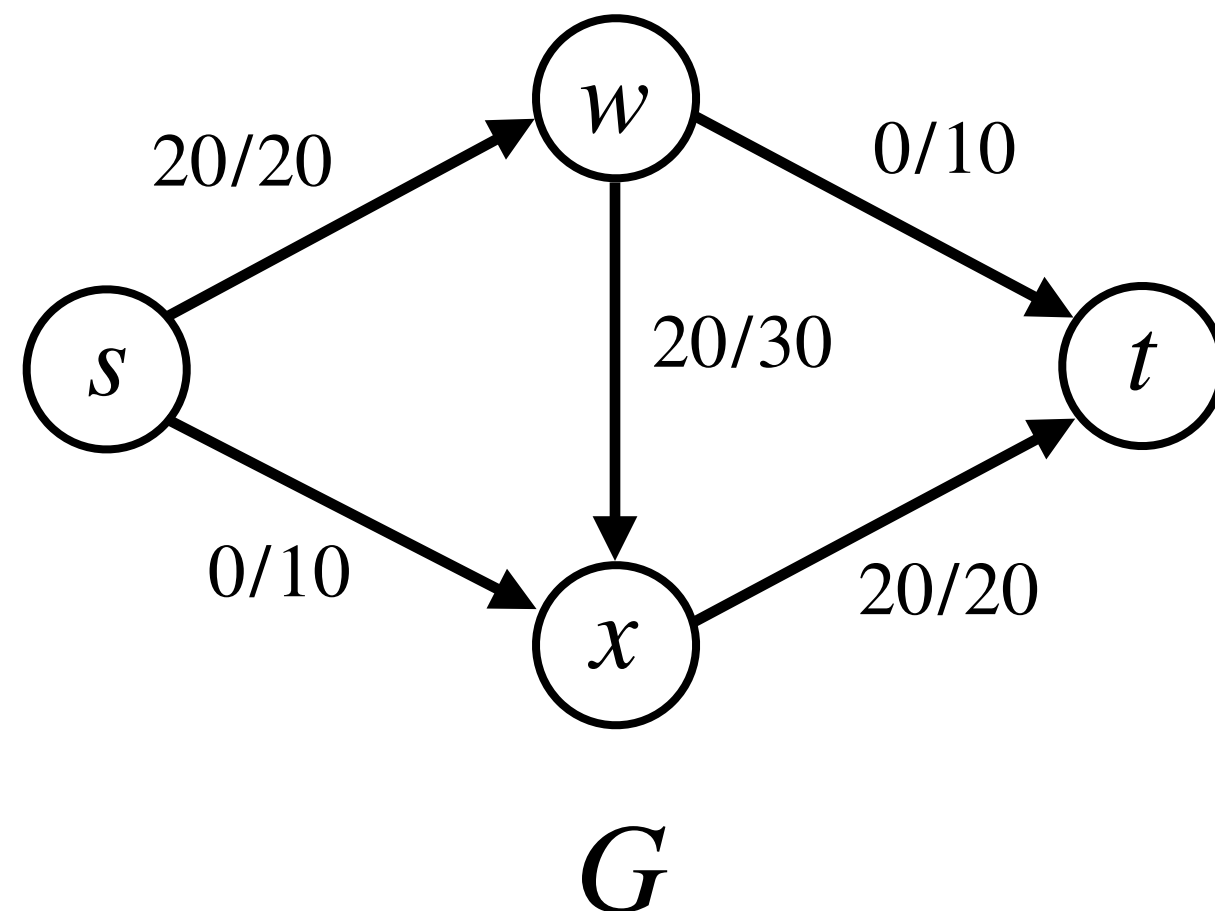


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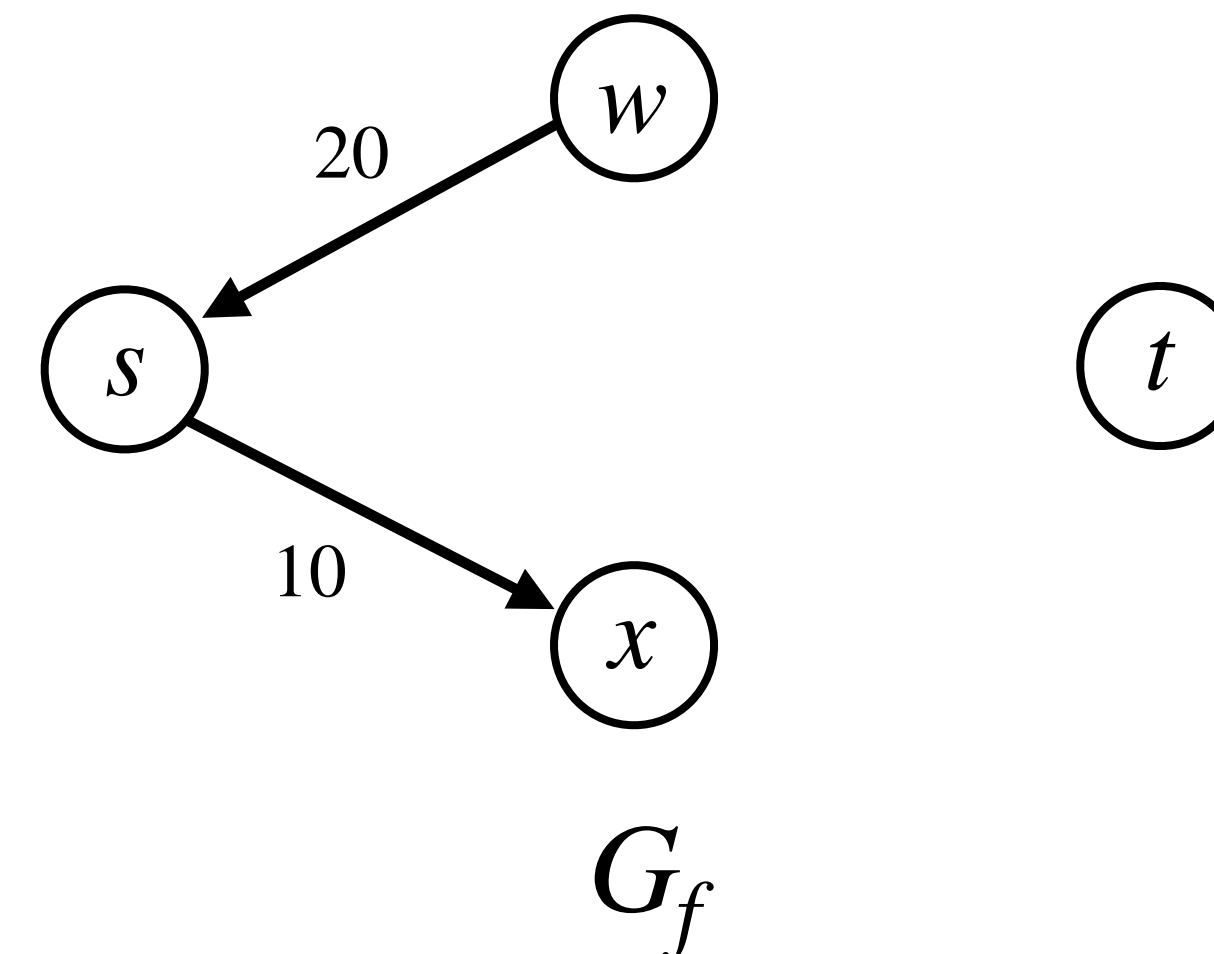
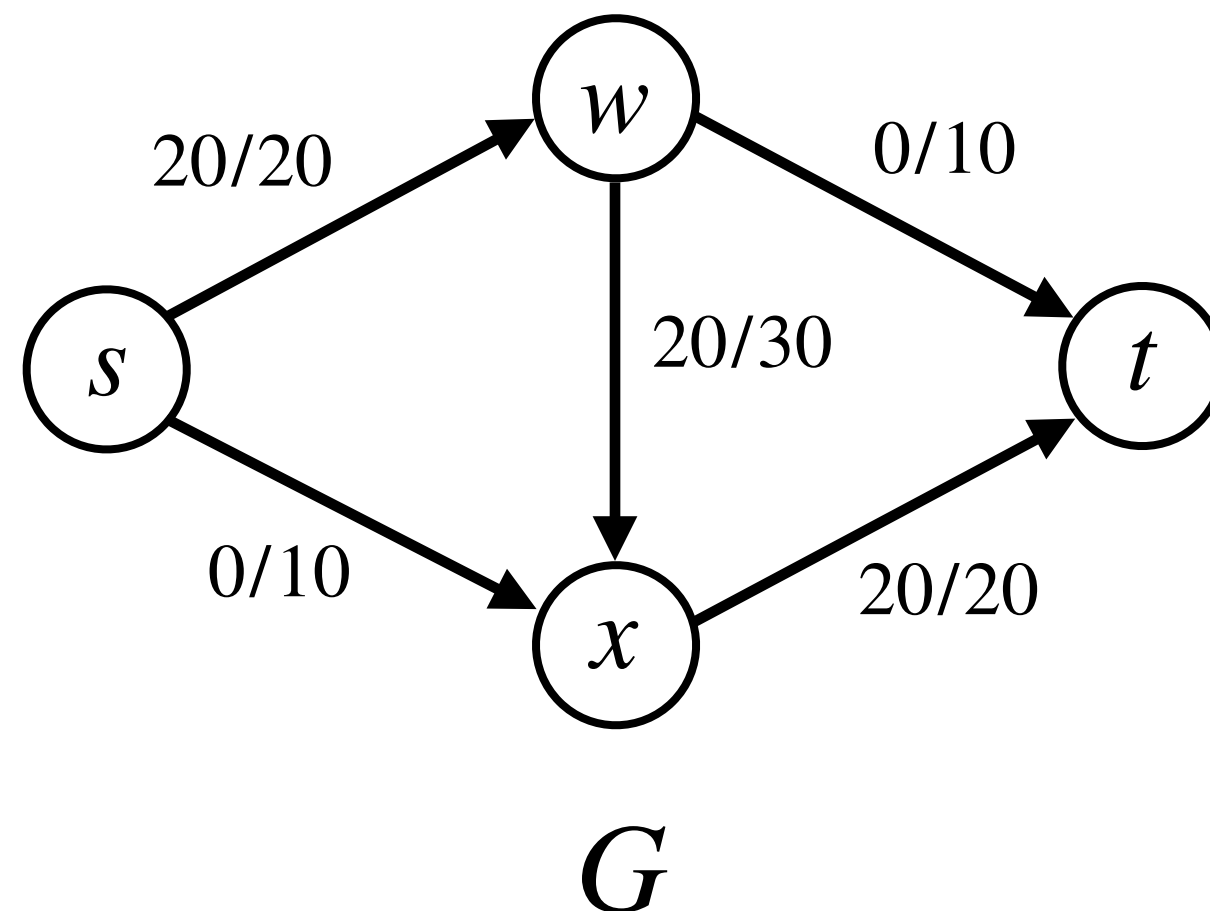


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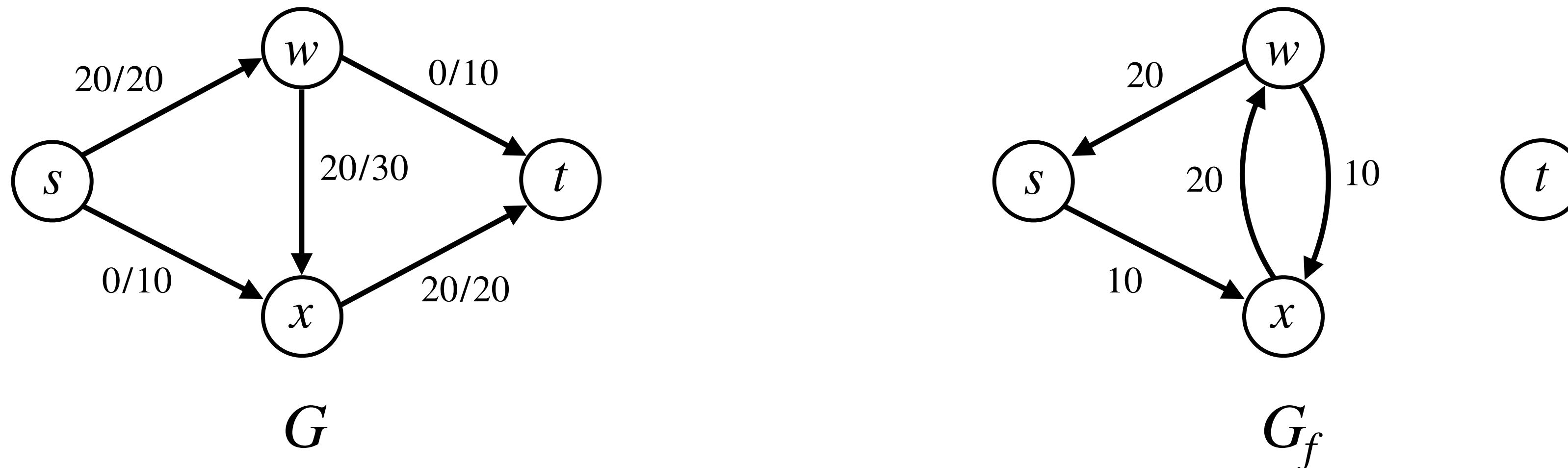


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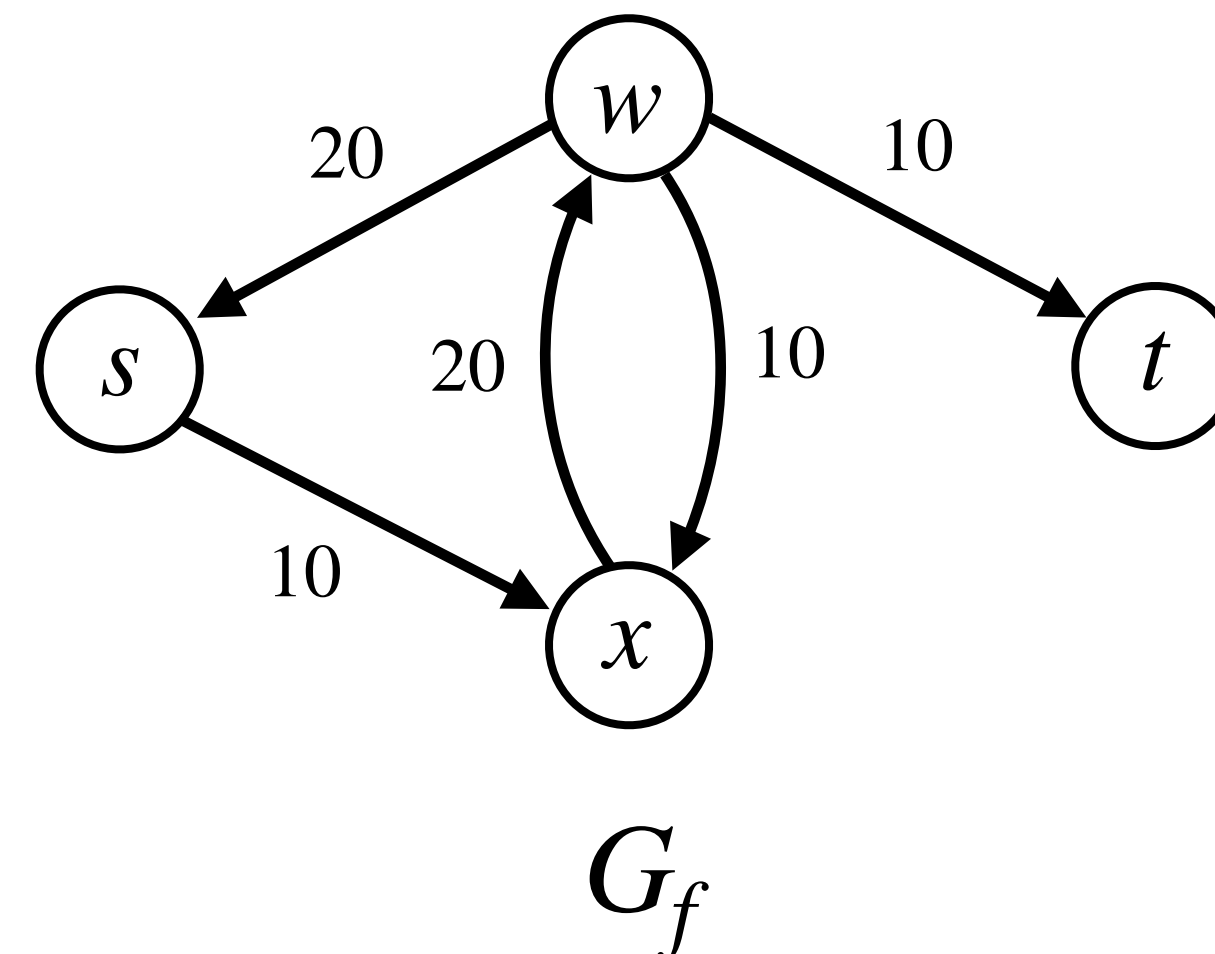
