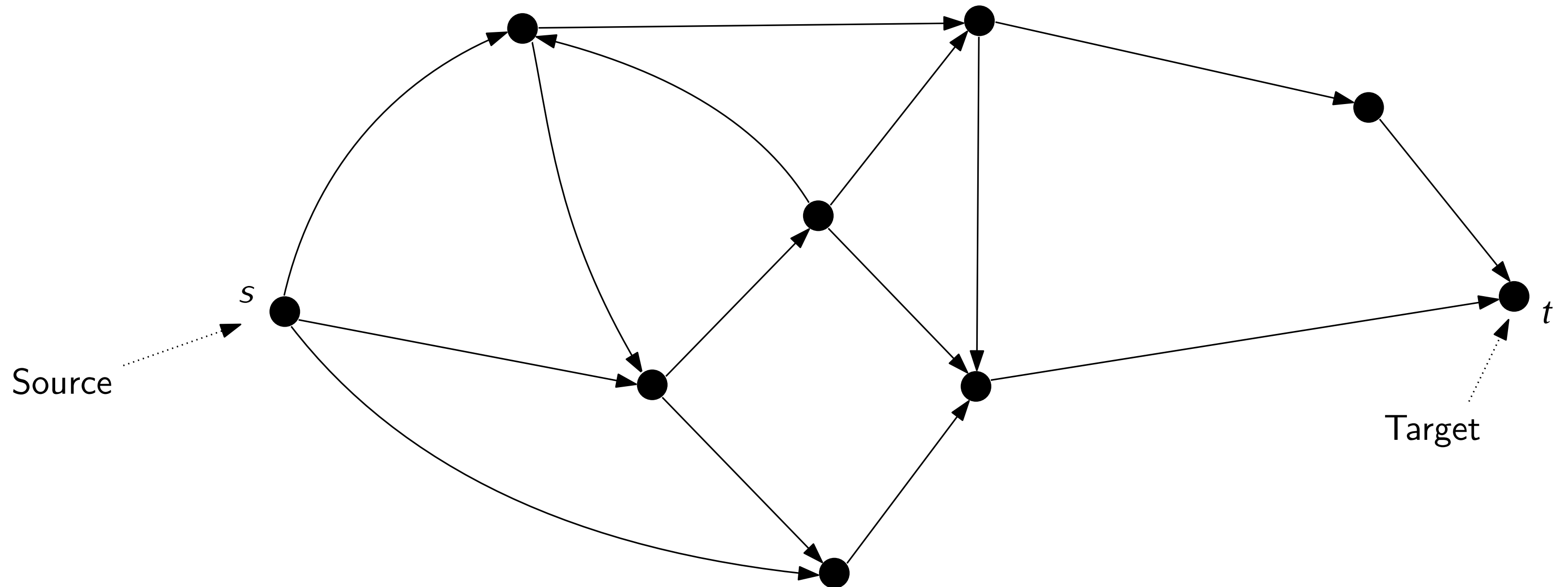


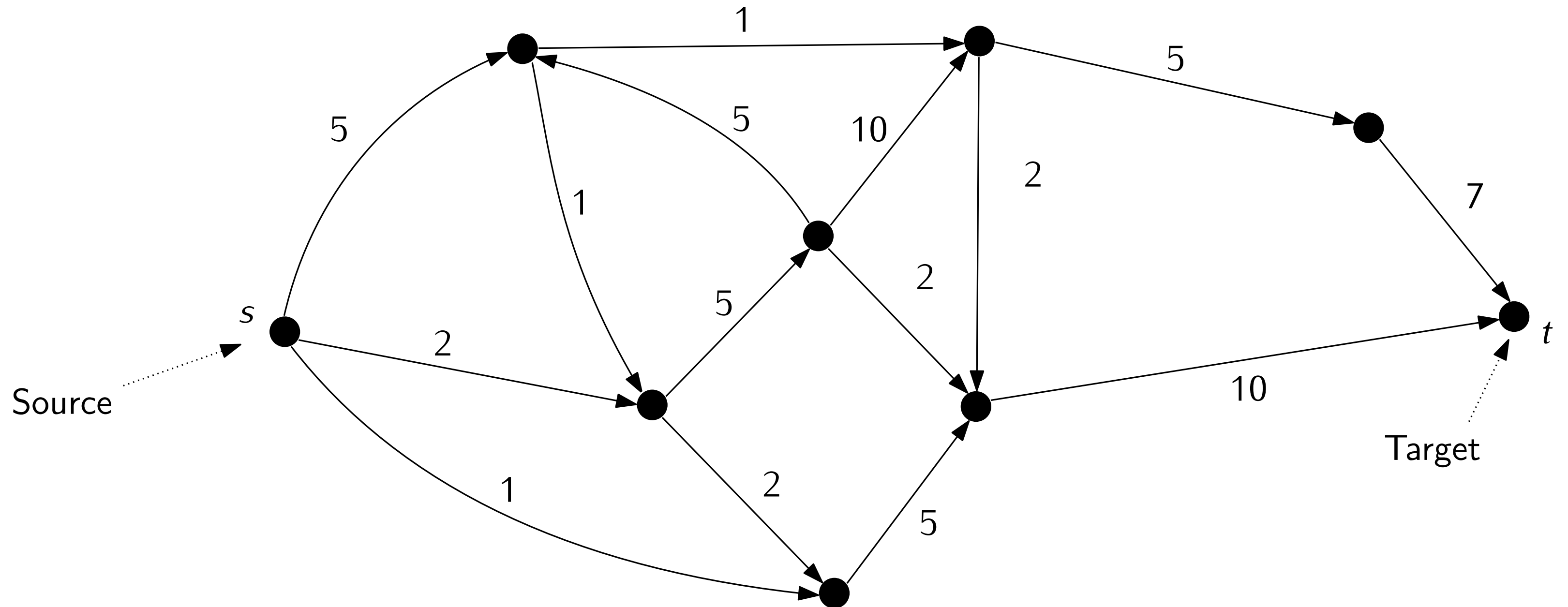
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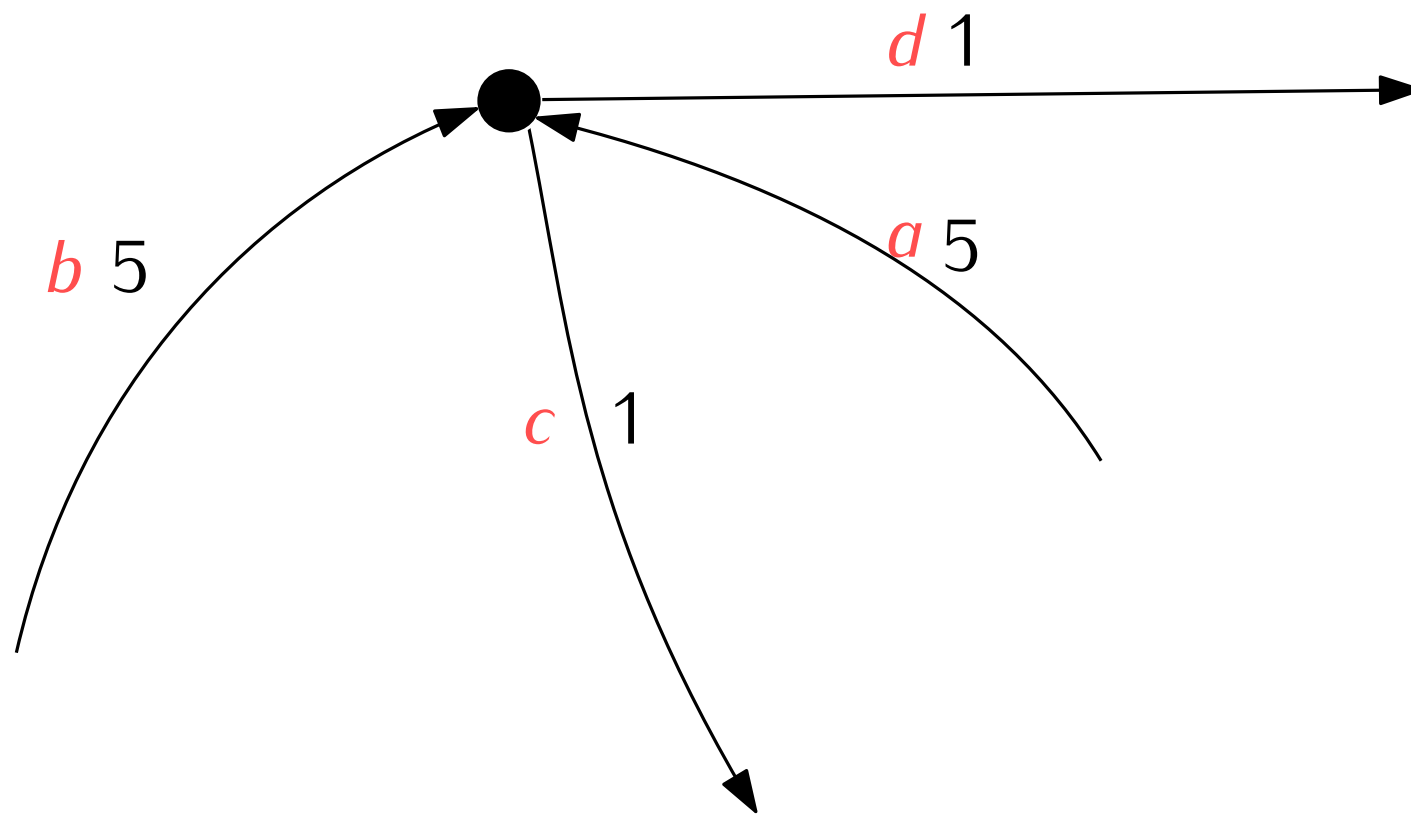
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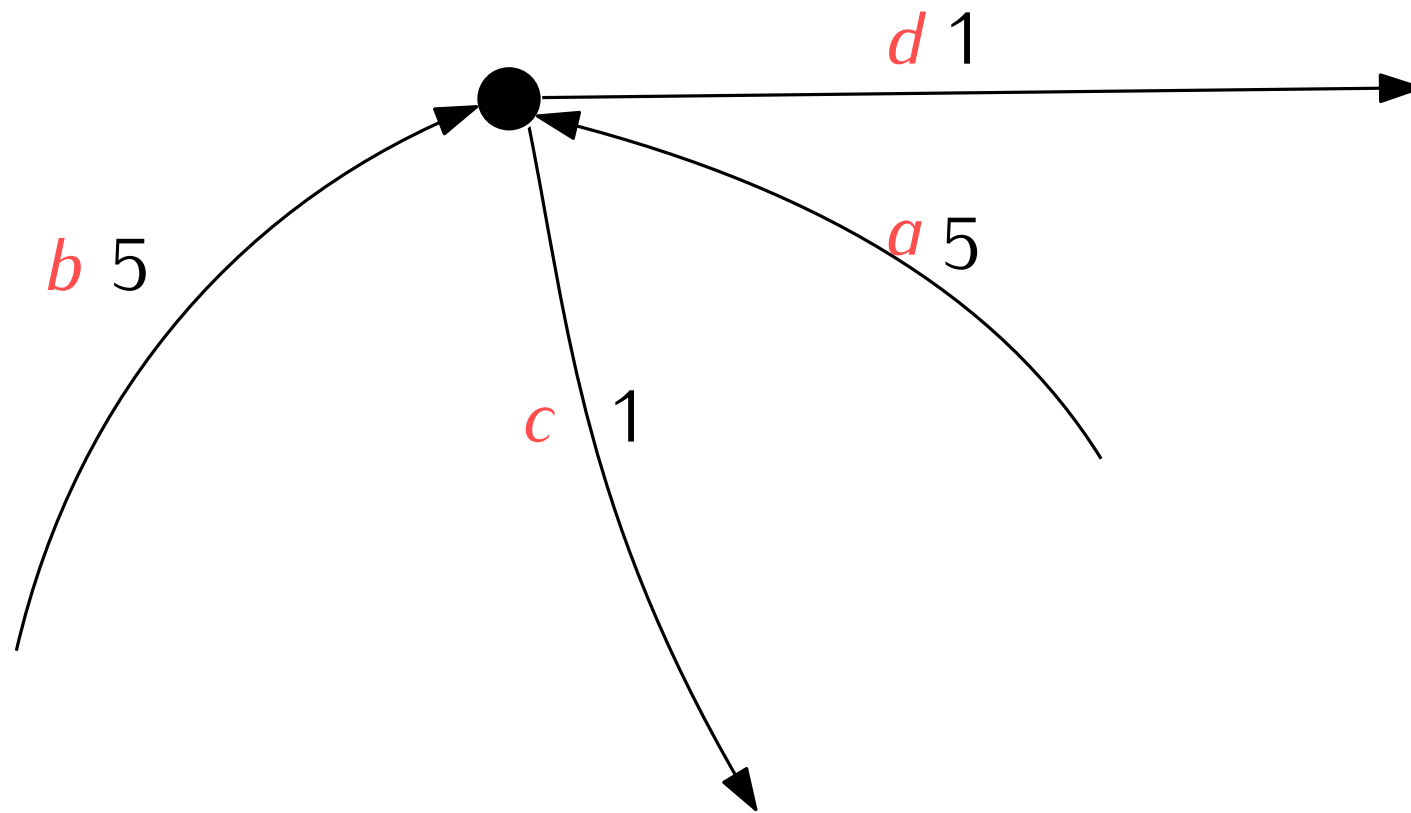
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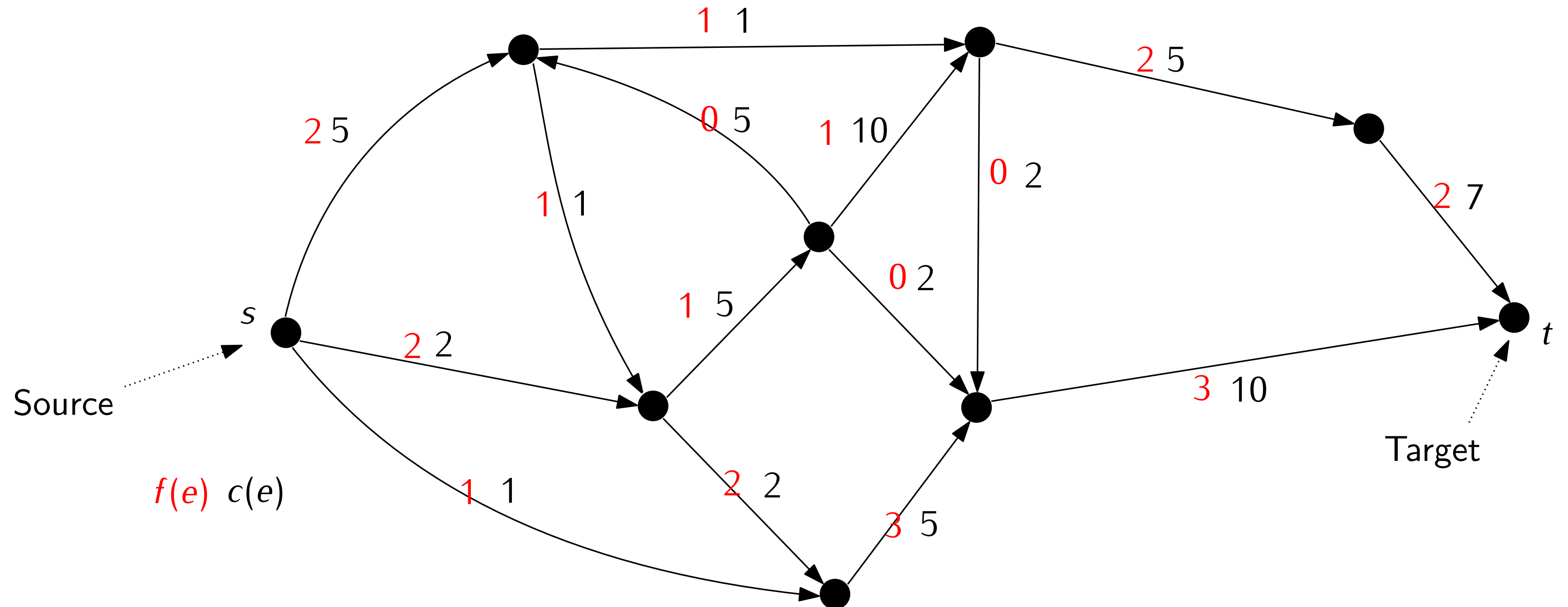
- For all $e \in E$, $0 \leq f(e) \leq c(e)$