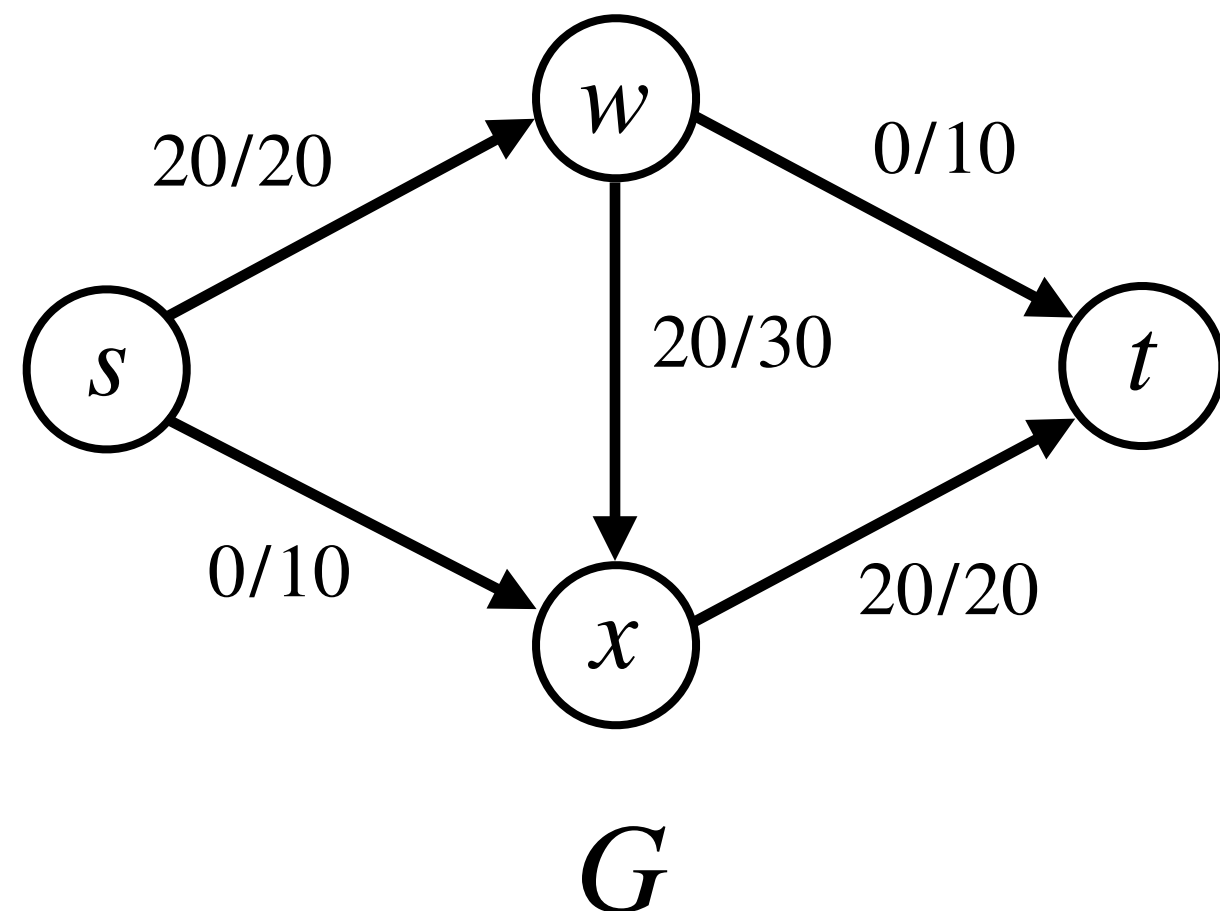


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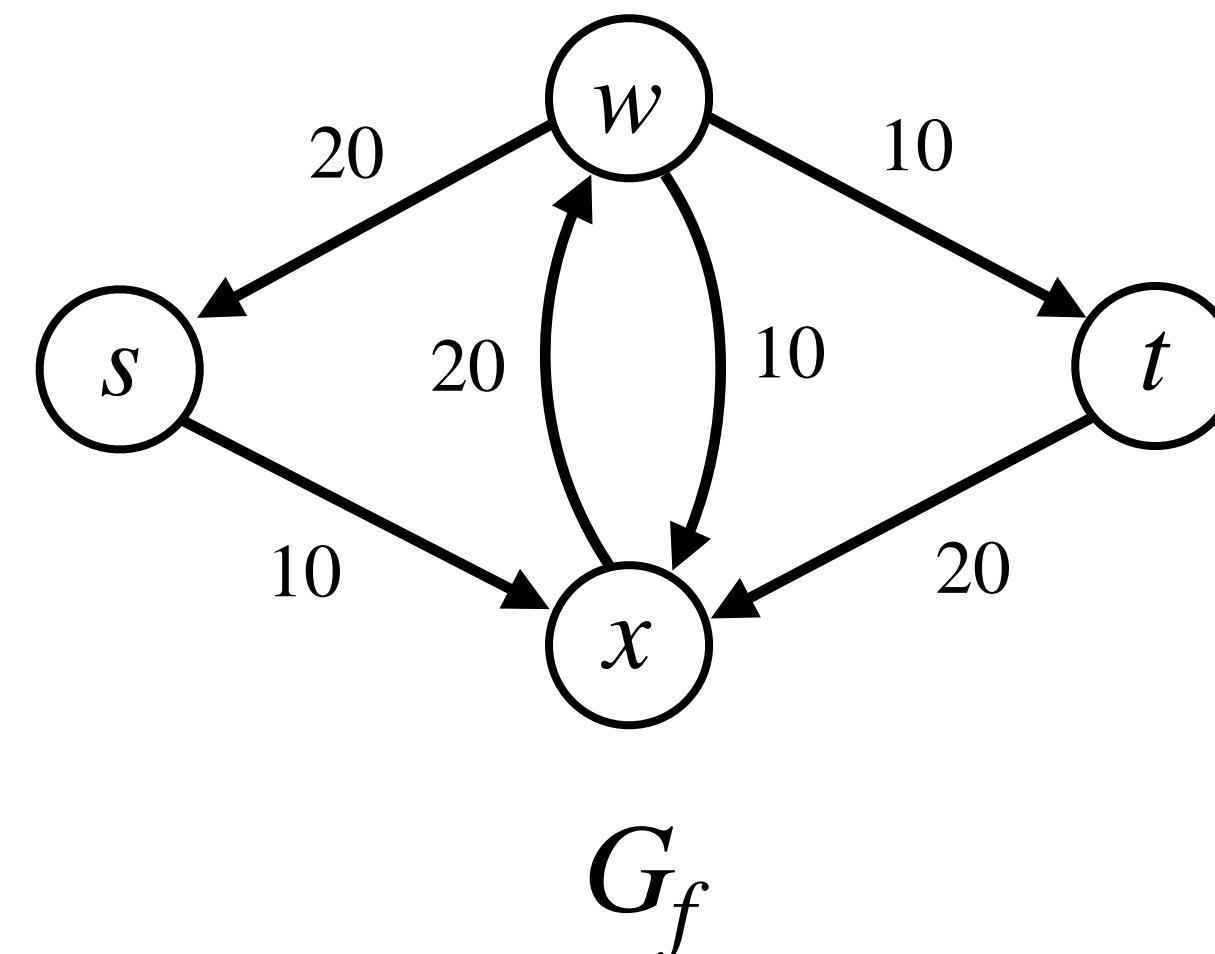
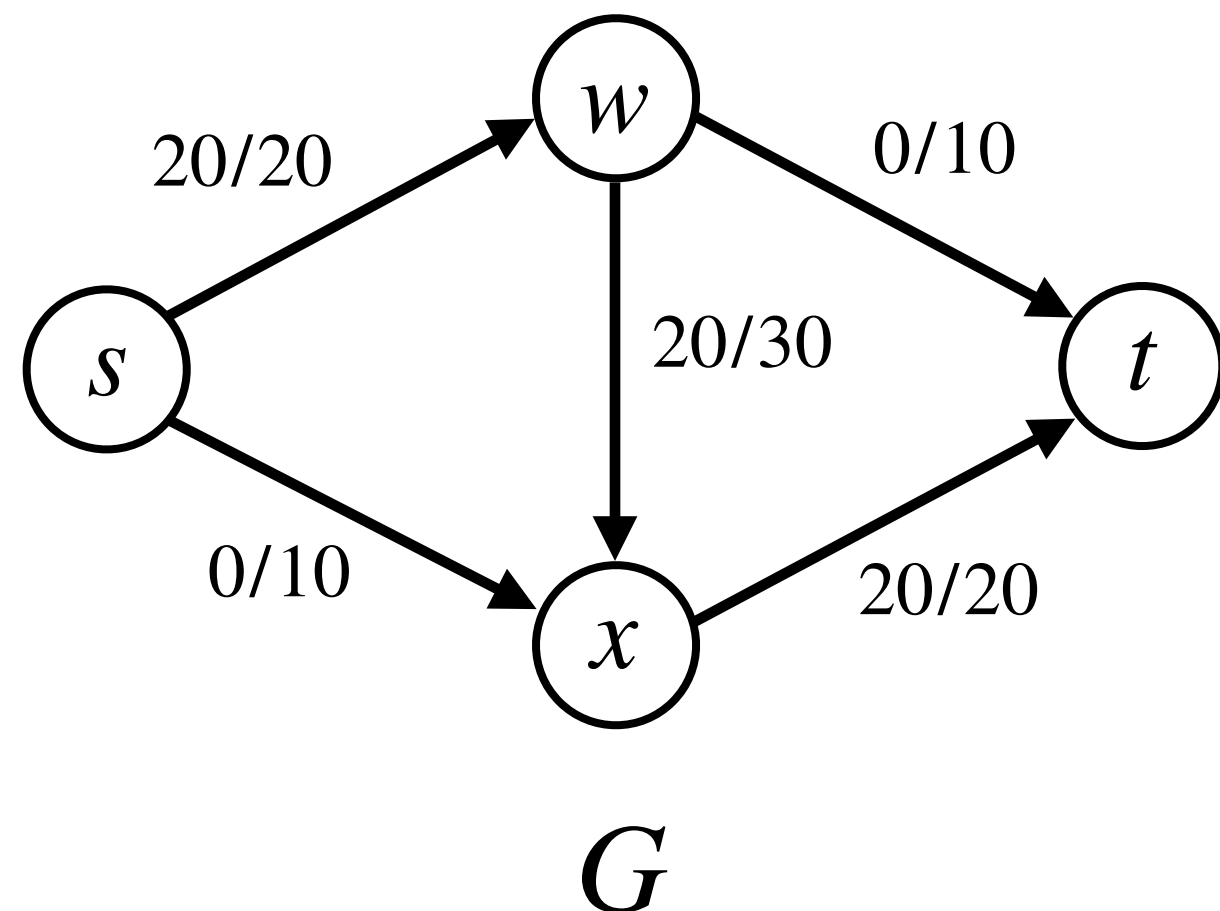


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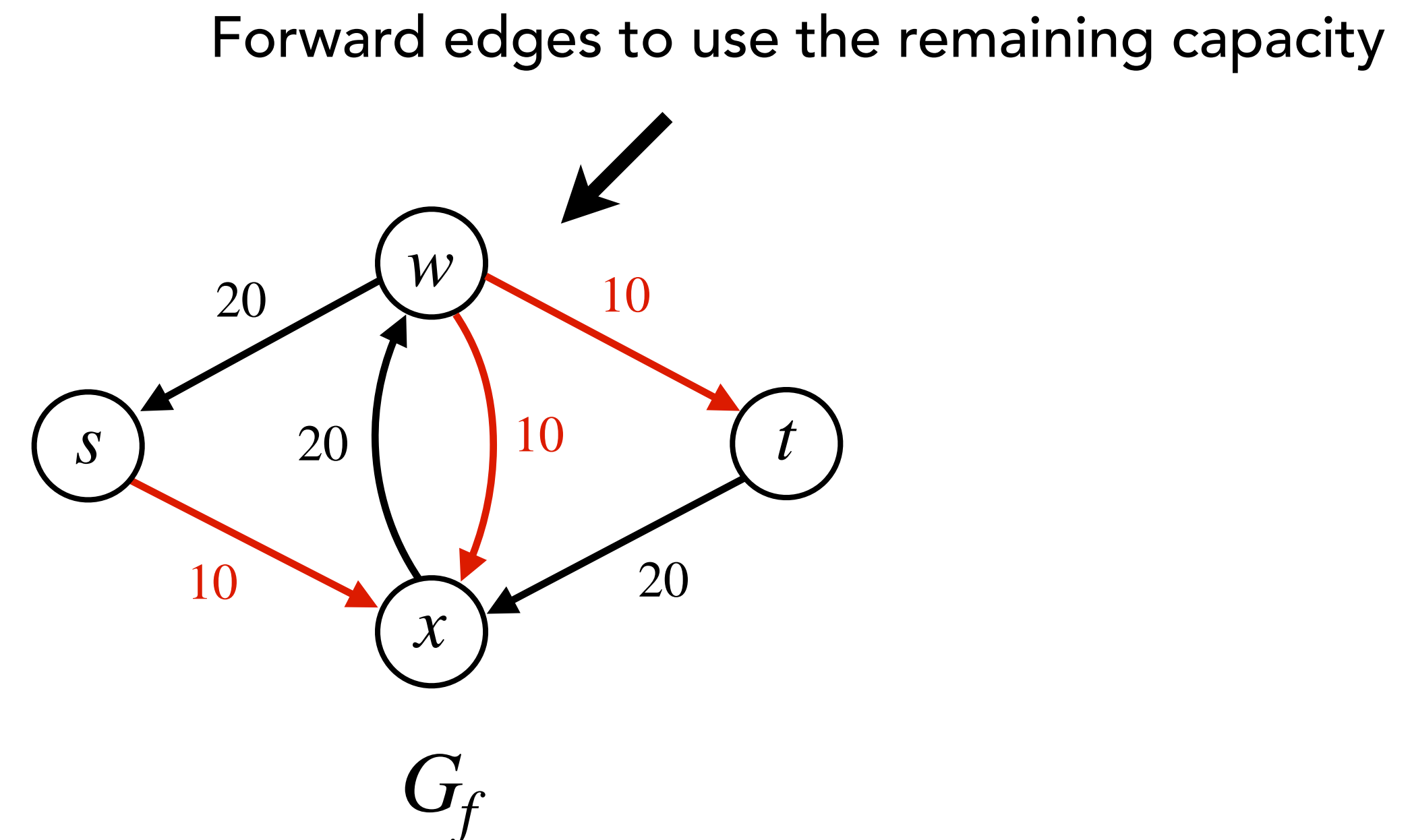
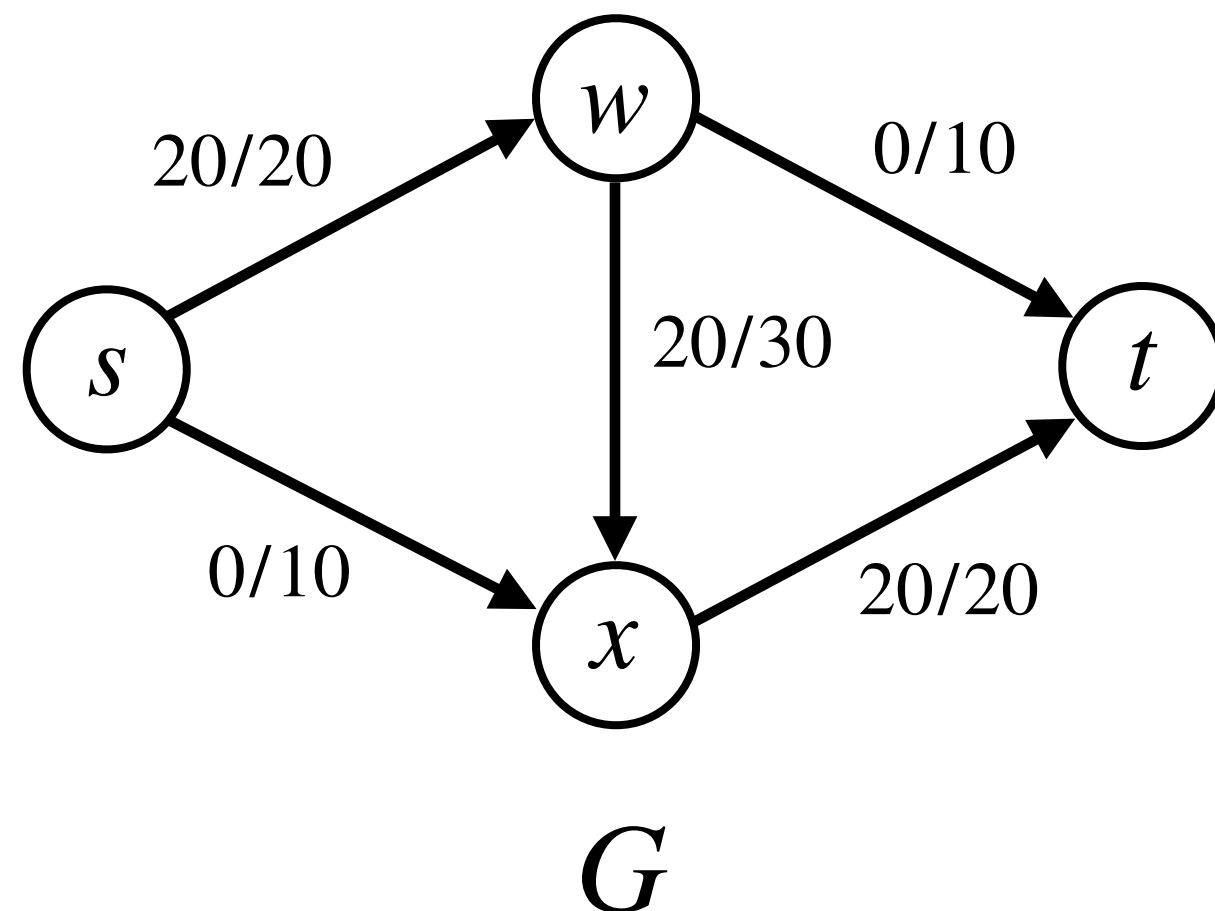