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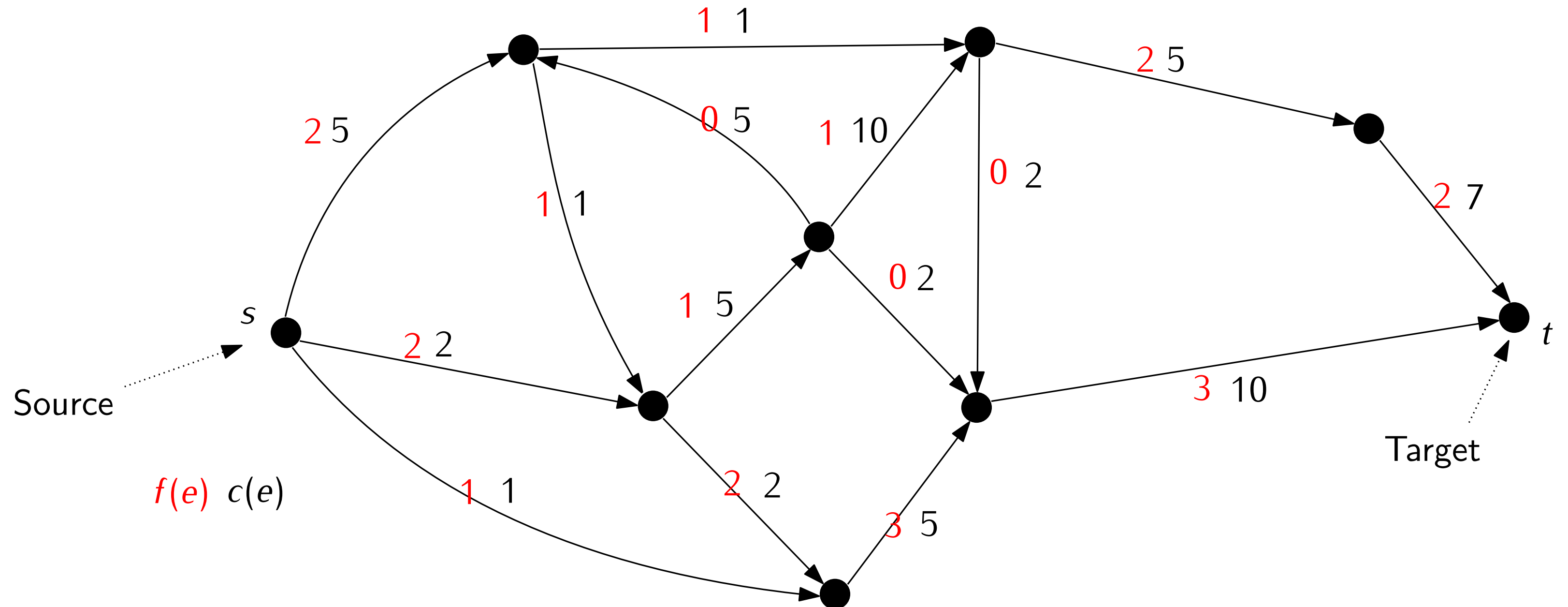
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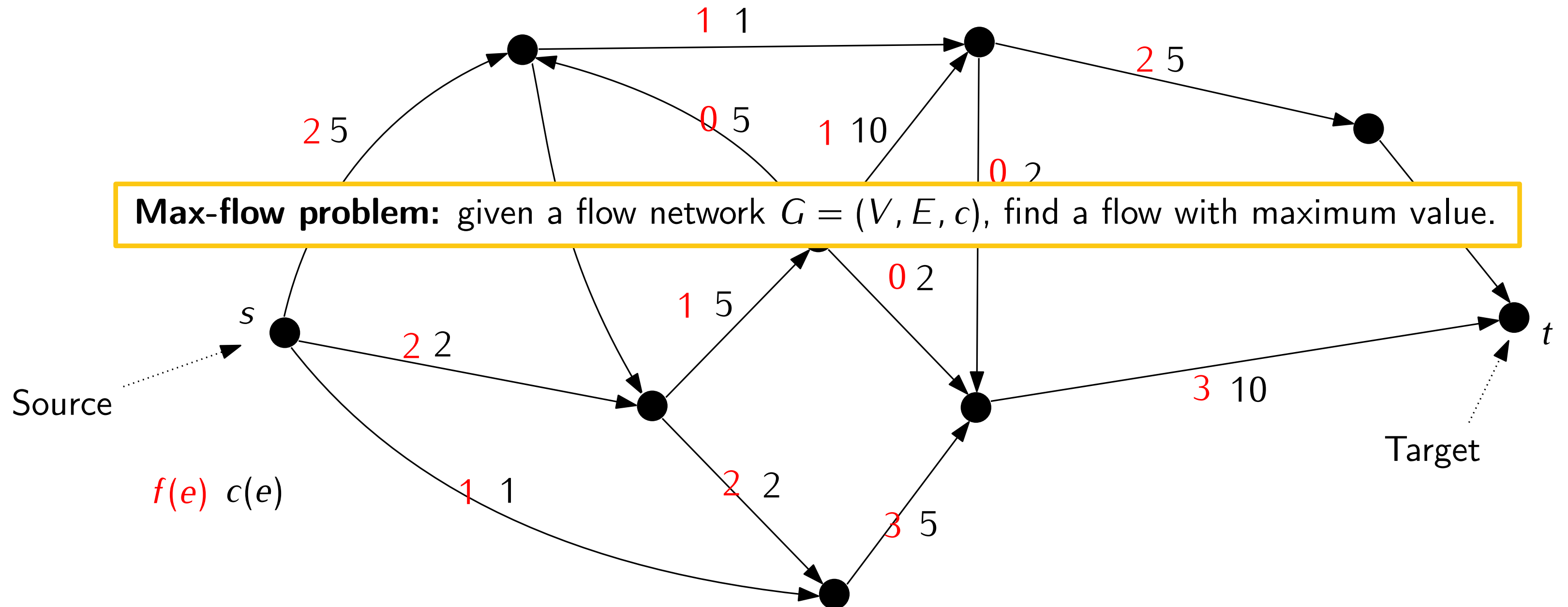
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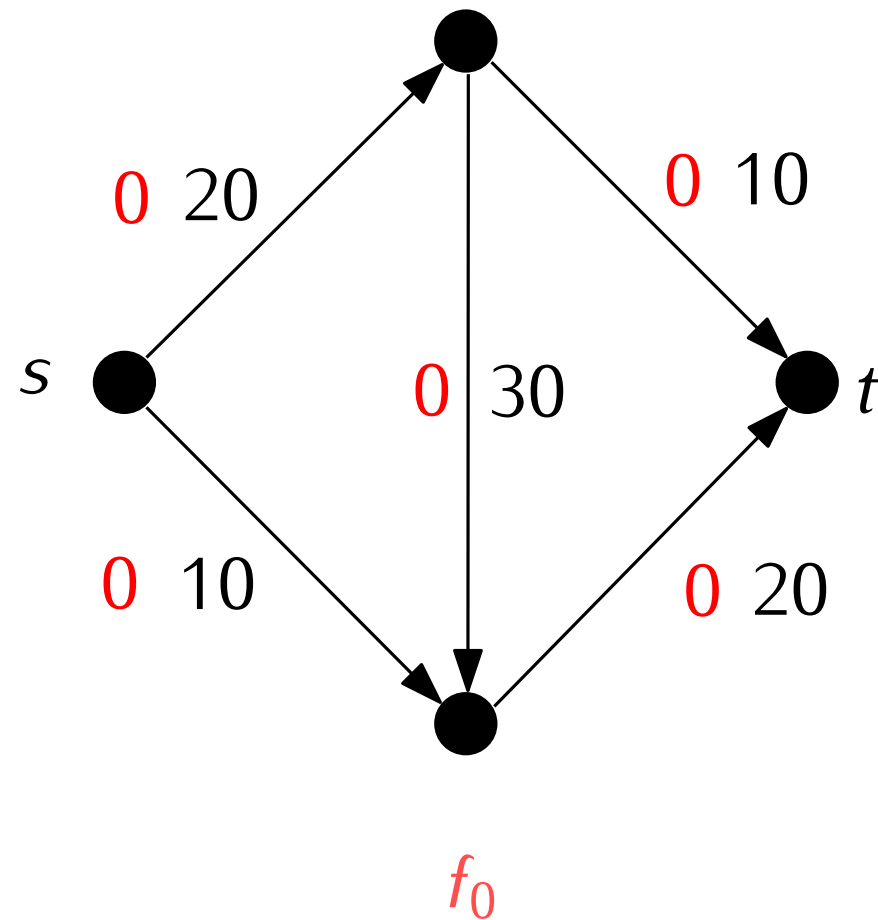
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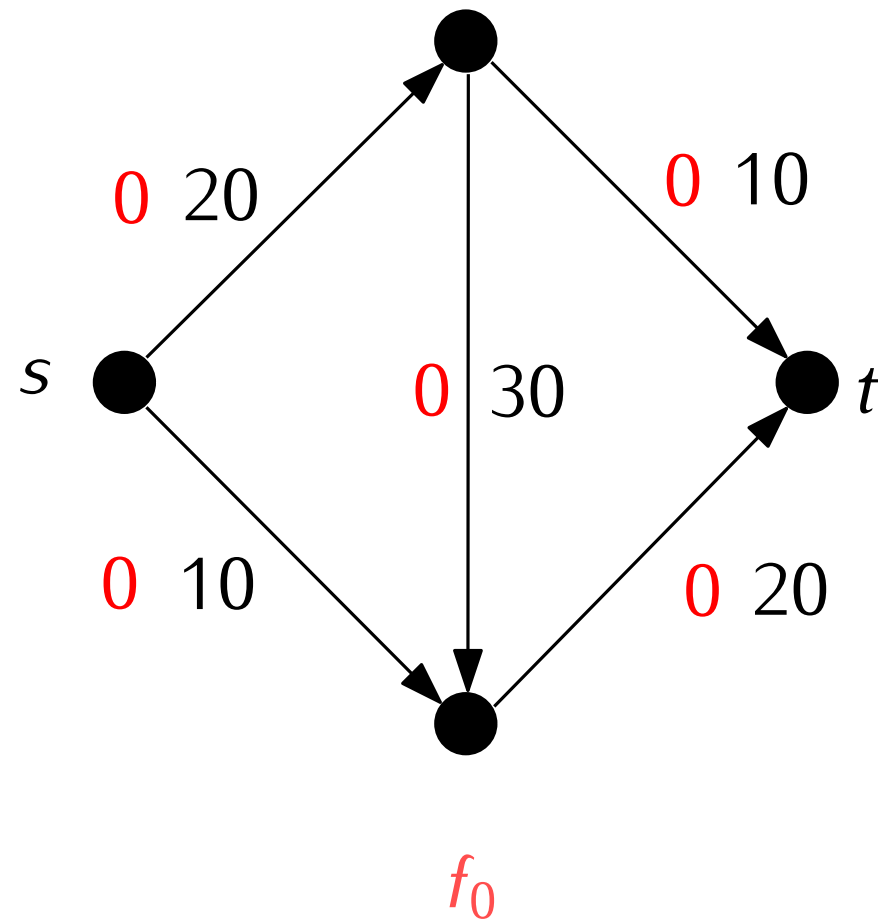
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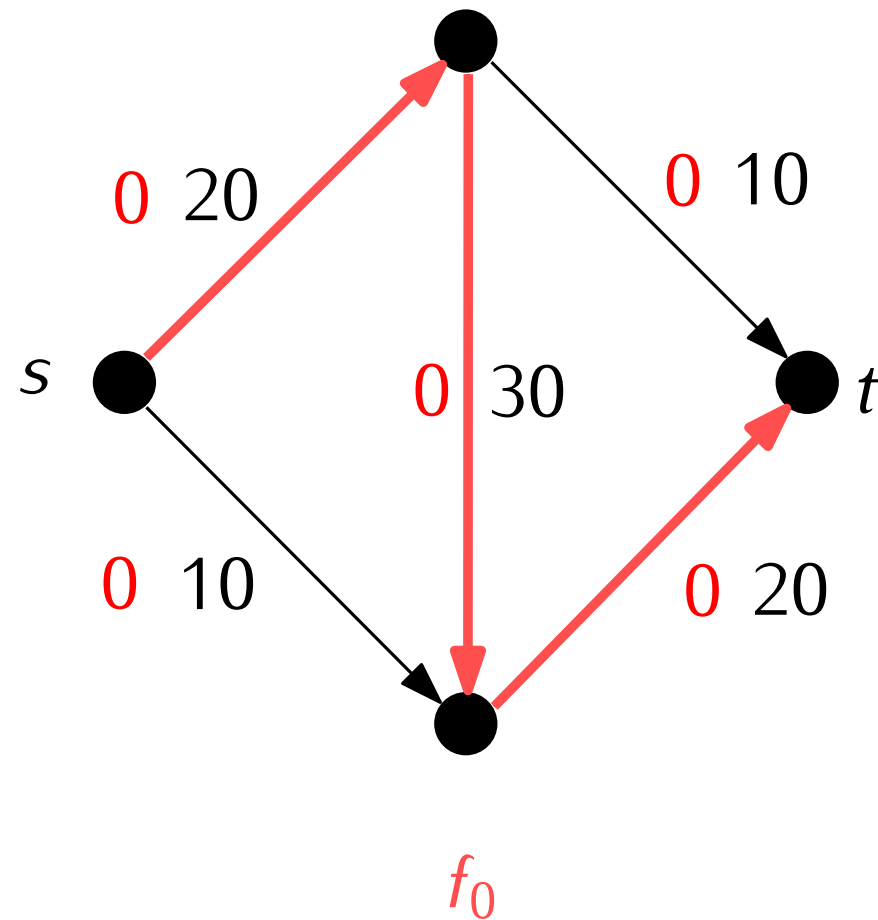
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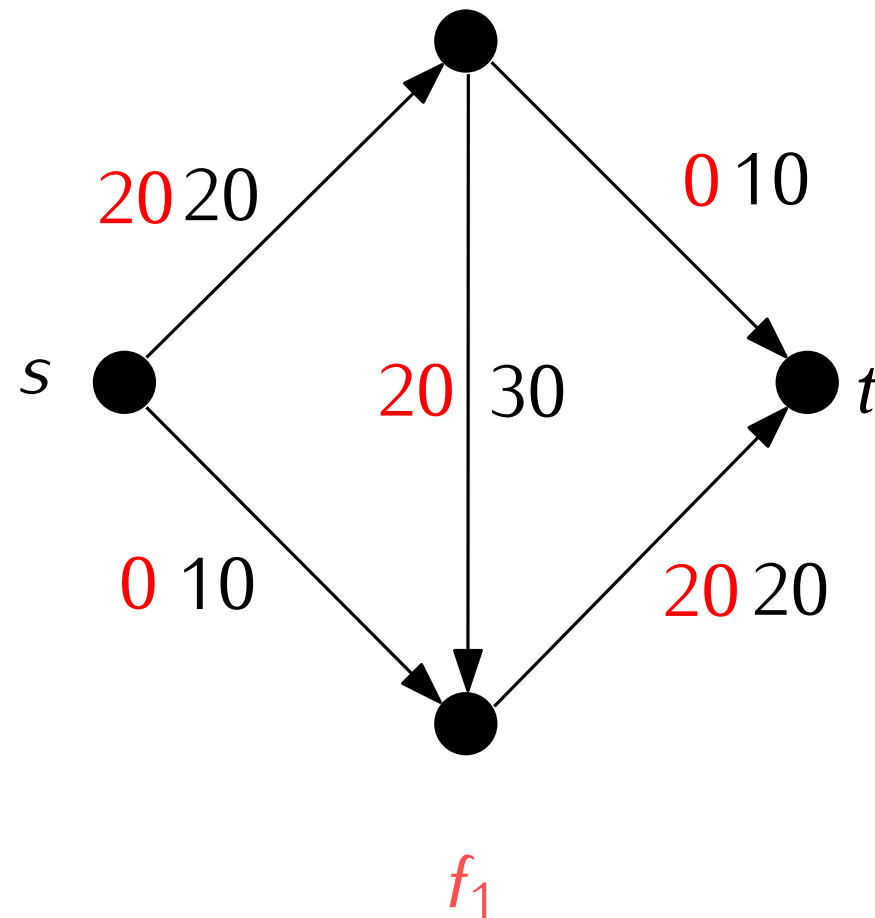
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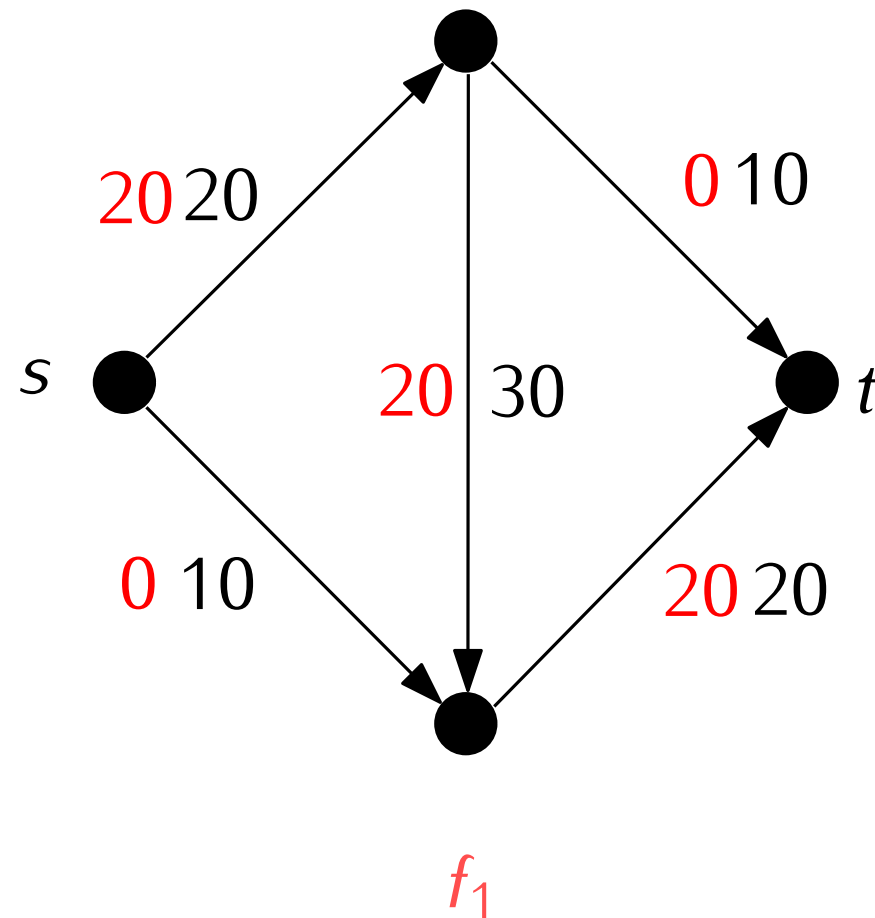
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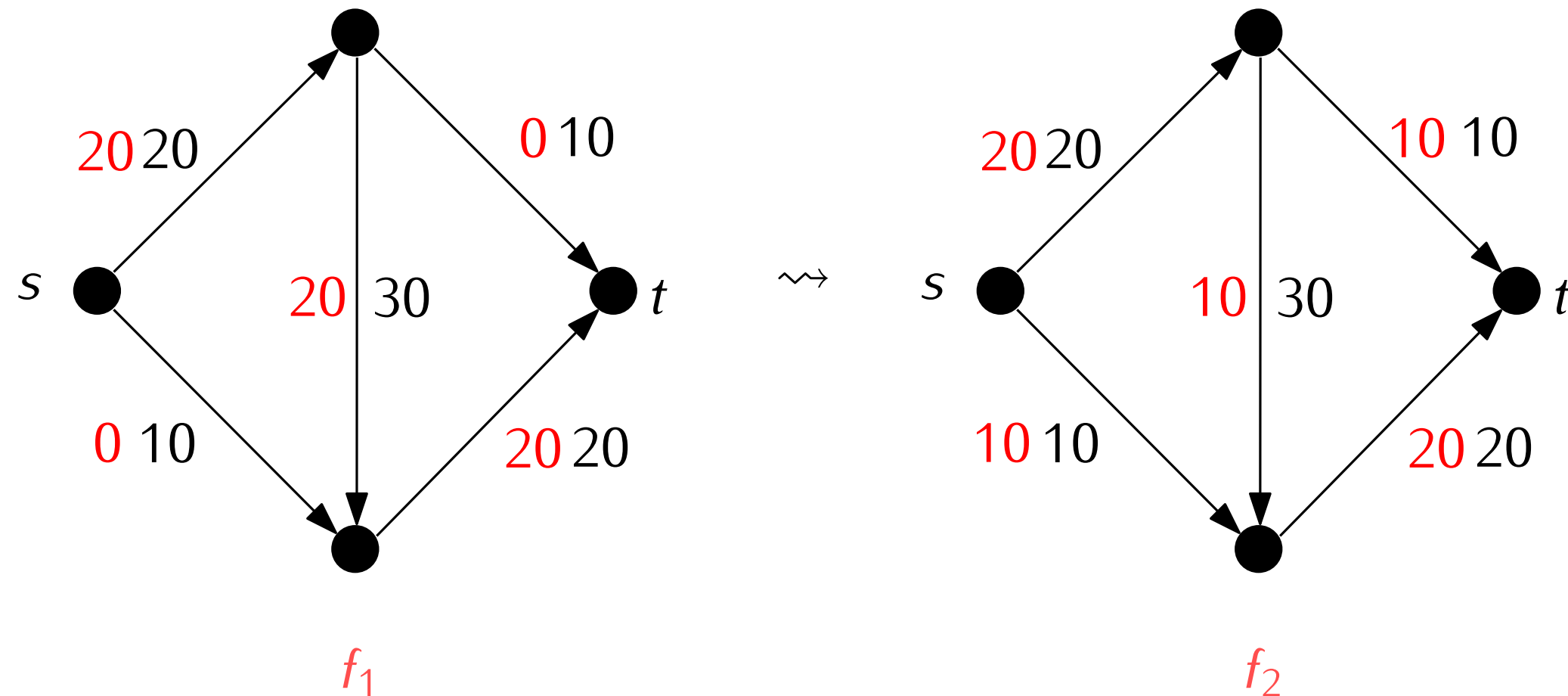