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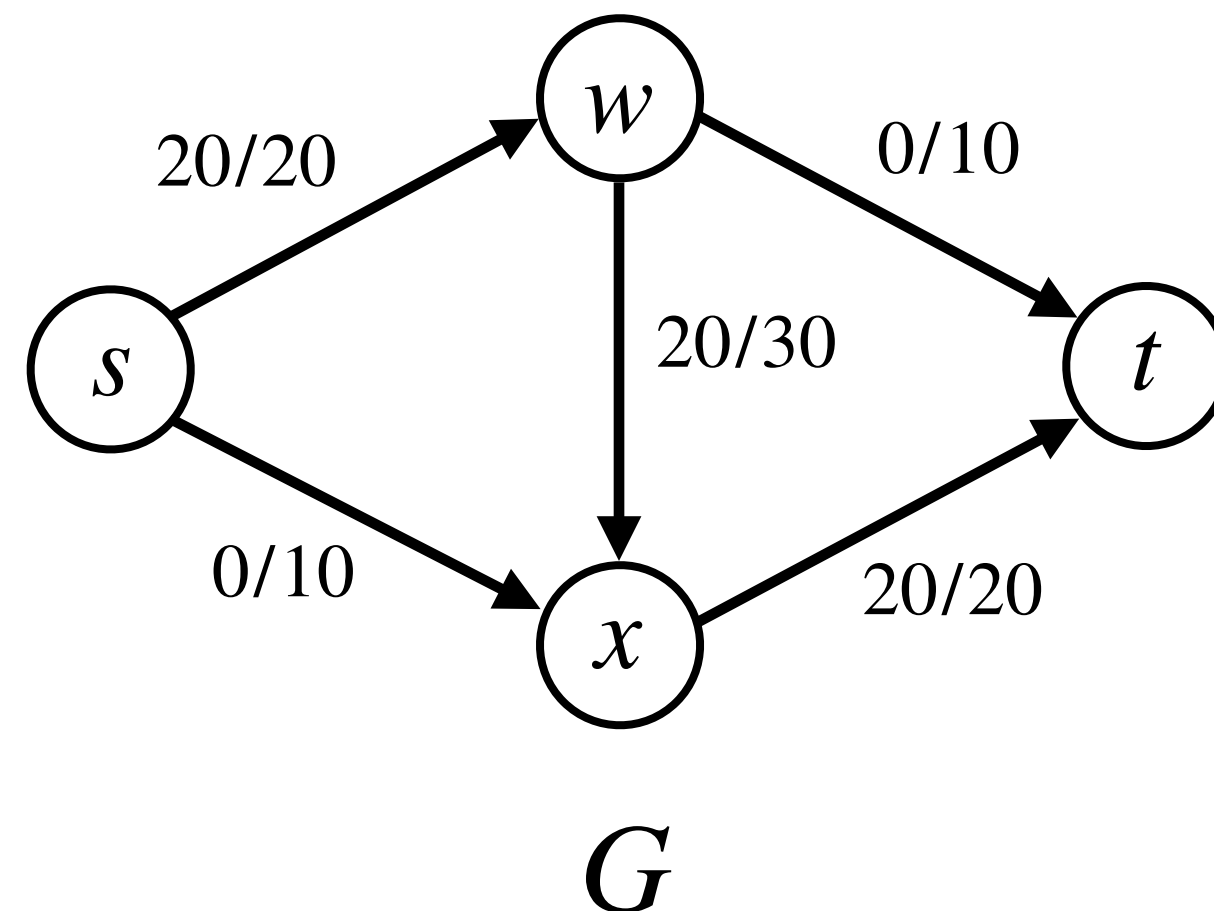


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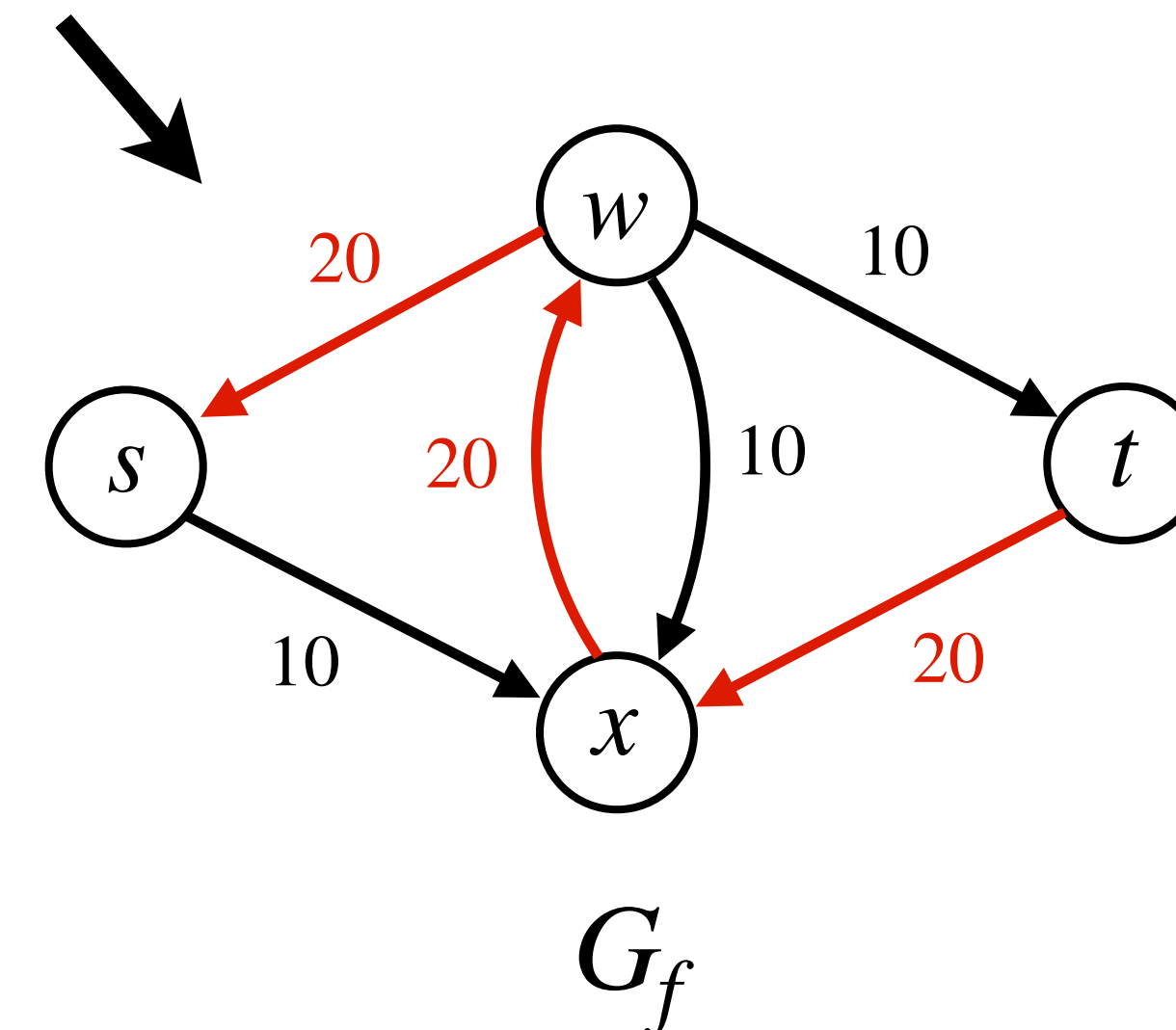
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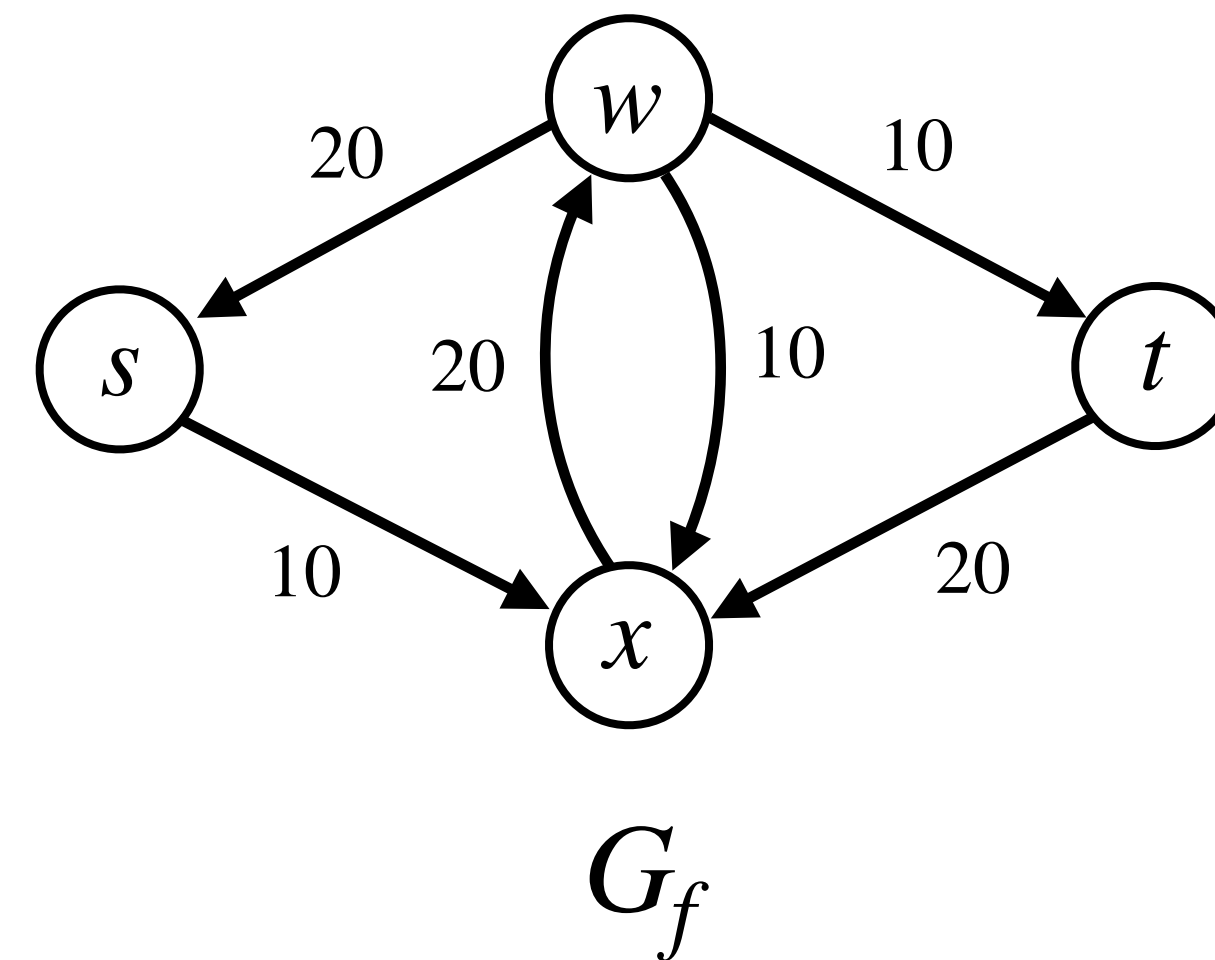
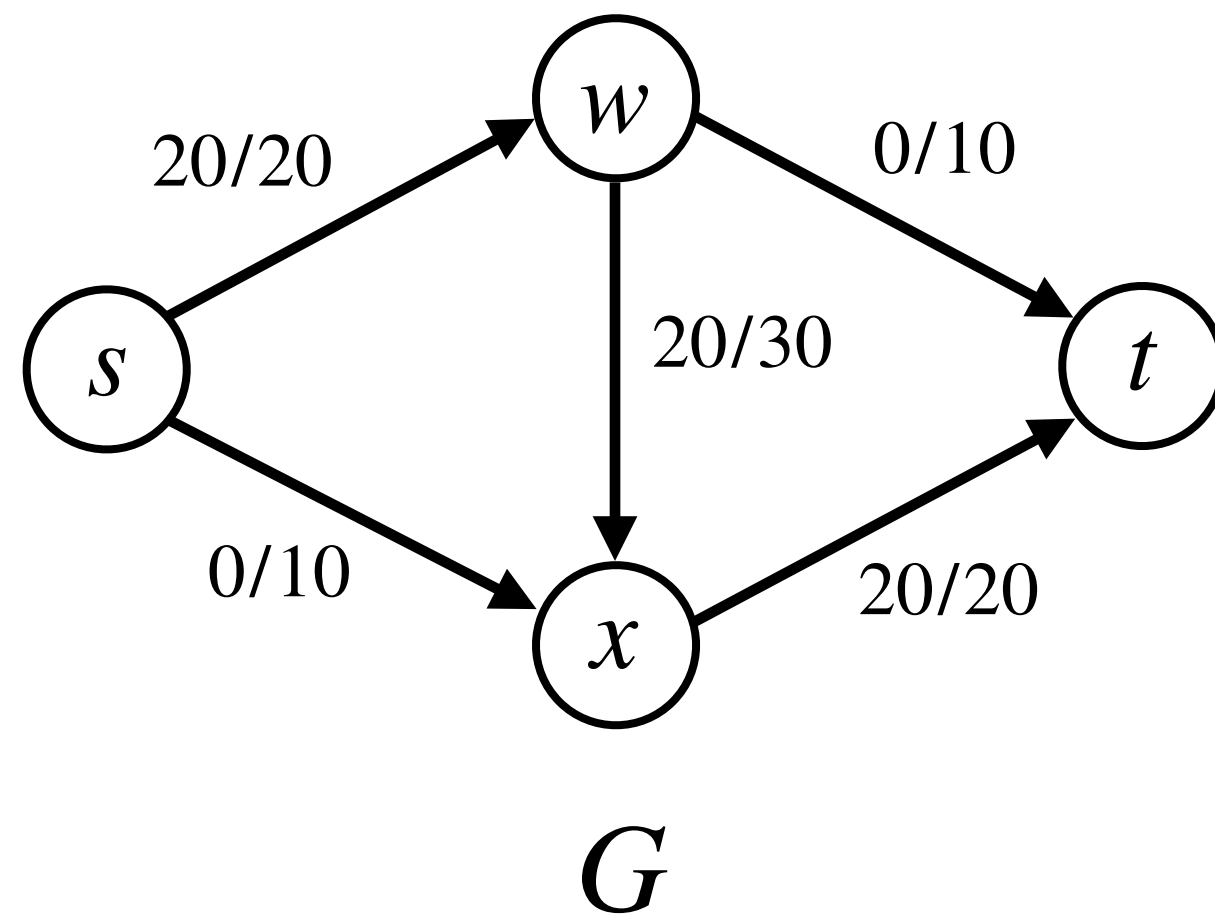