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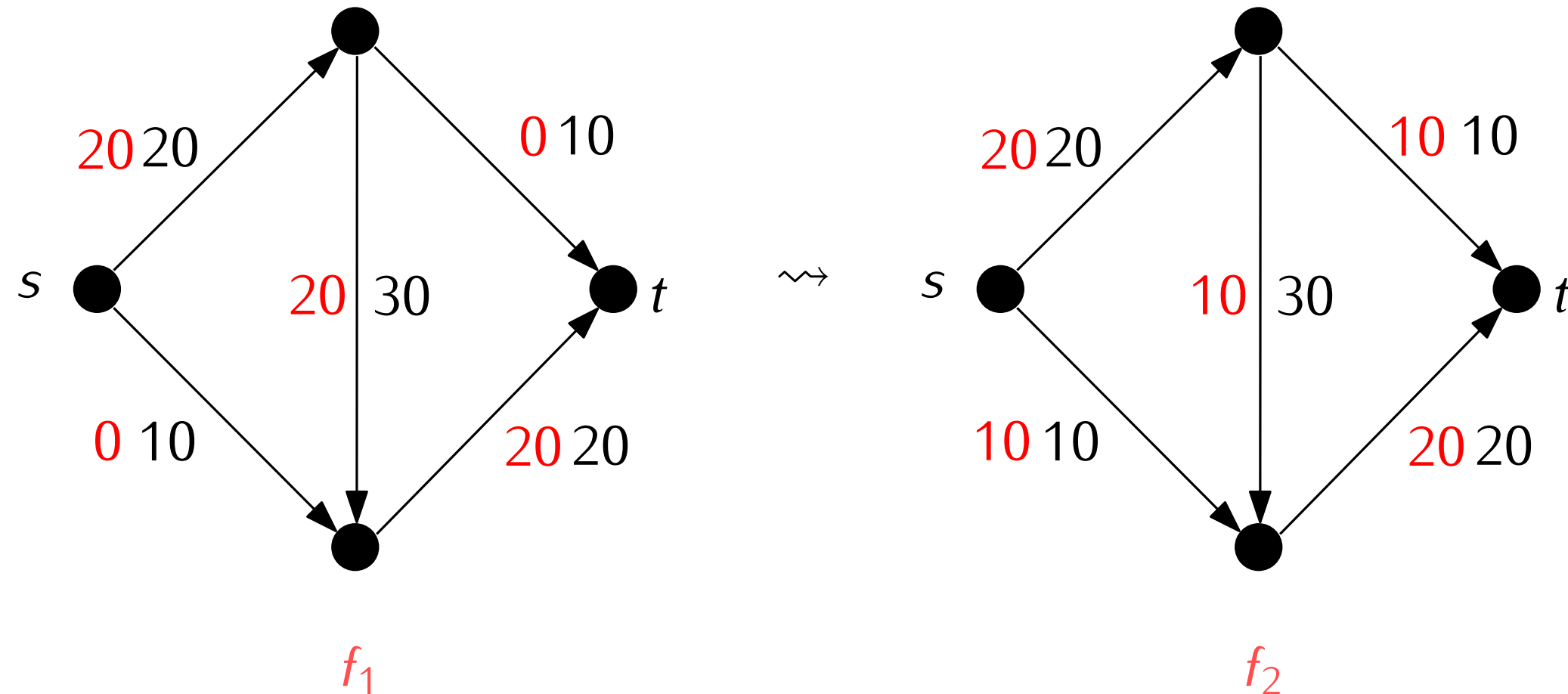
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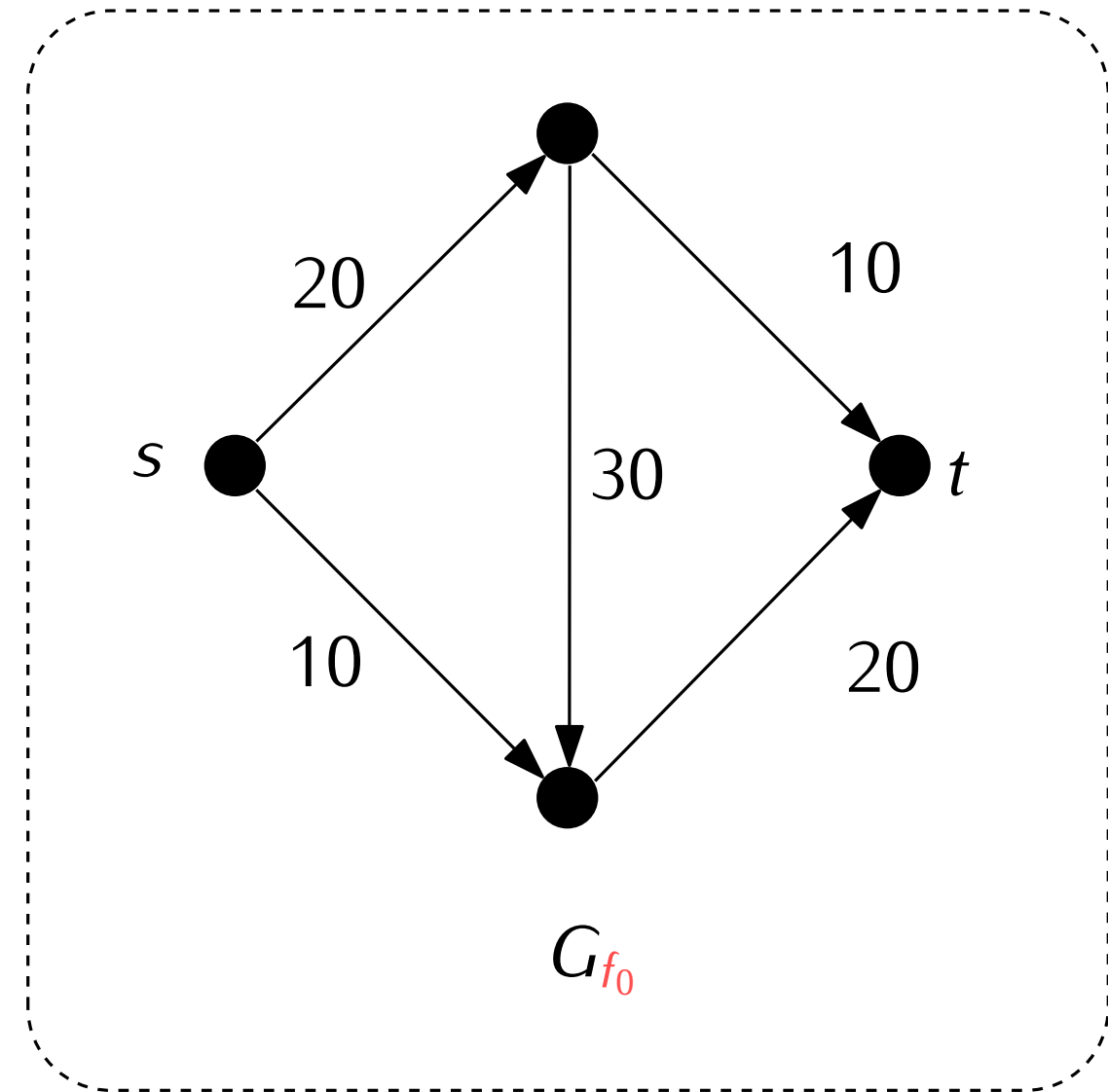
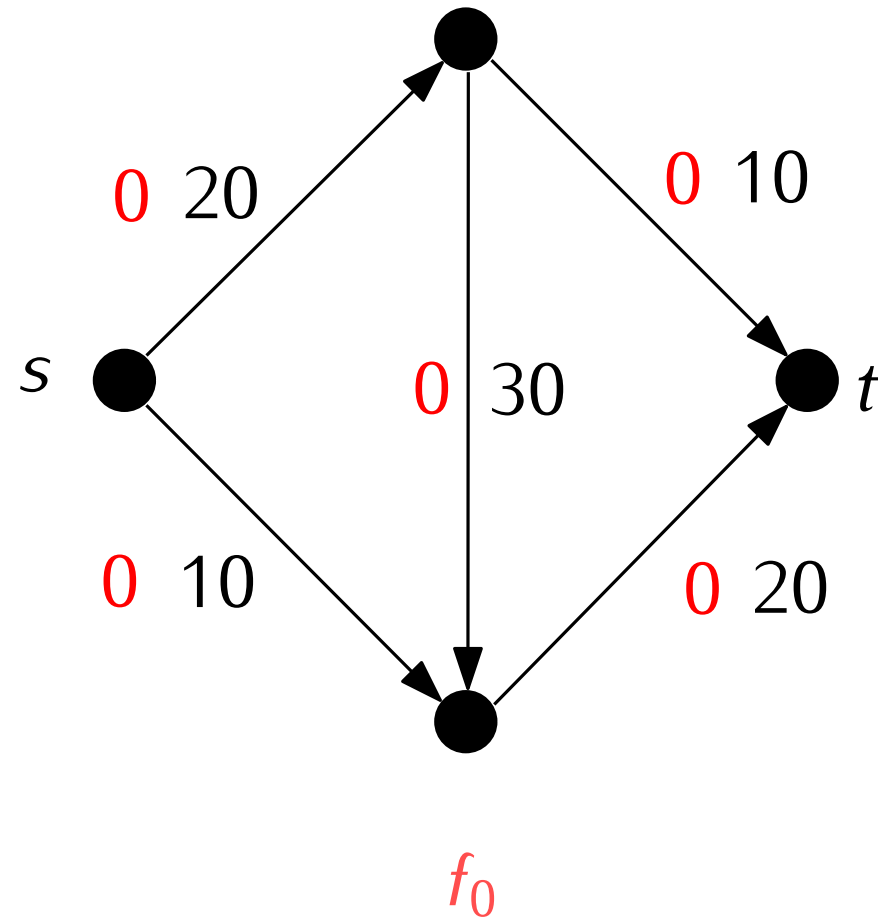
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- **Want:** data structure to find those steps efficiently



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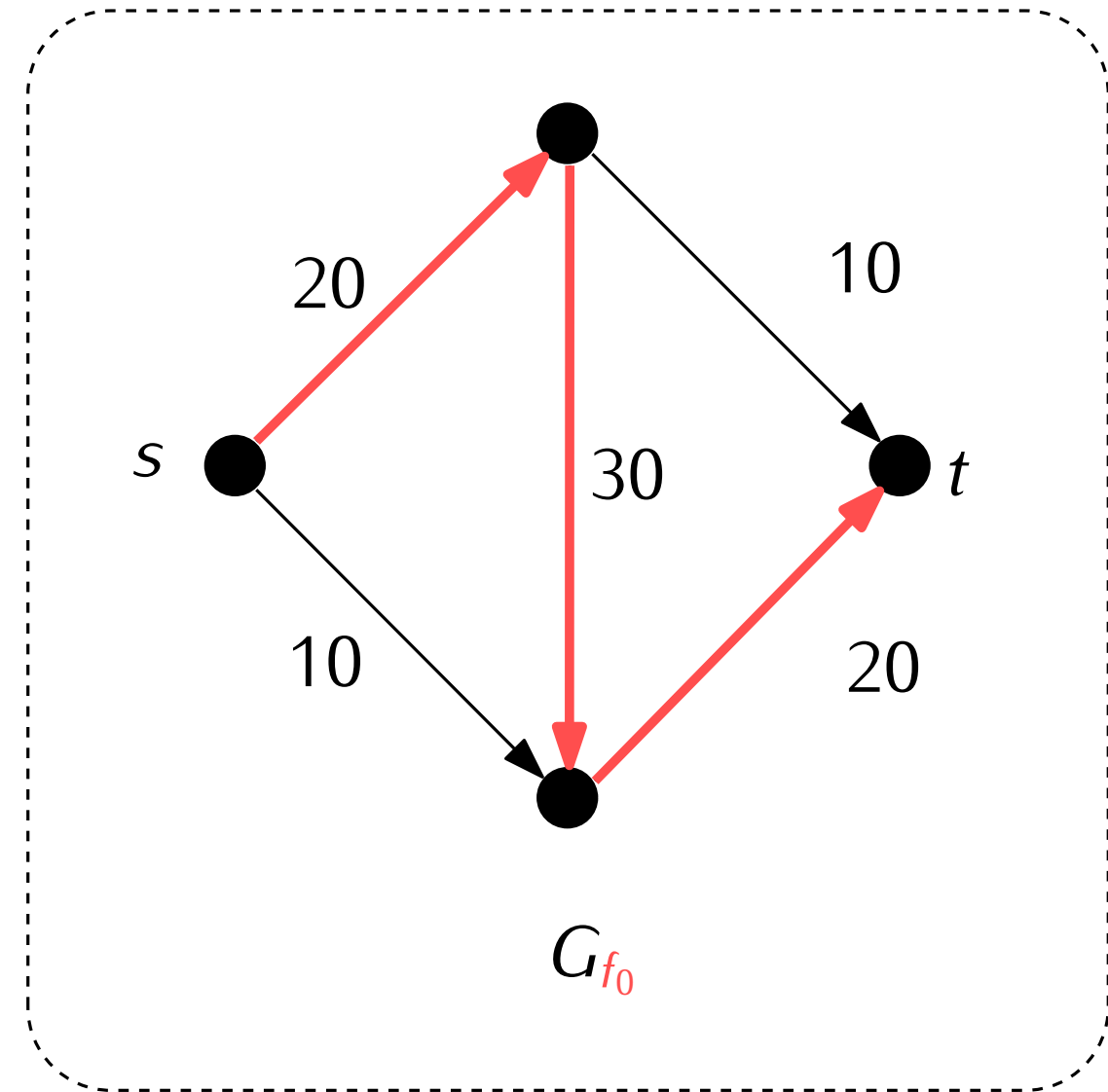