Backward edges to decrease the flow on some edges



Augmenting Flows via Residual Networks

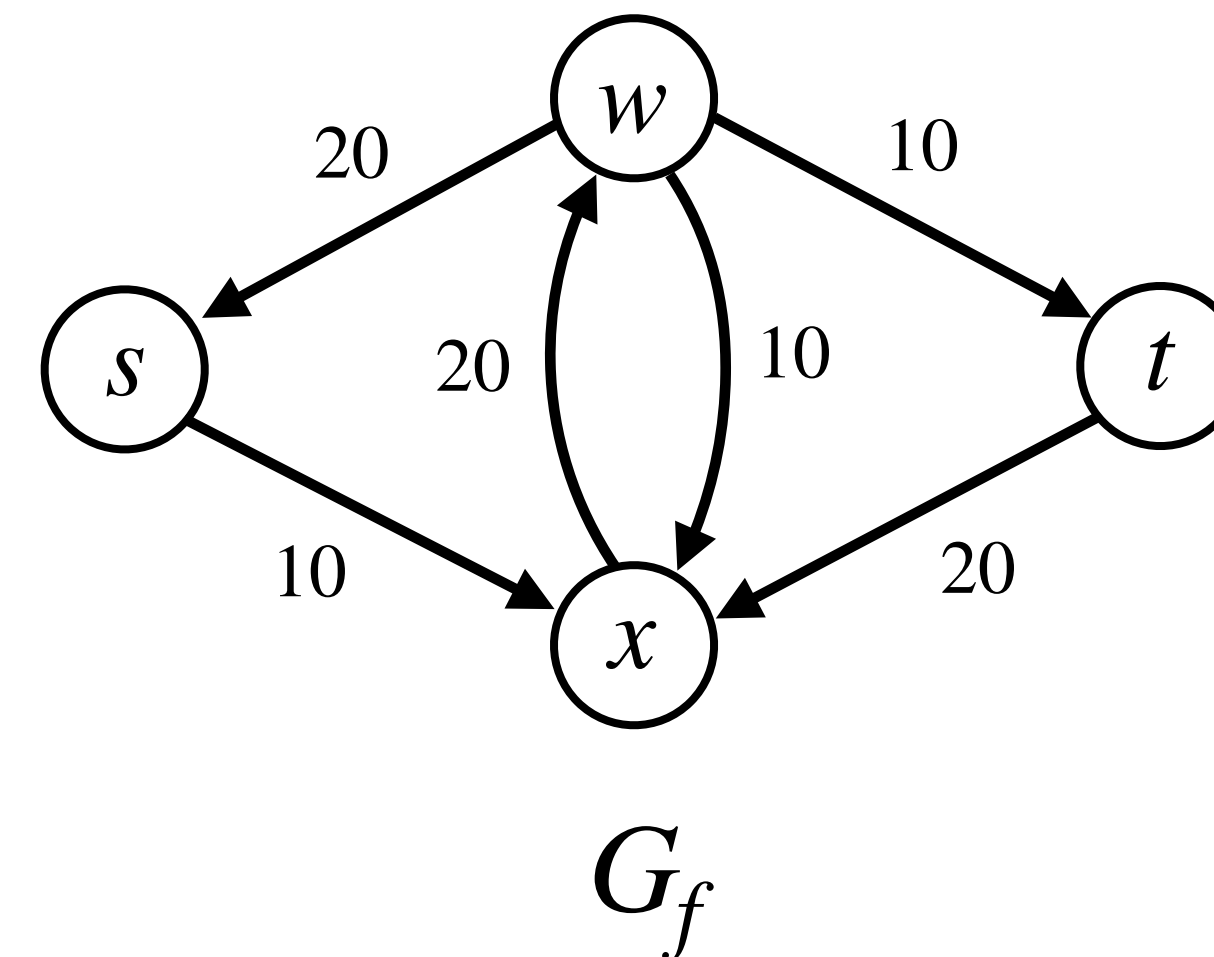
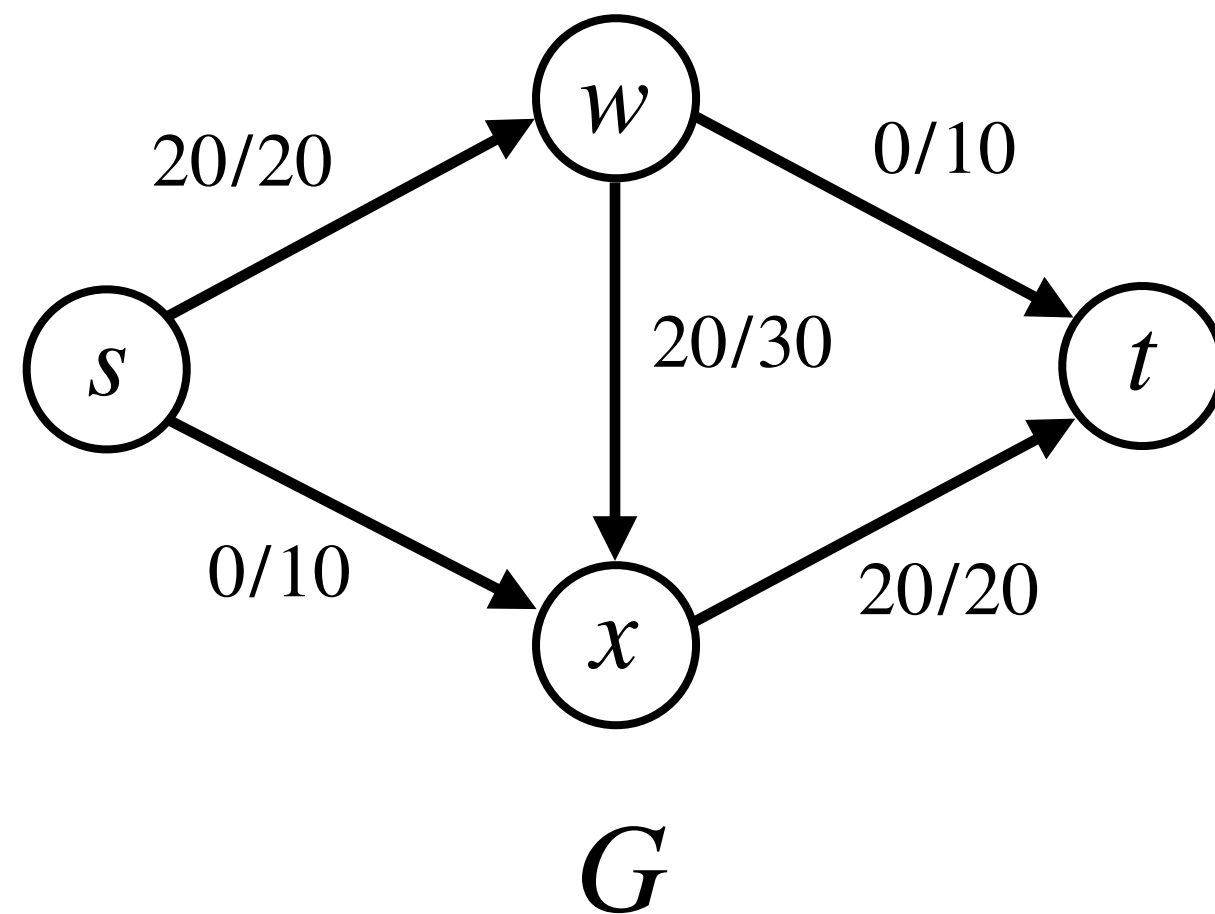
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- Find $s \rightsquigarrow t$ path P in the residual network G_f and its bottleneck capacity δ .