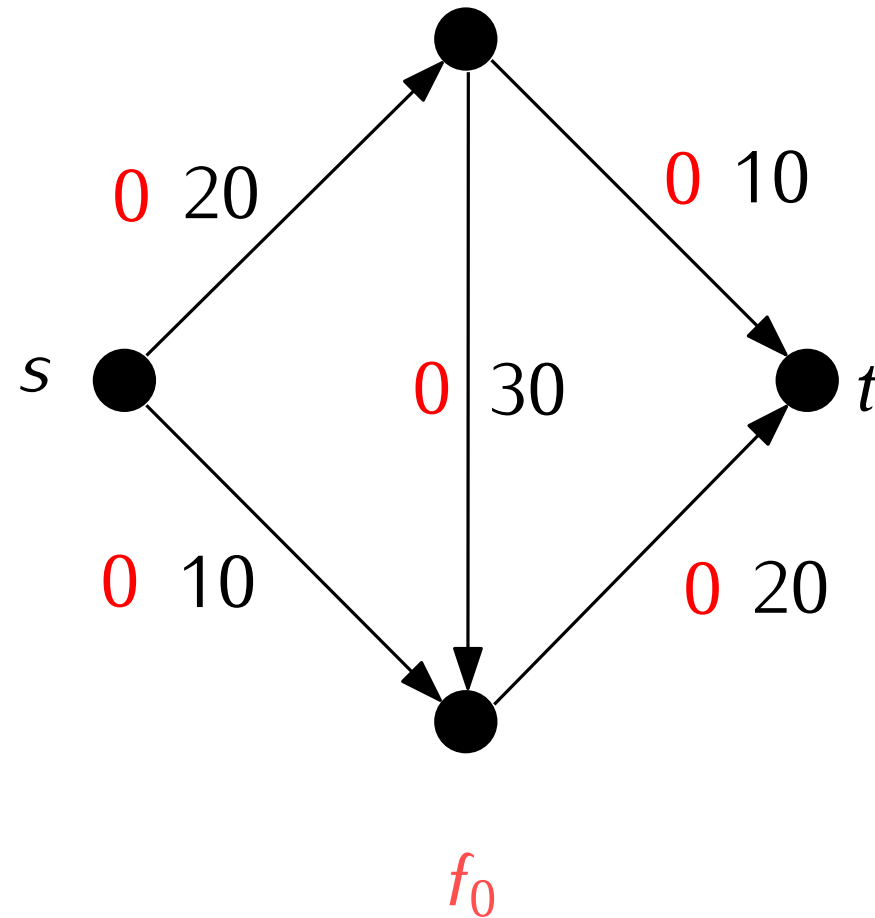
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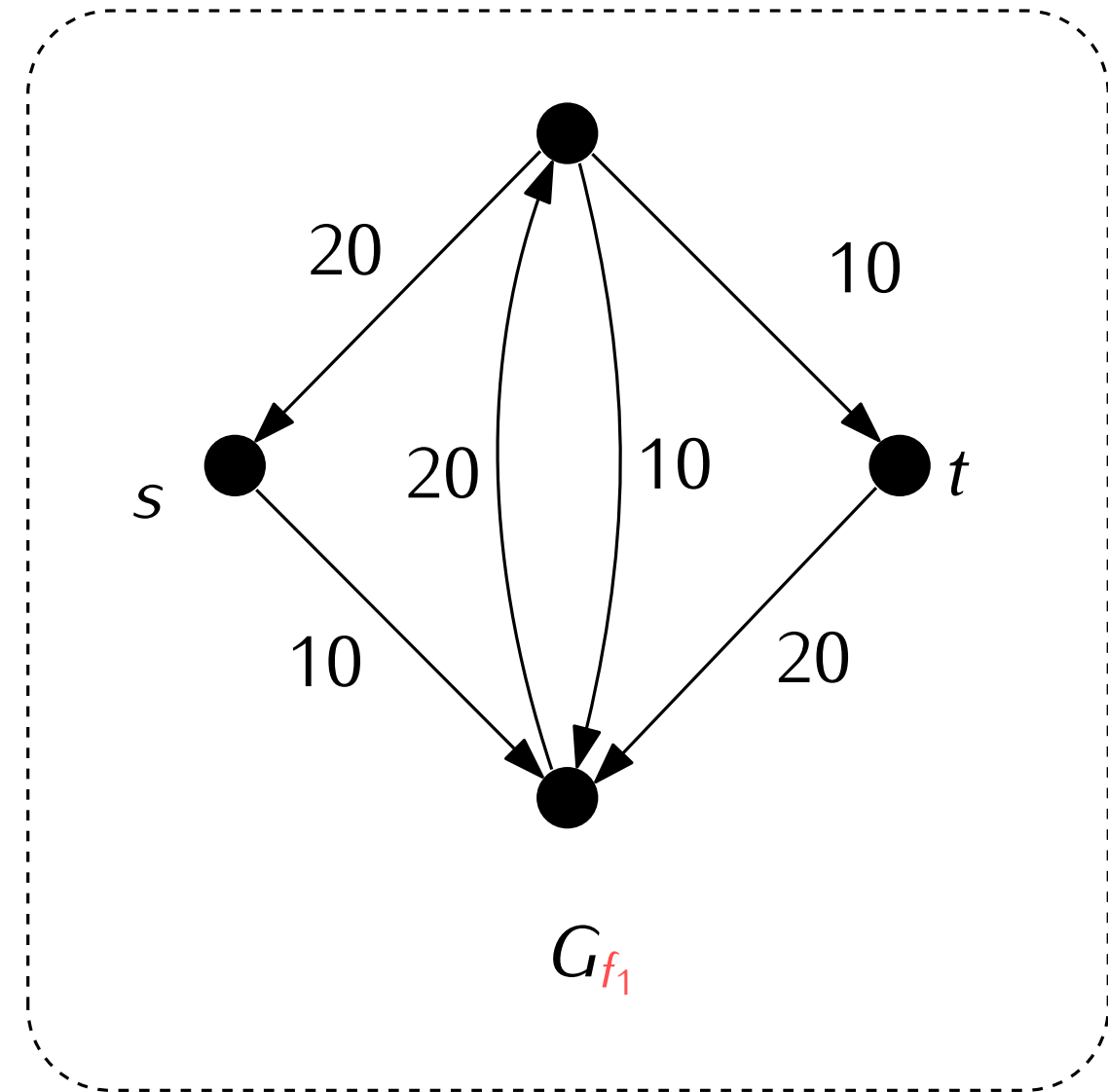
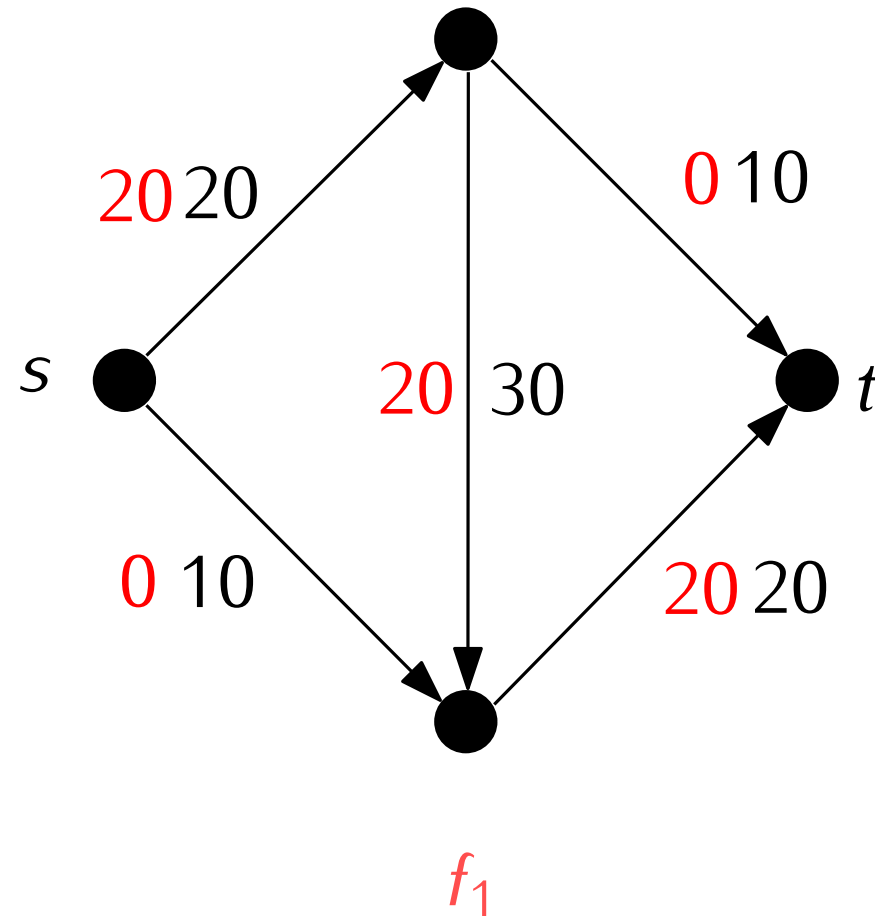
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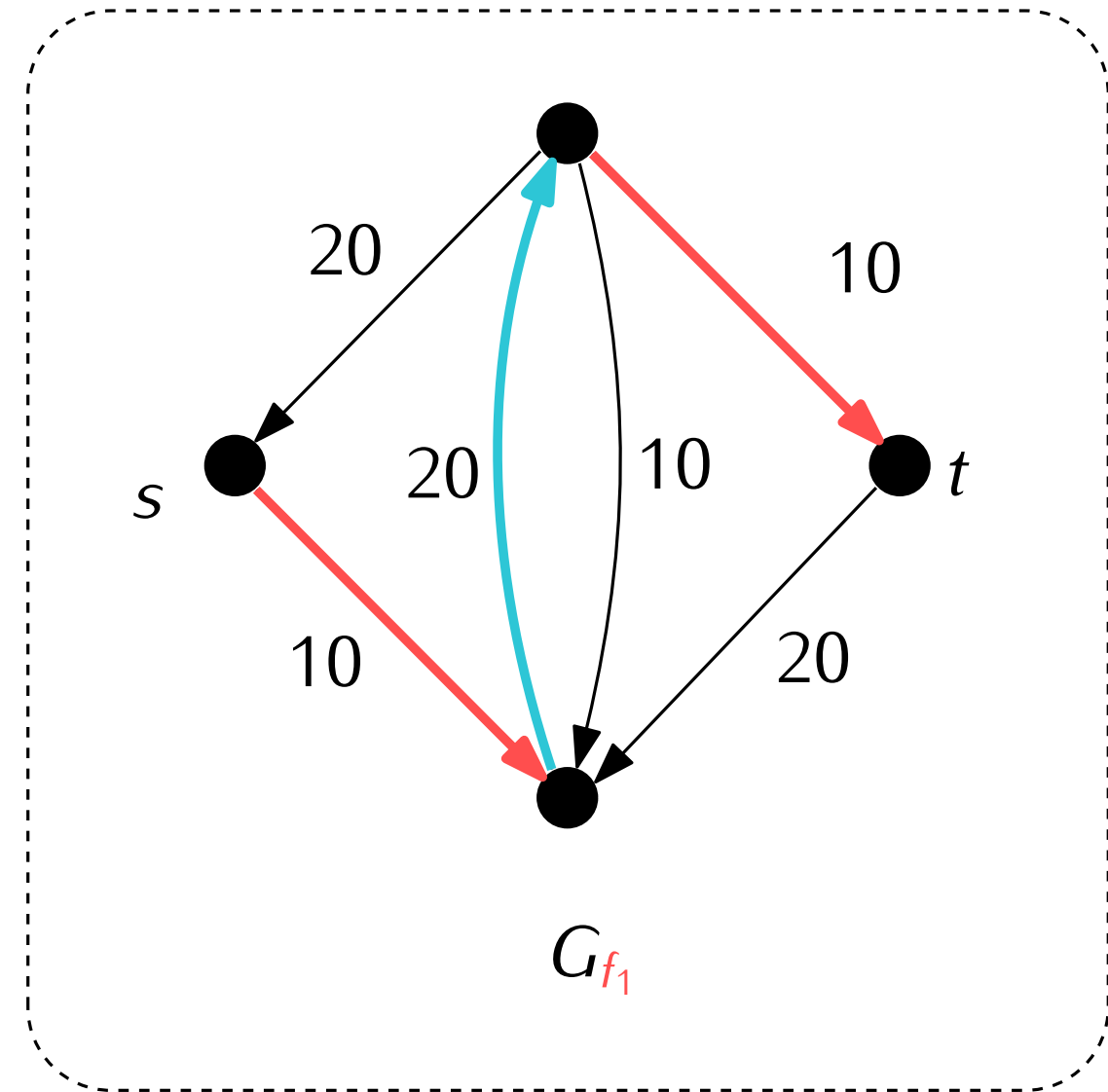
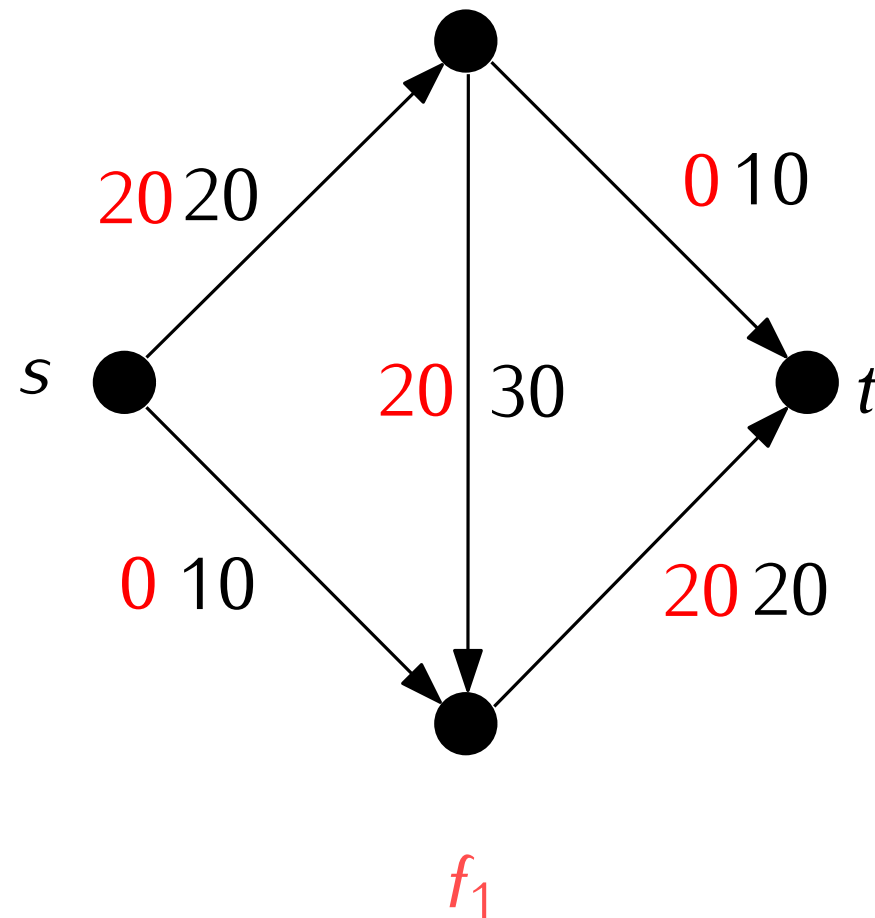
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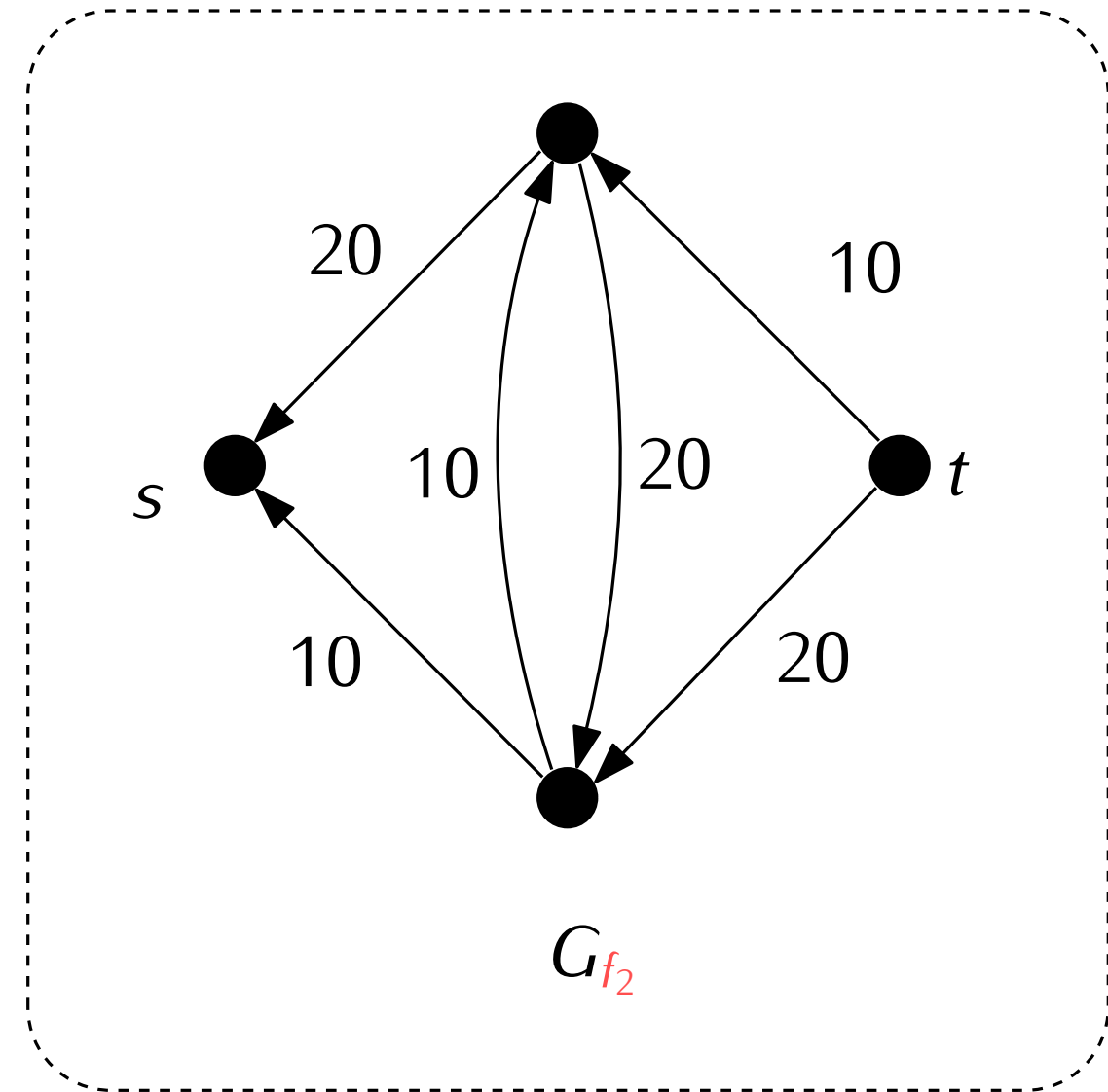