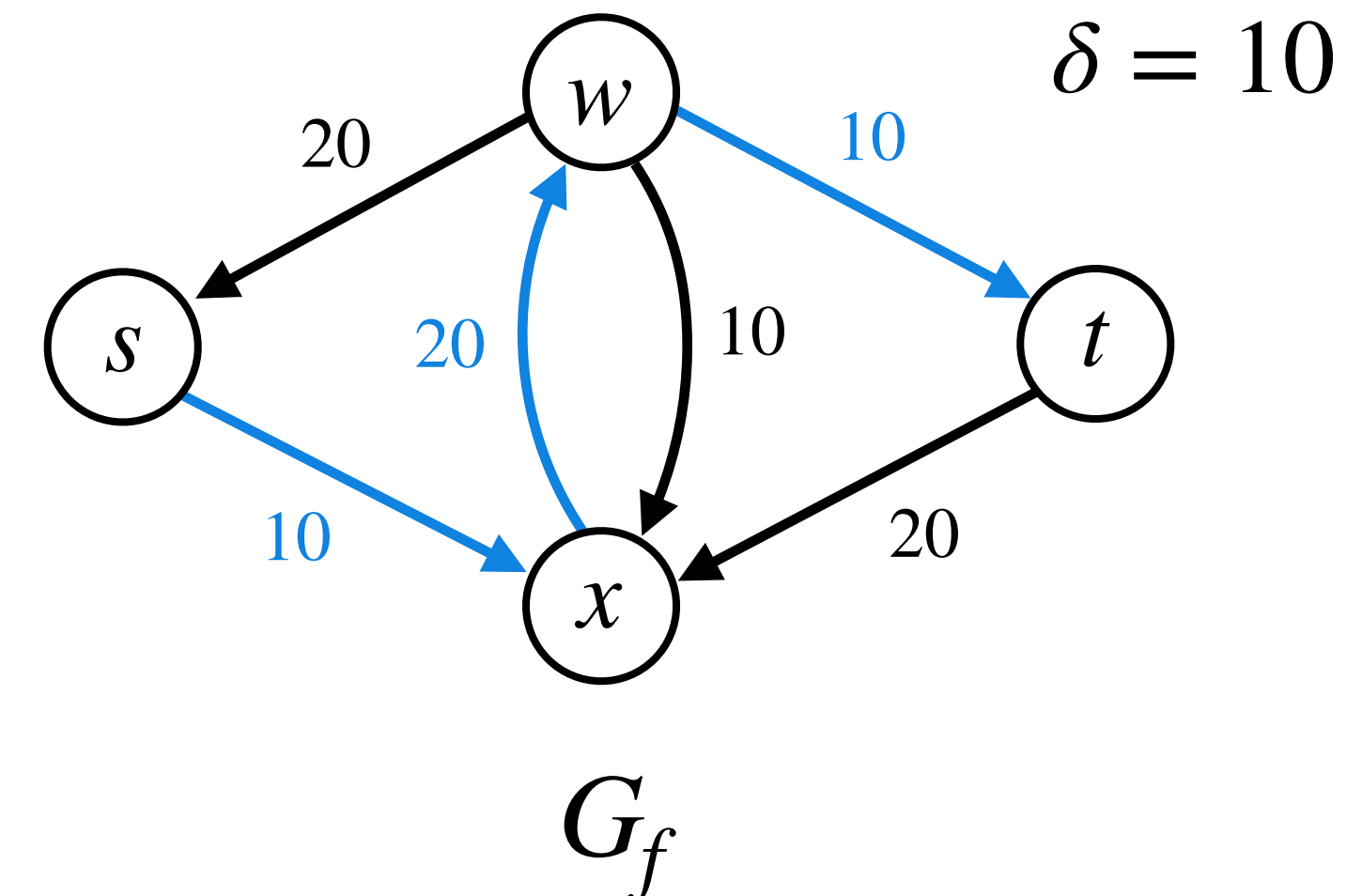
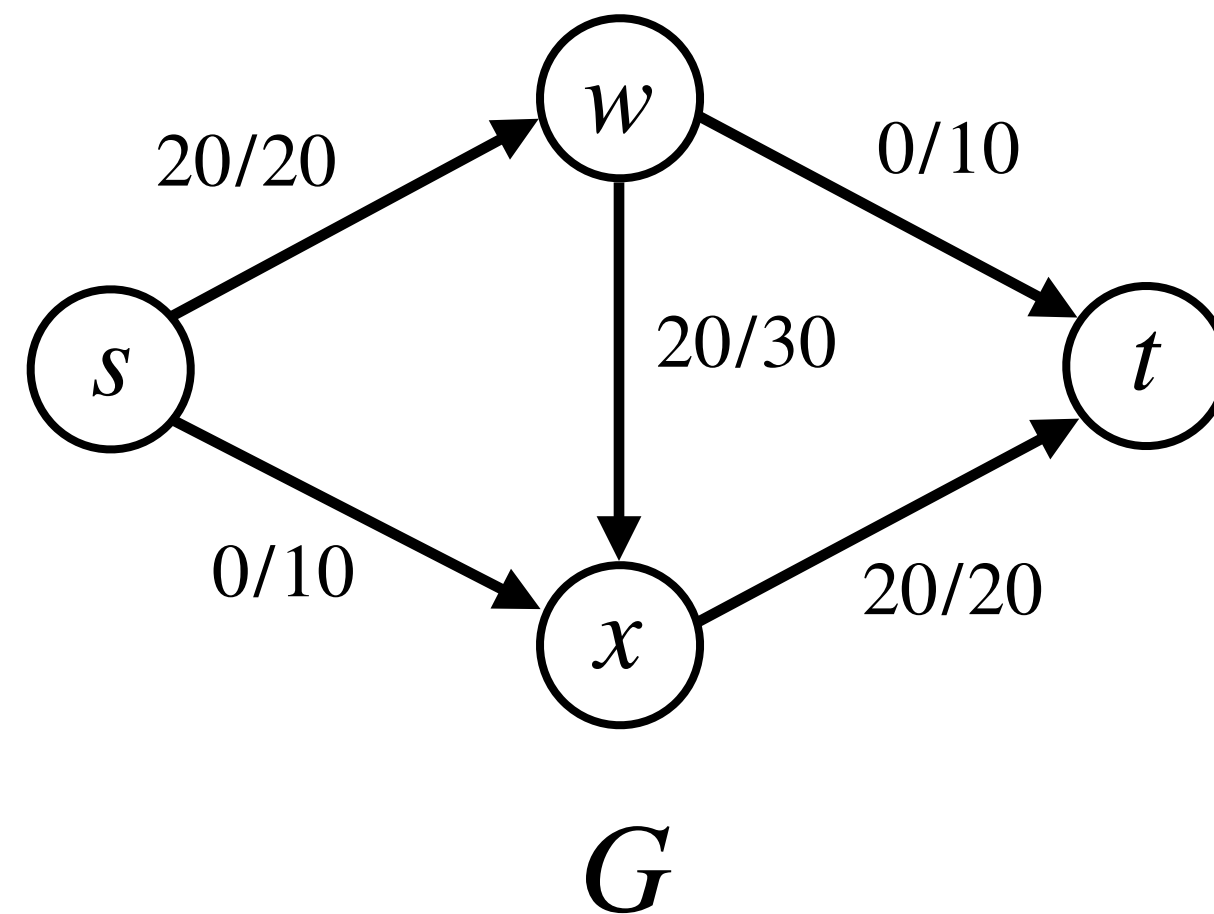
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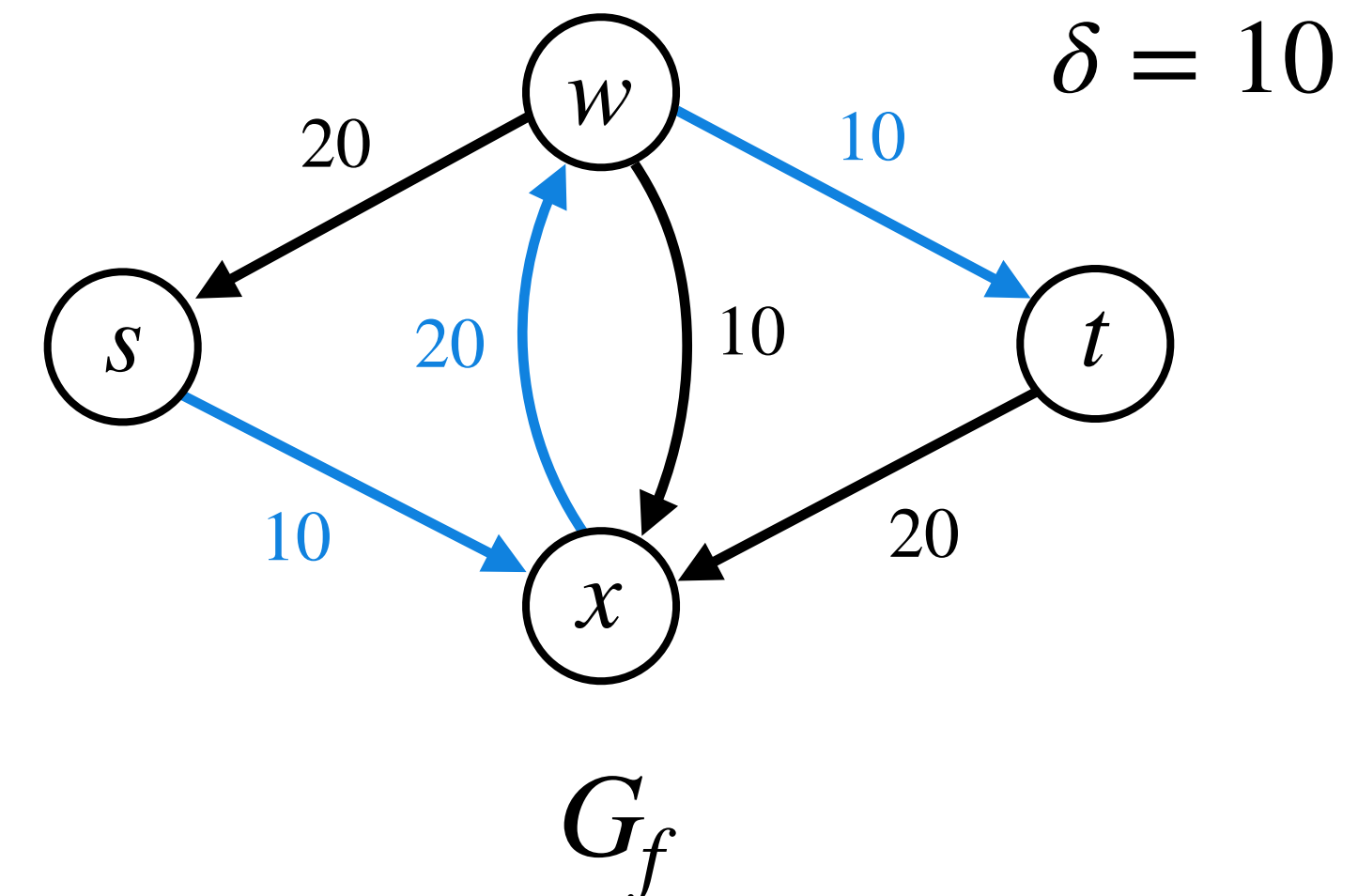
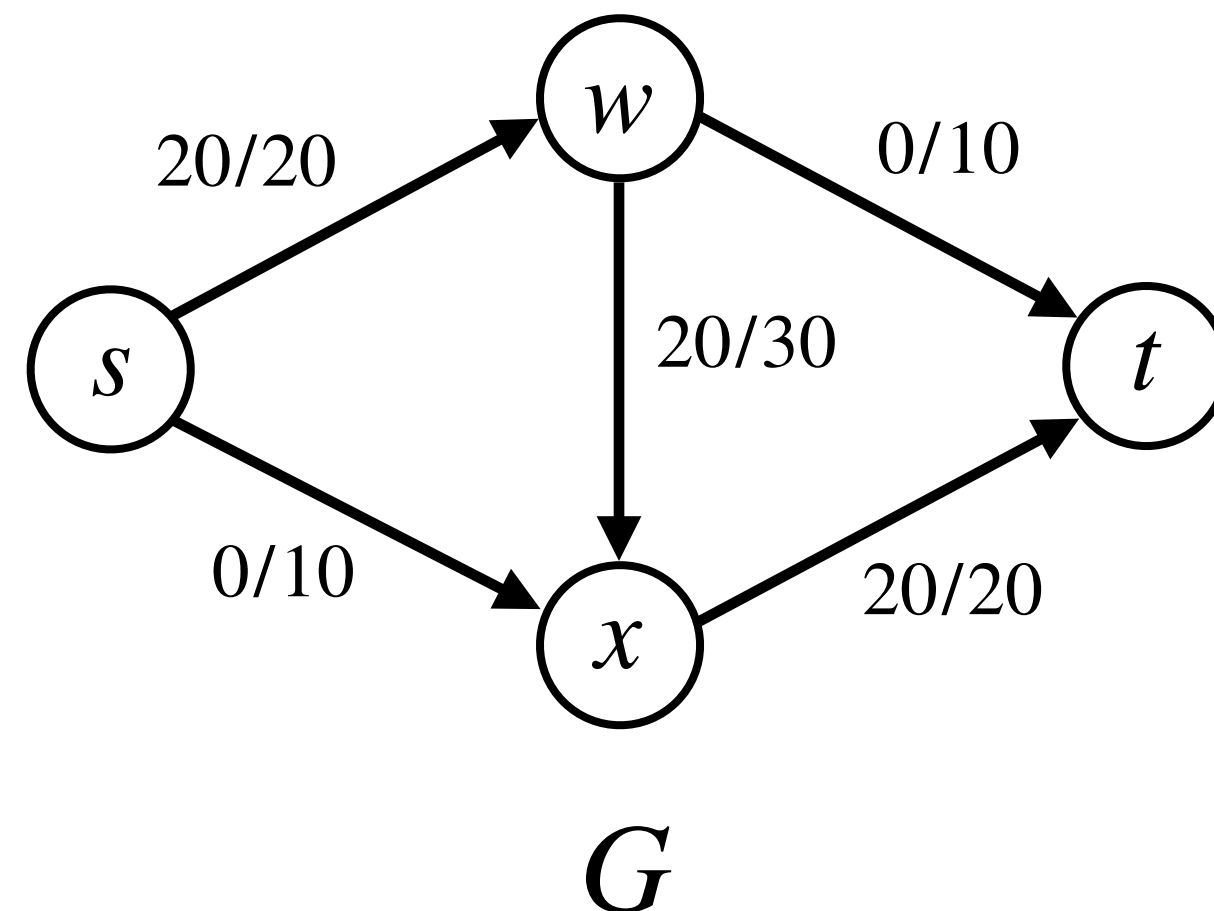
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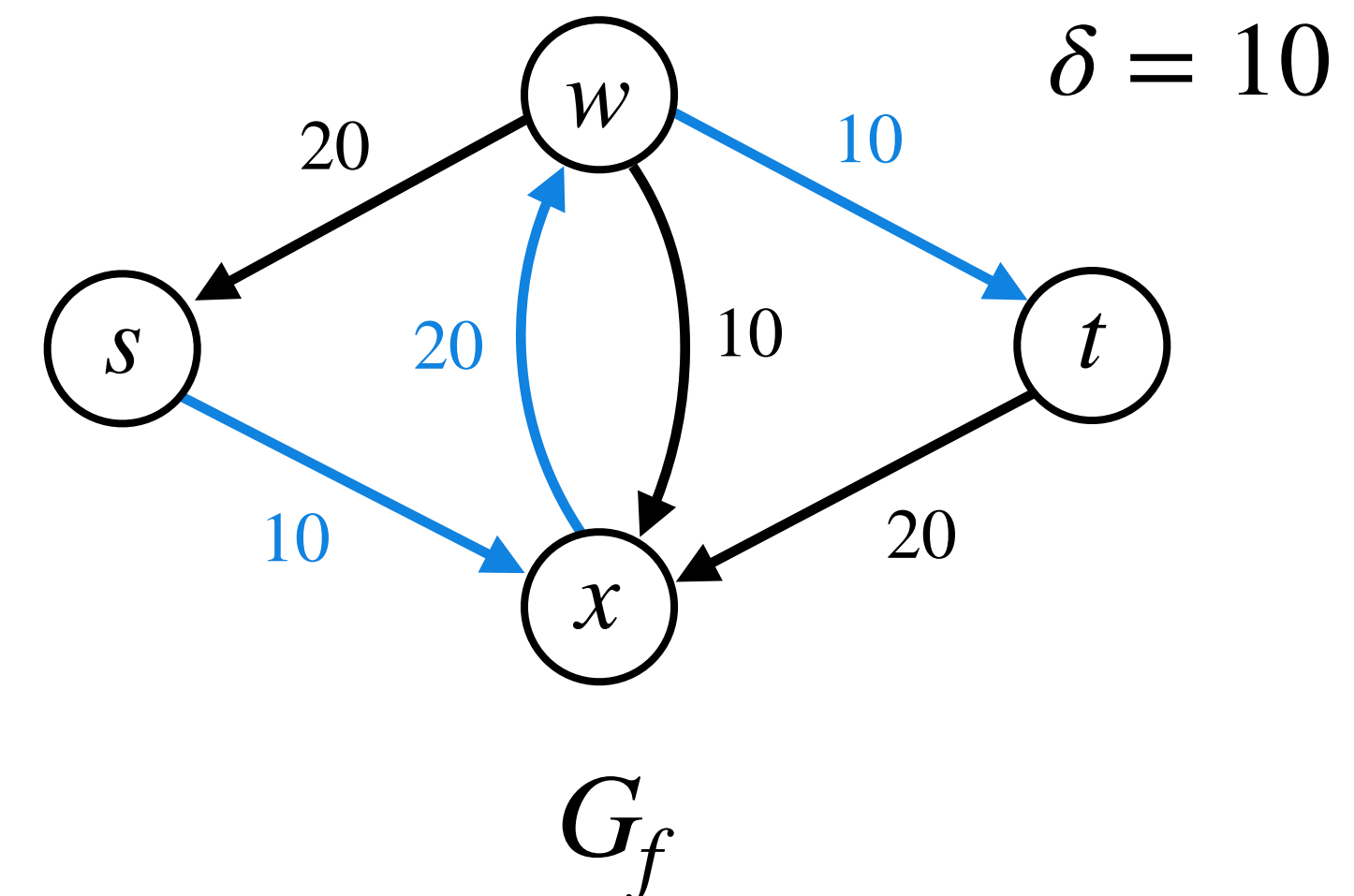
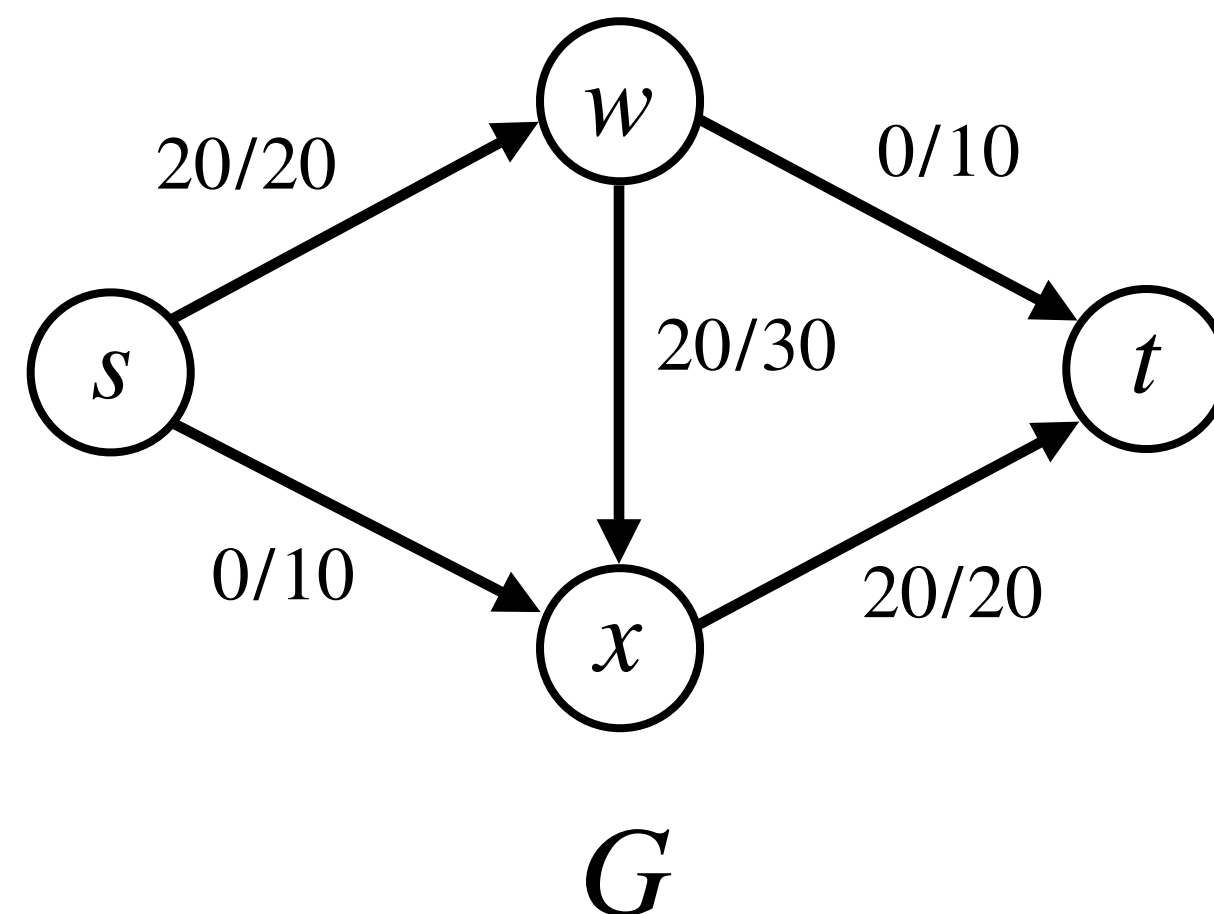
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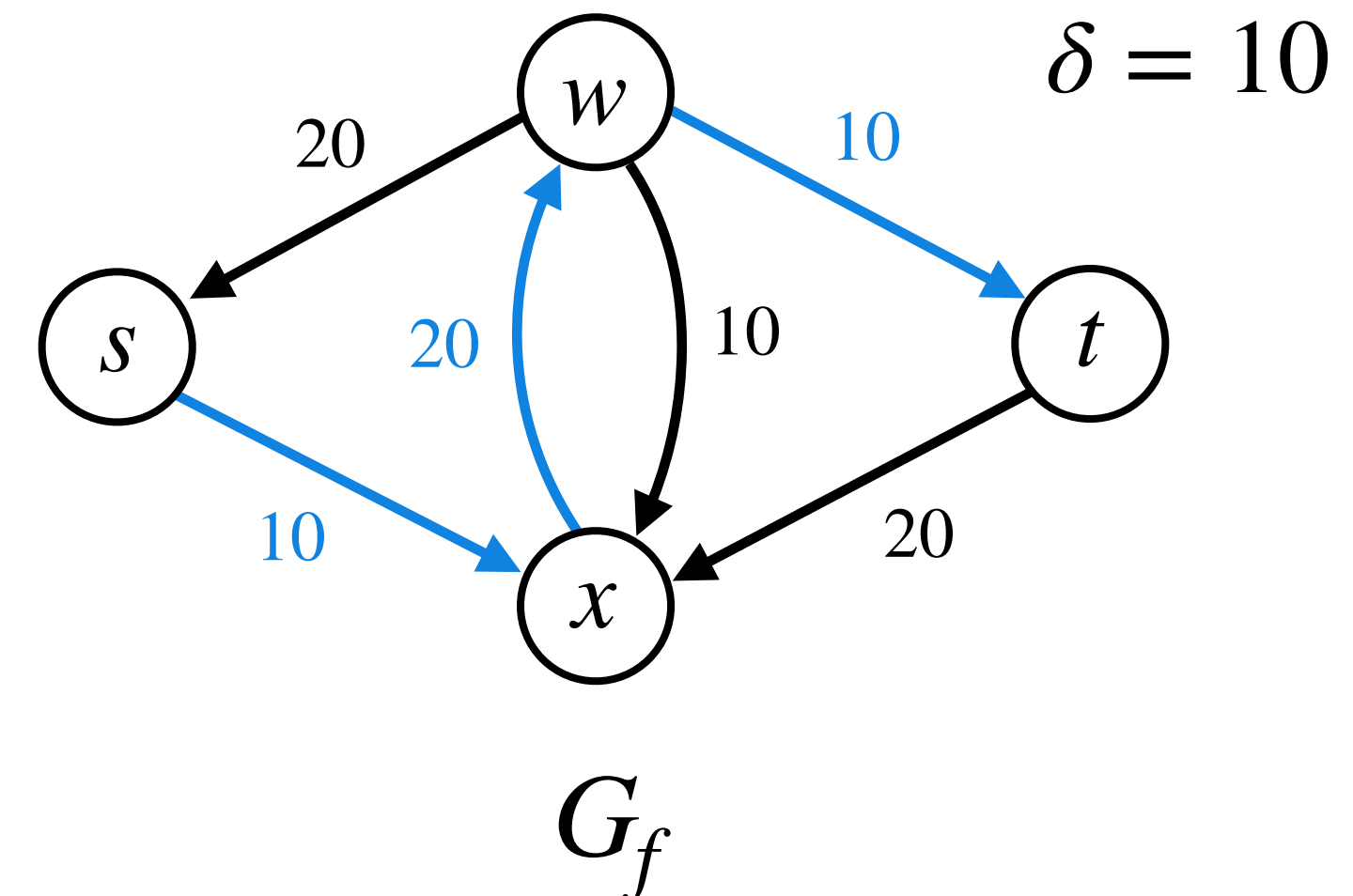
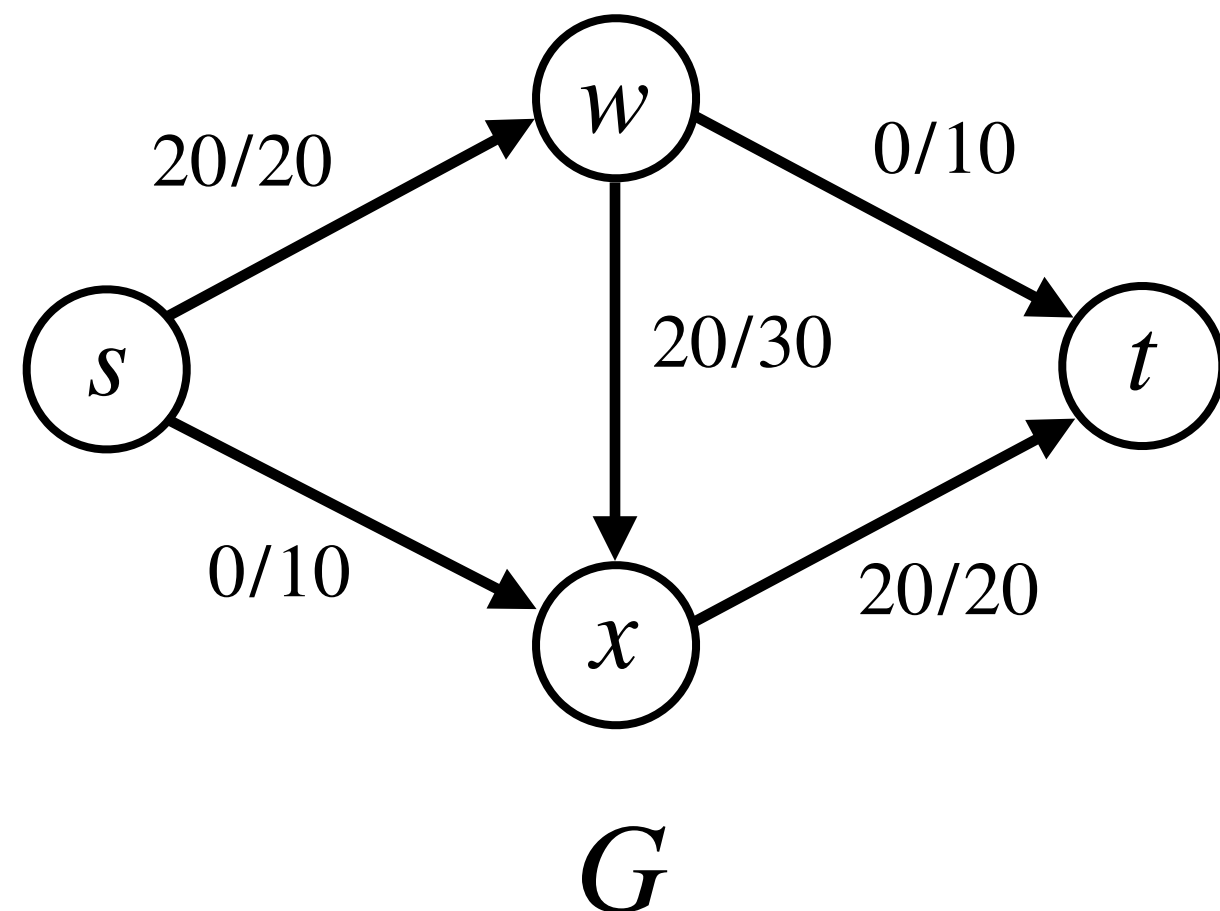