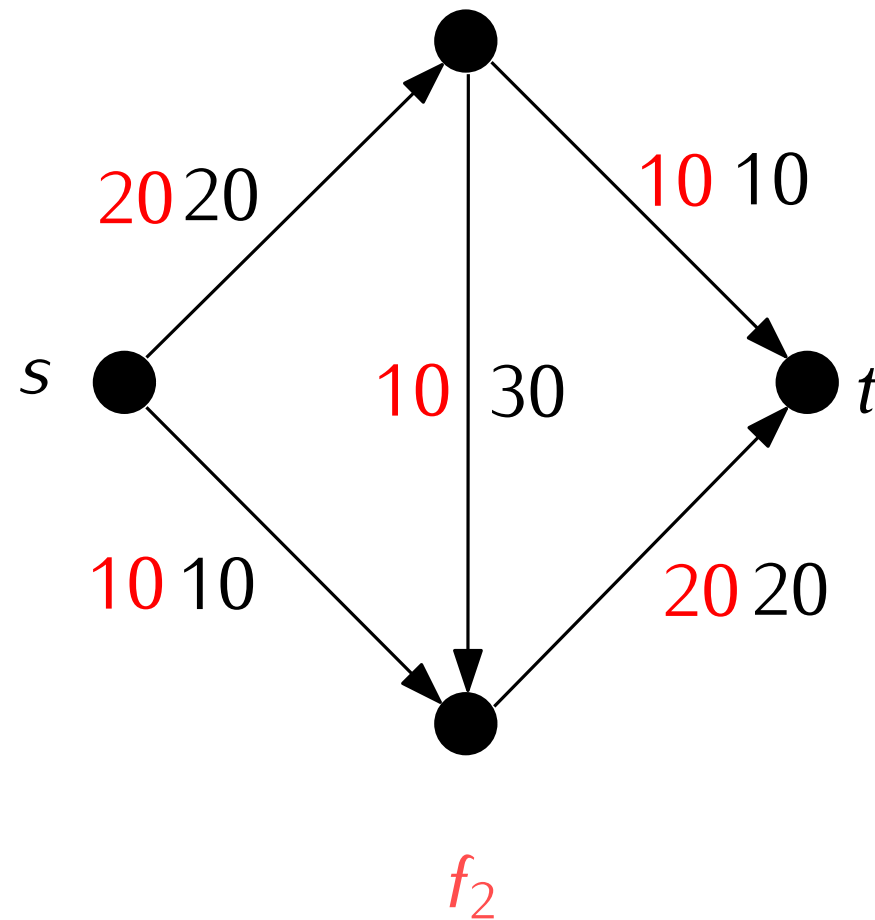
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 - $f'(e_i) := f(e_i) + m$ for all $i \in \{1, \dots, k\}$ if $e_i \in E$
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Questions:

- Does it terminate? 2. could loop forever...
- How fast does it terminate?
- Is f a flow?
- Does f always have maximal value?