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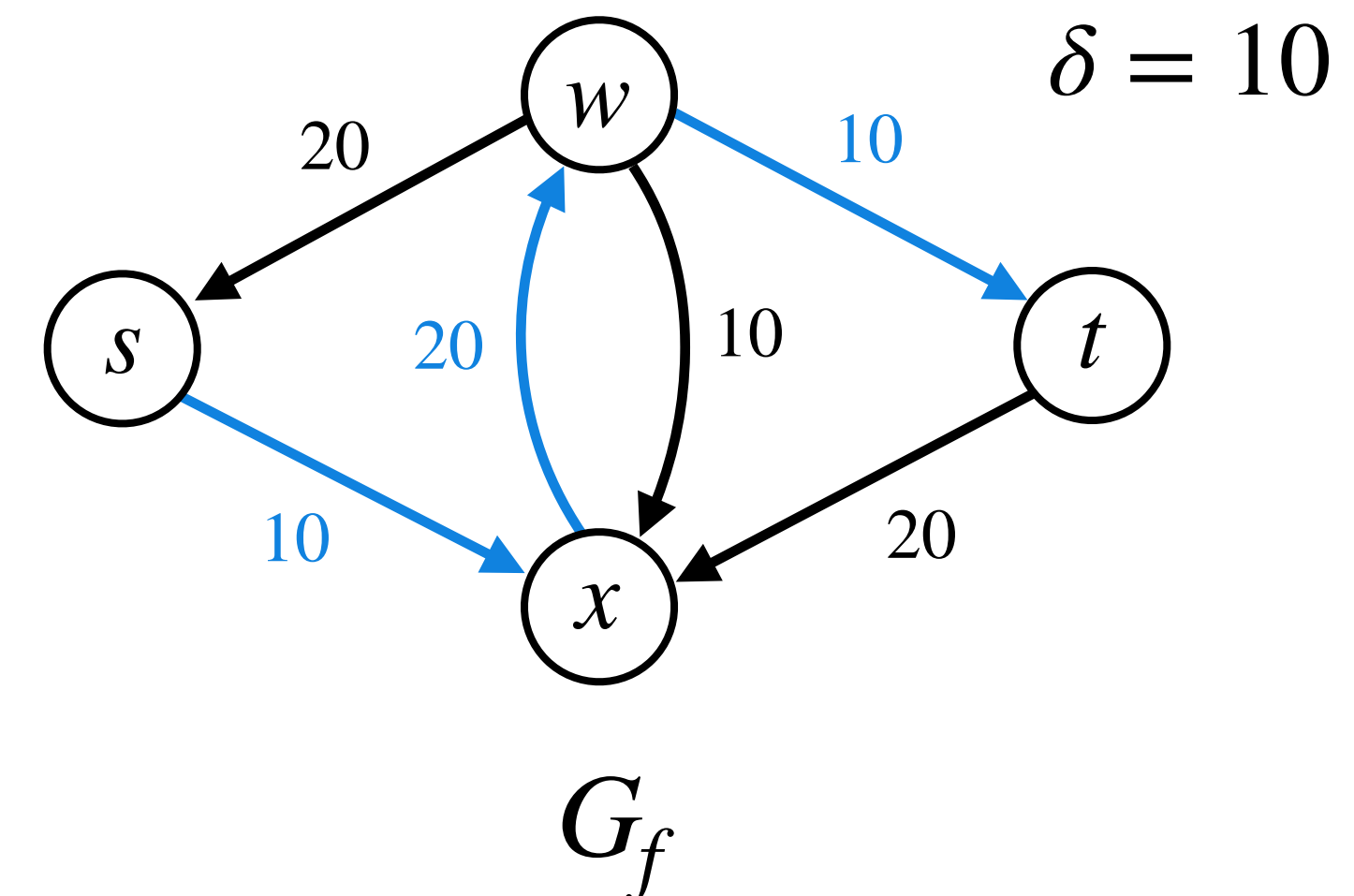
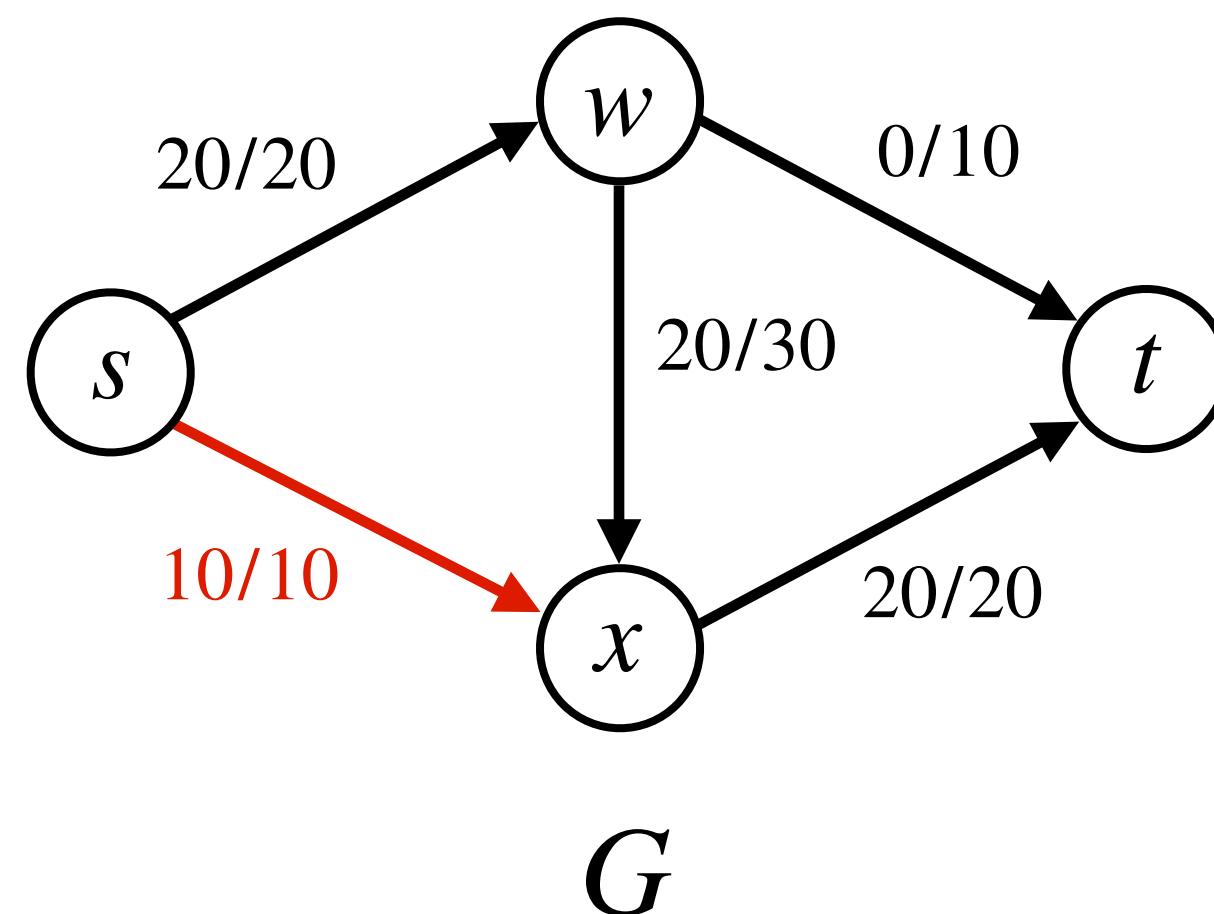
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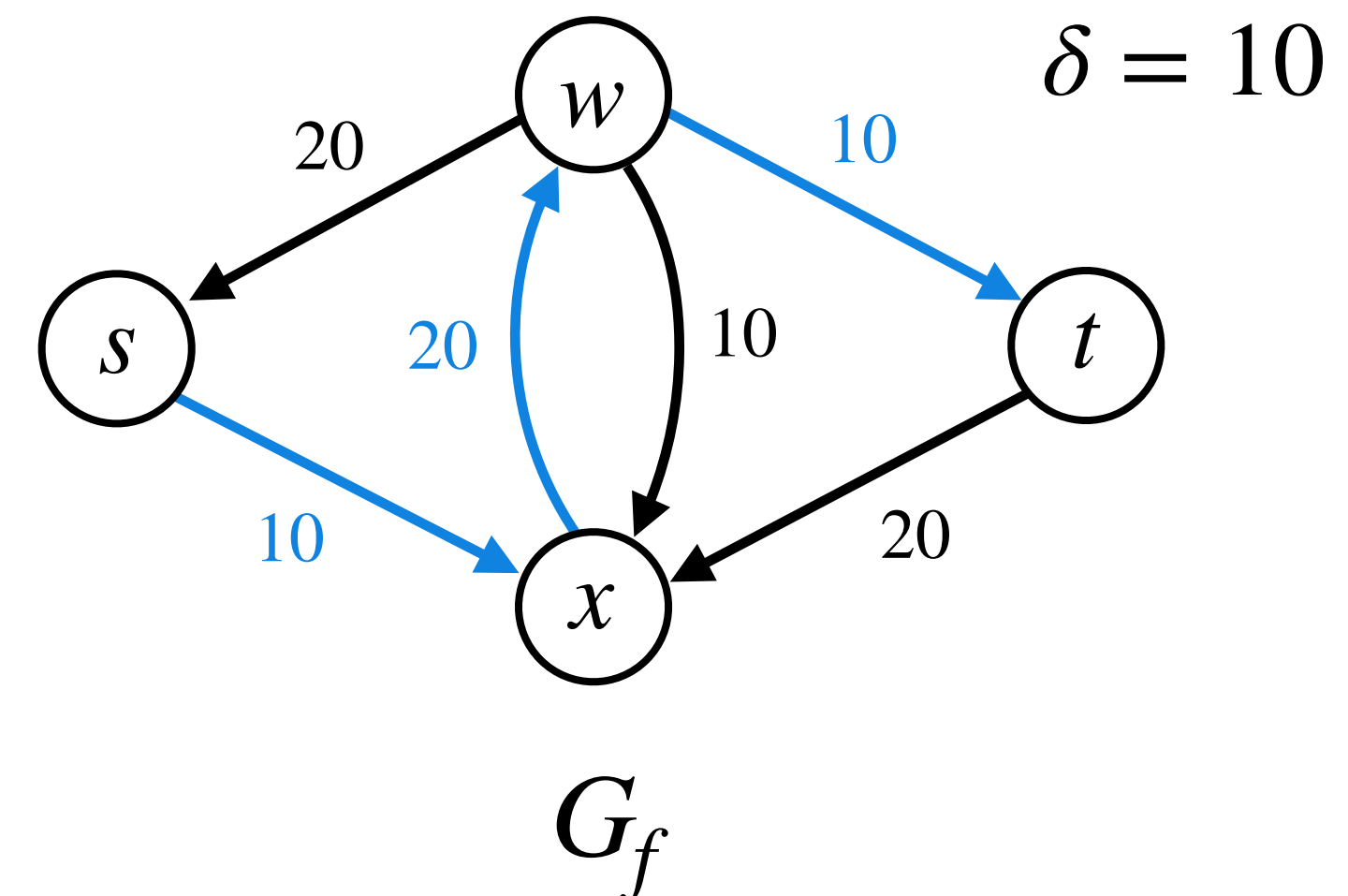
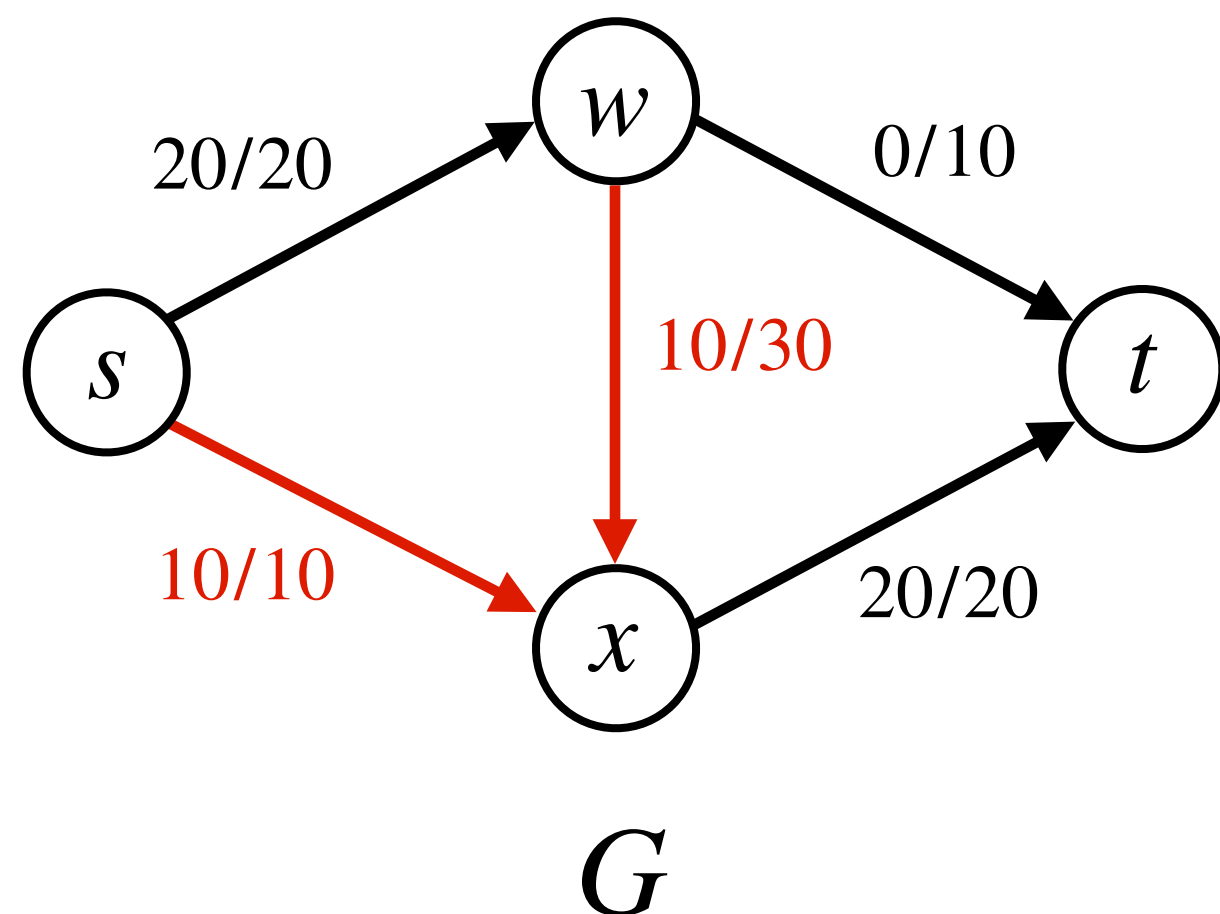