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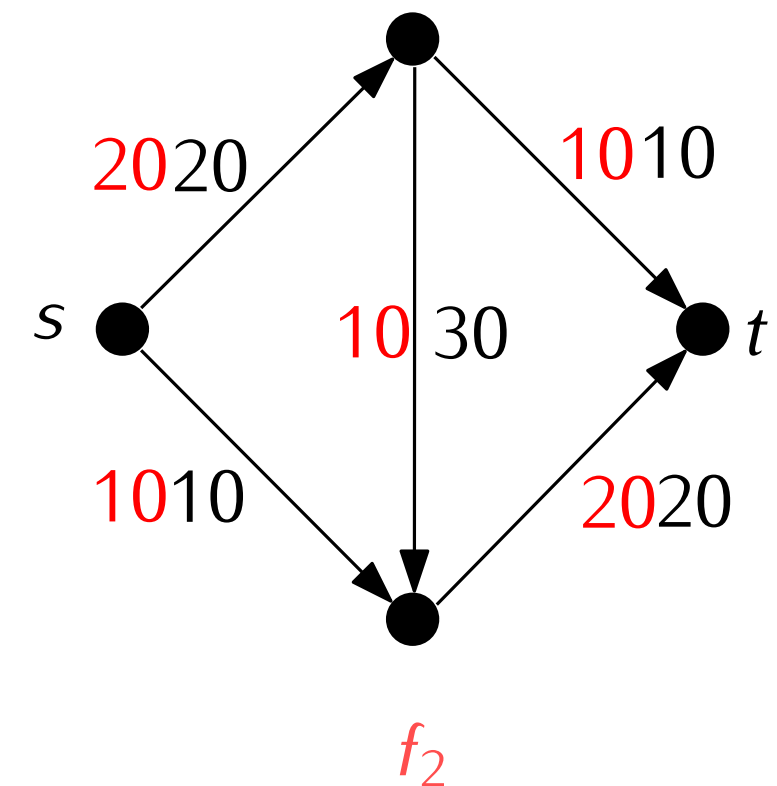
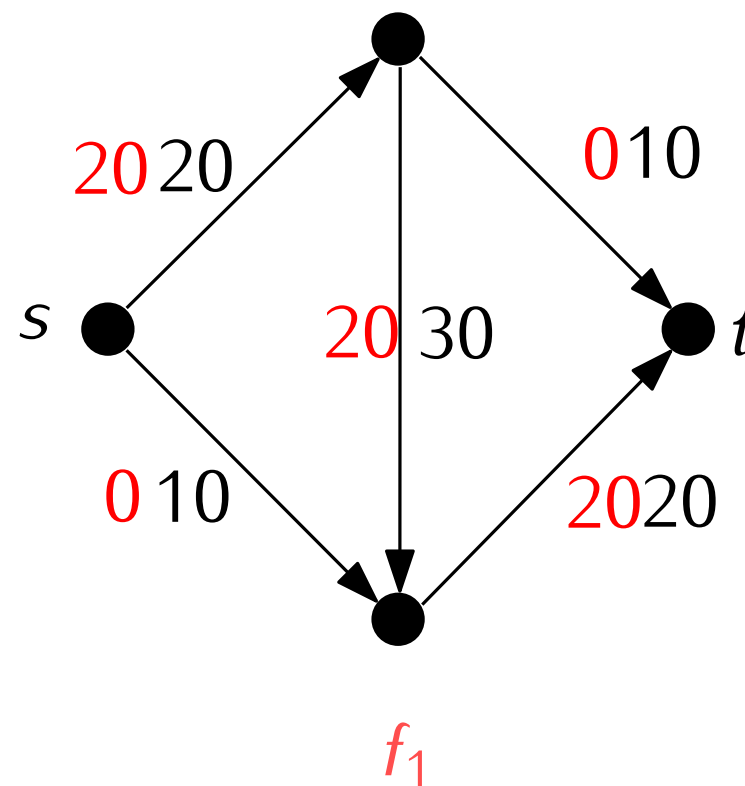
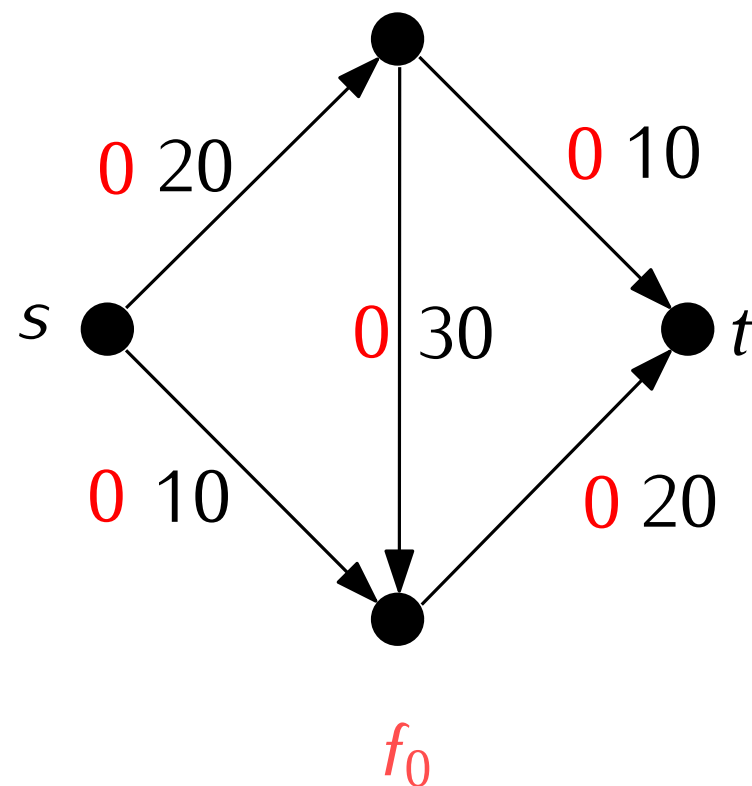
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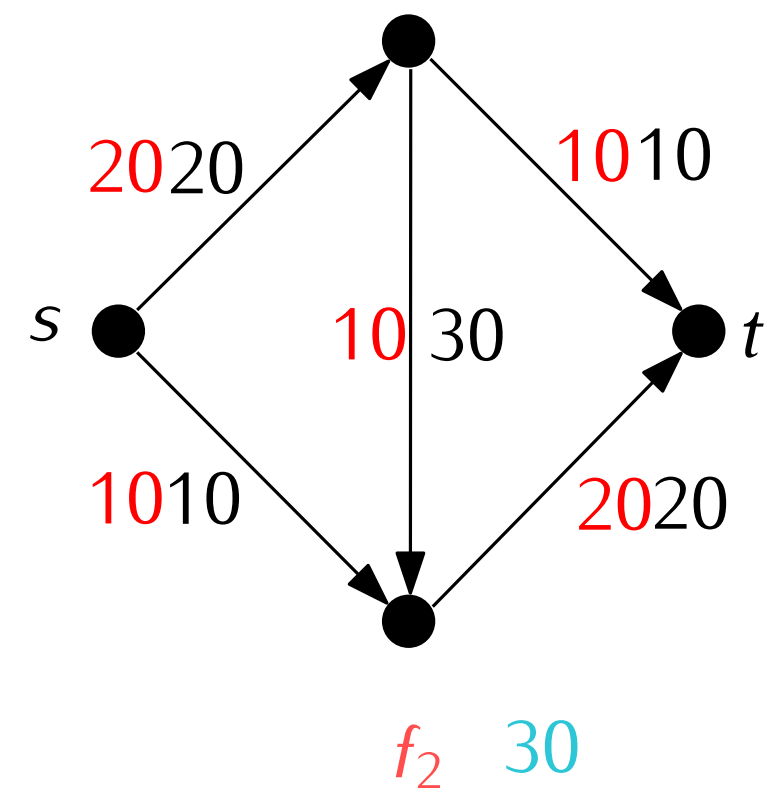
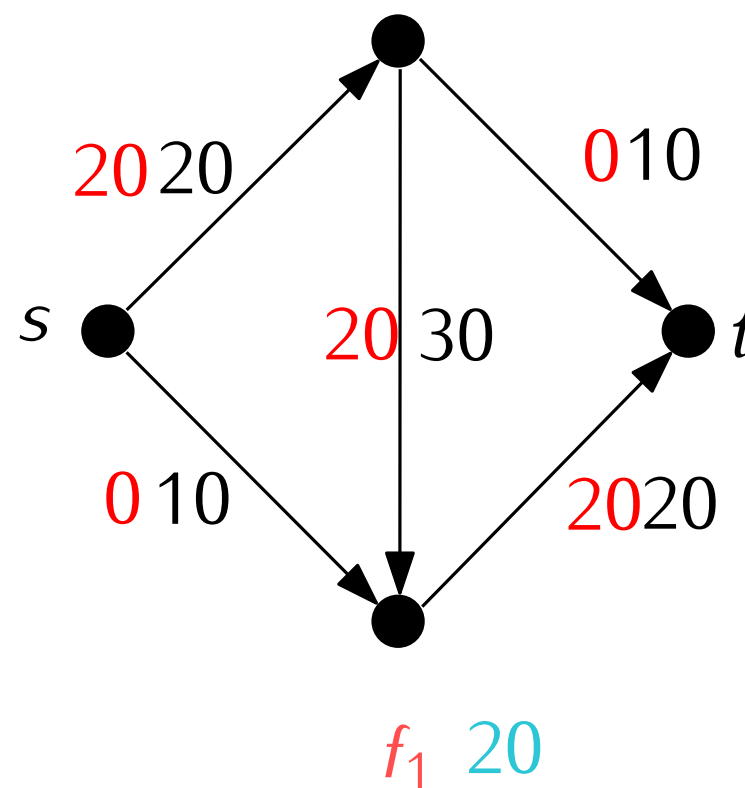
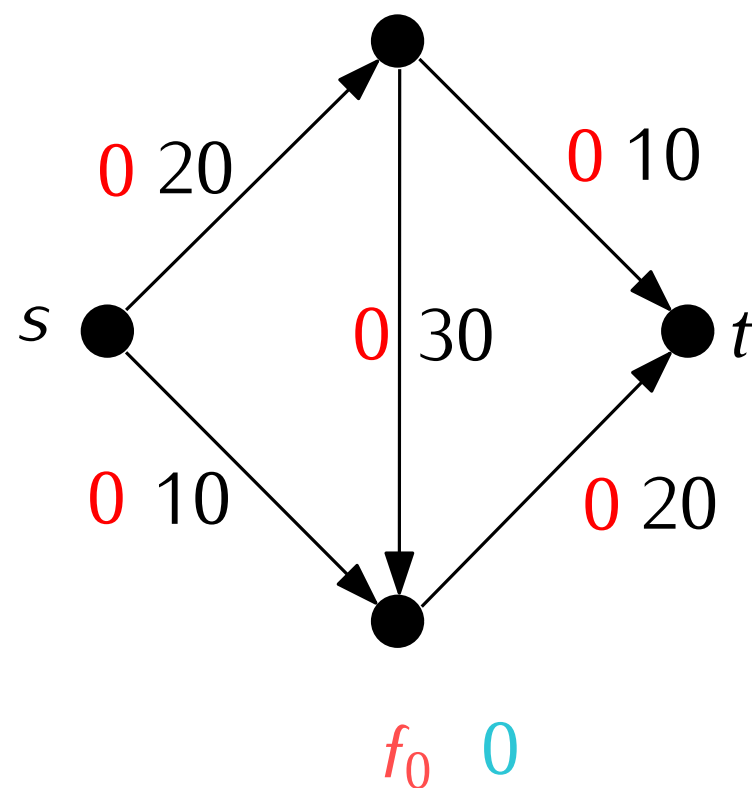


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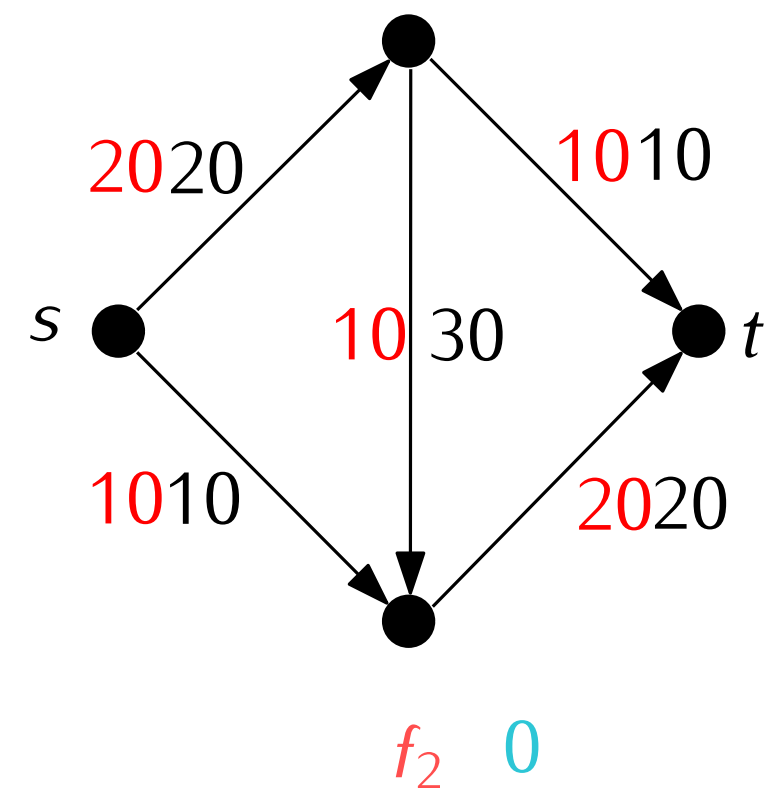
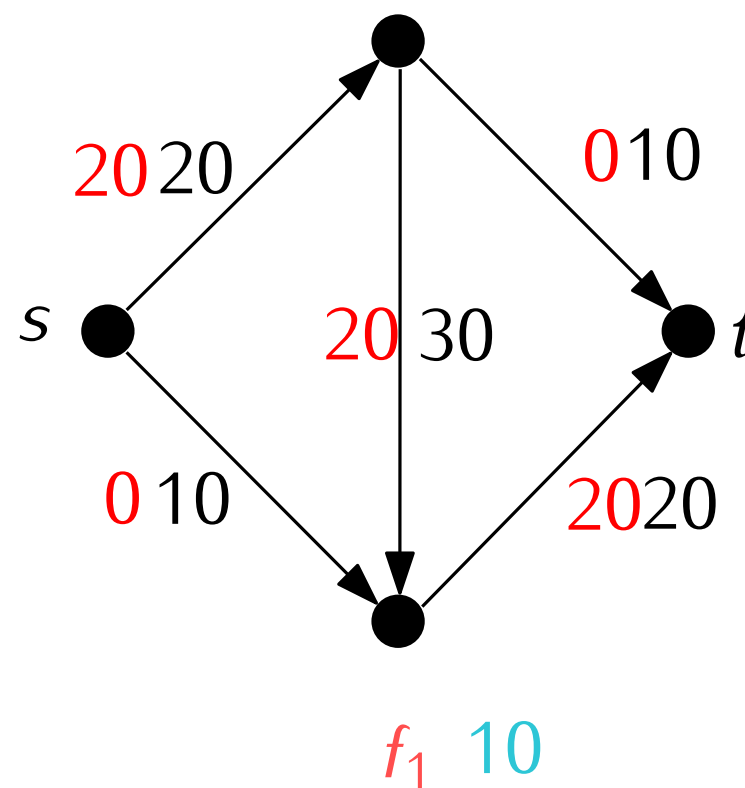
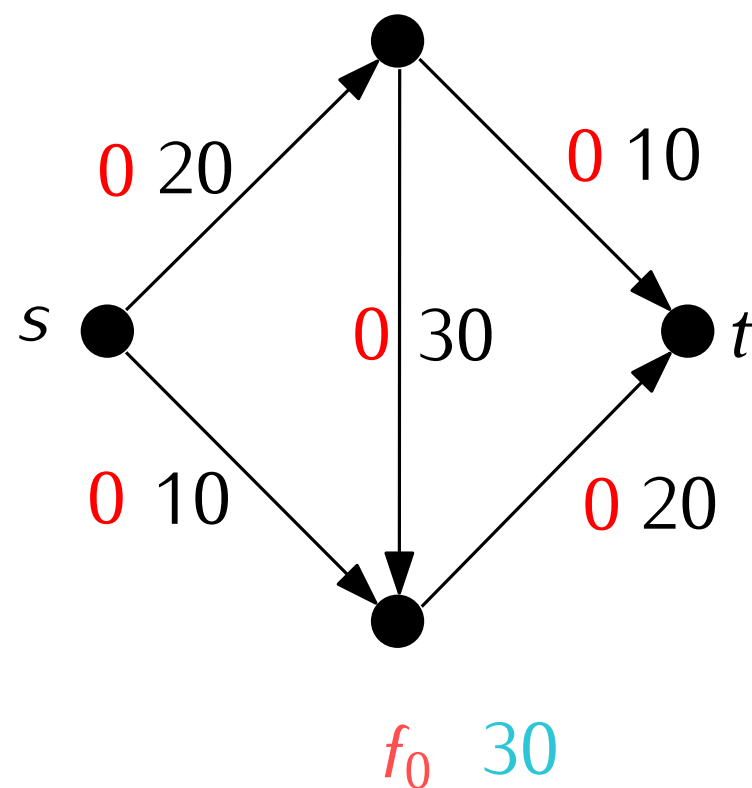


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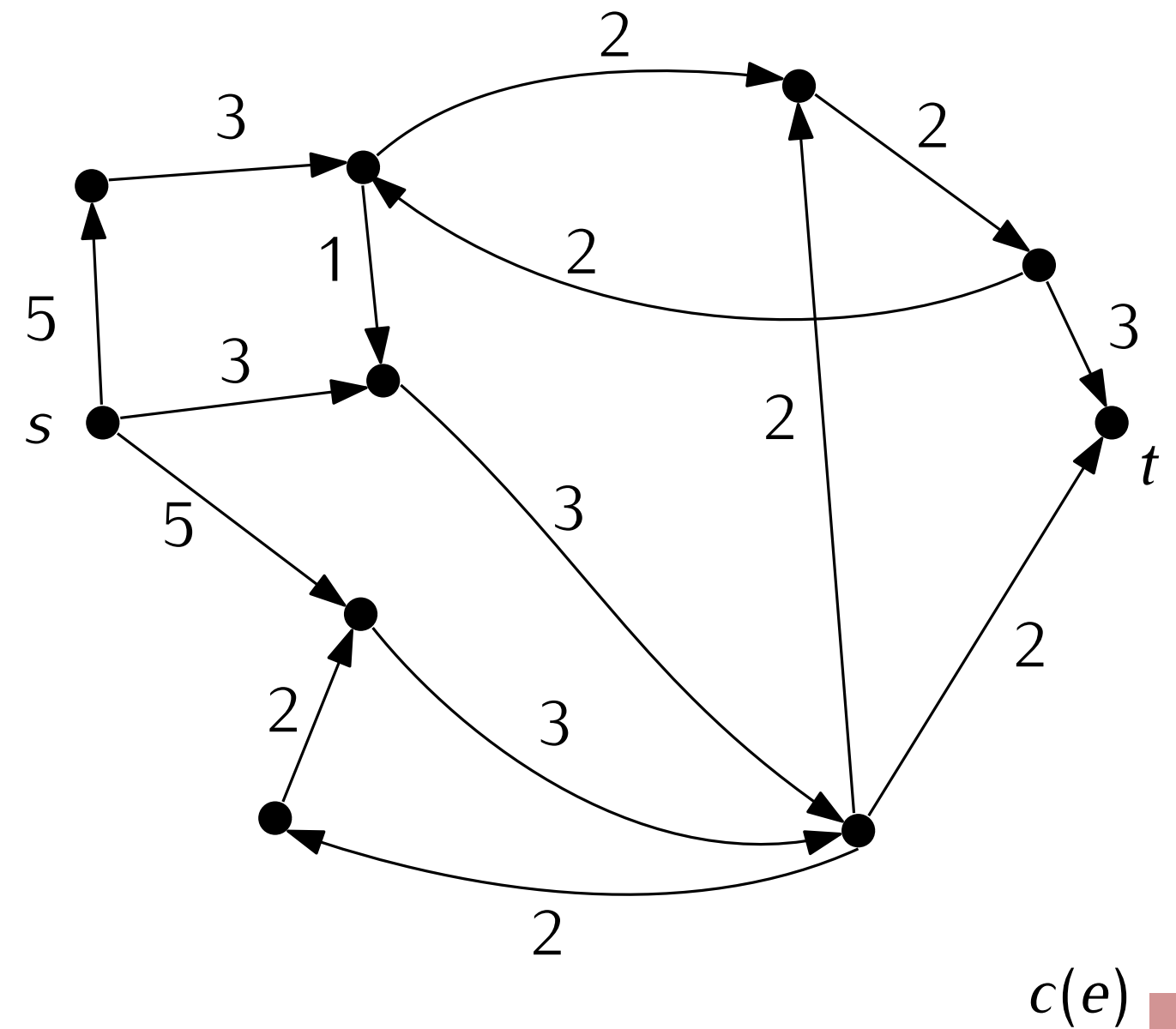
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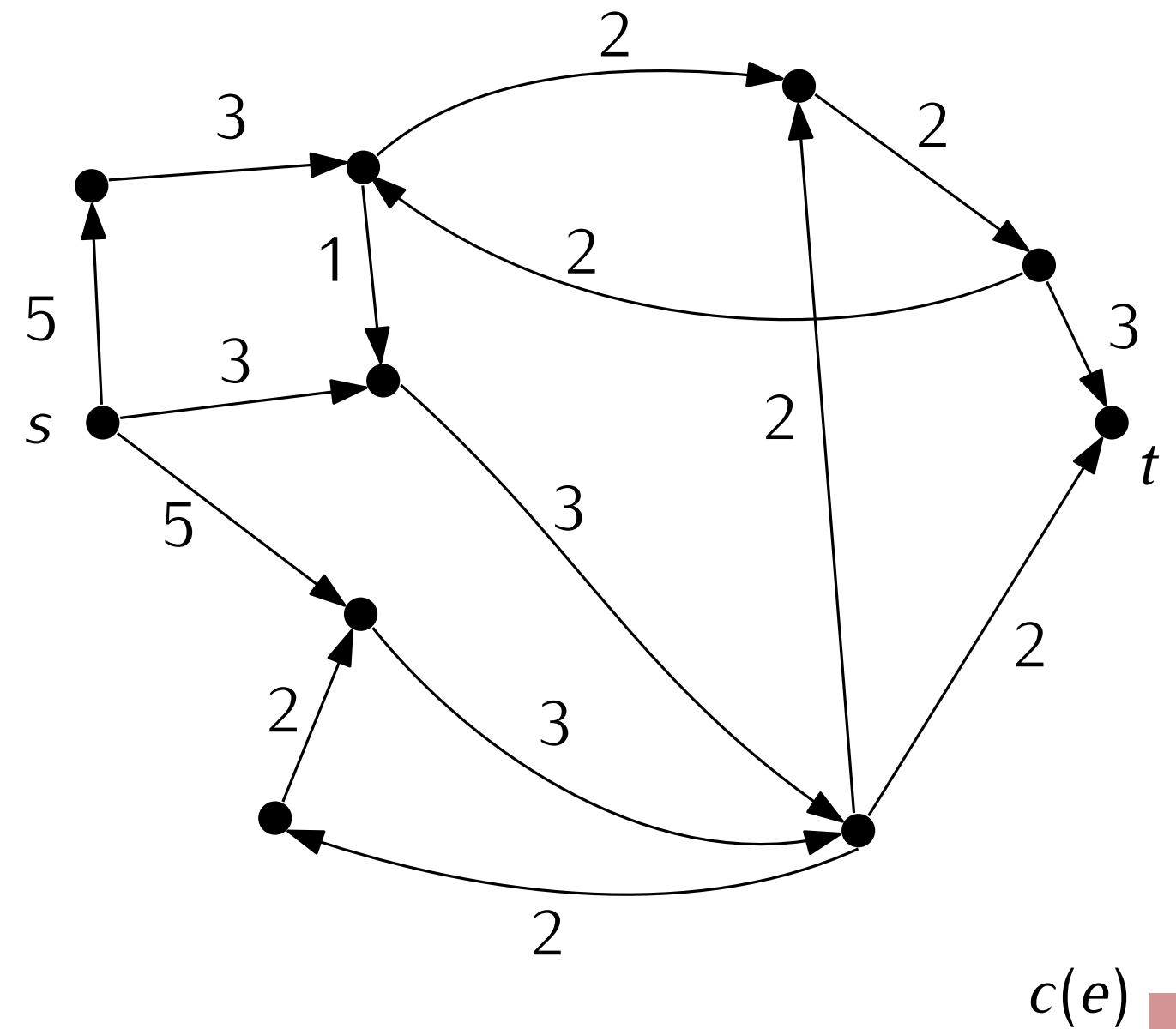
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- We later see that there are at most $|V||E|$ iterations if we choose shortest path at 2.
 \rightsquigarrow Edmonds-Karp algorithm

What are values that can in principle be achieved?



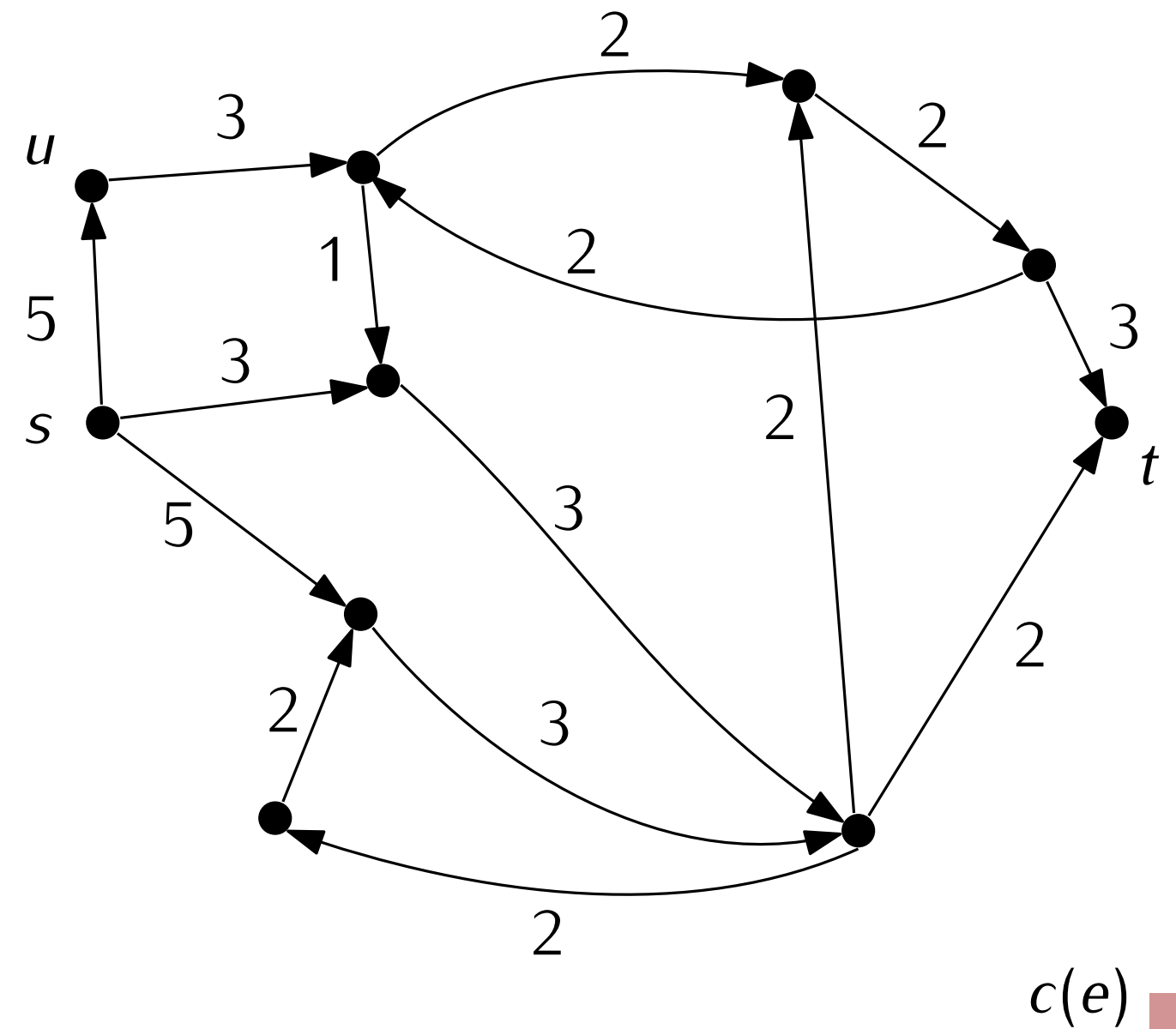
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