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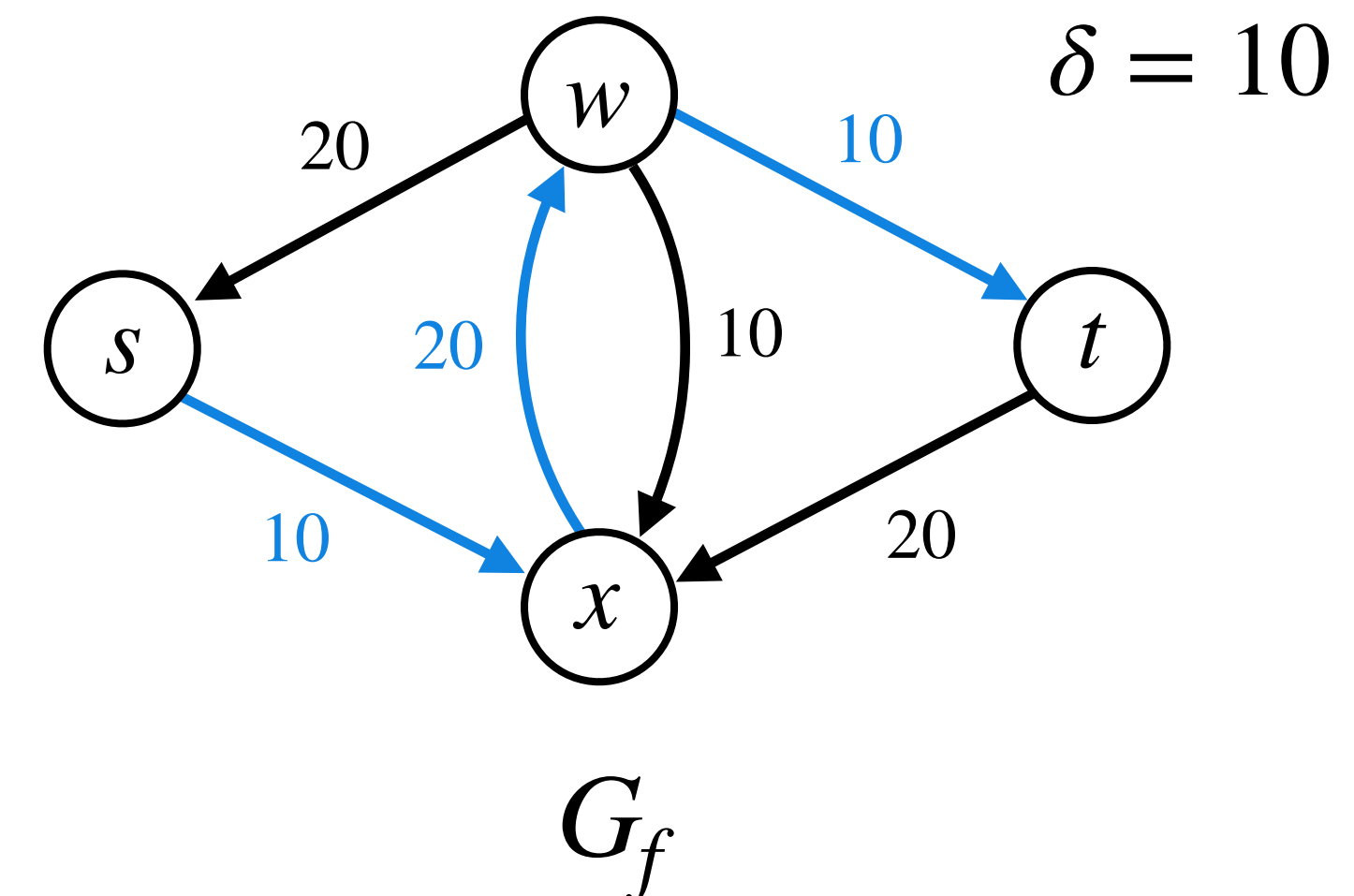
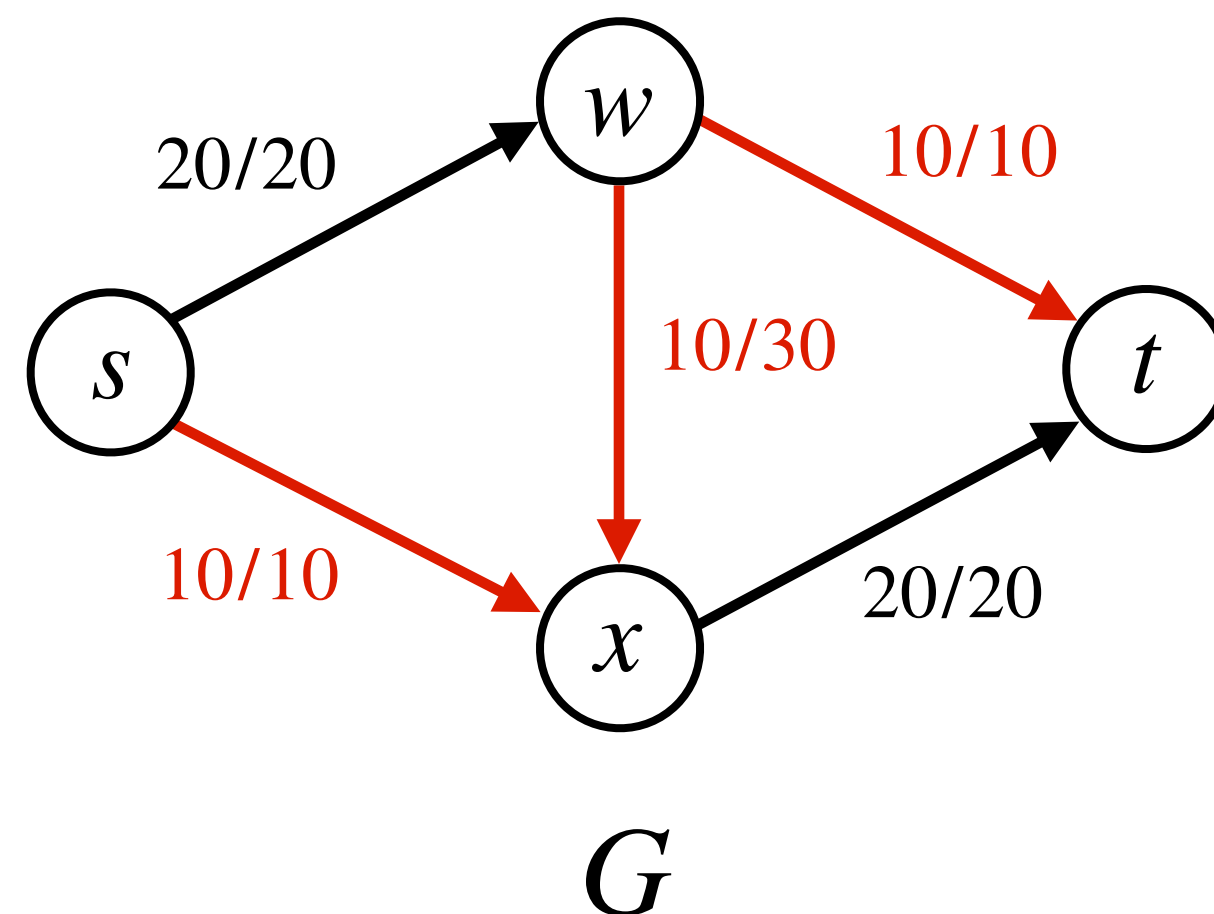
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- What about capacity, conservation constraints?

Example:

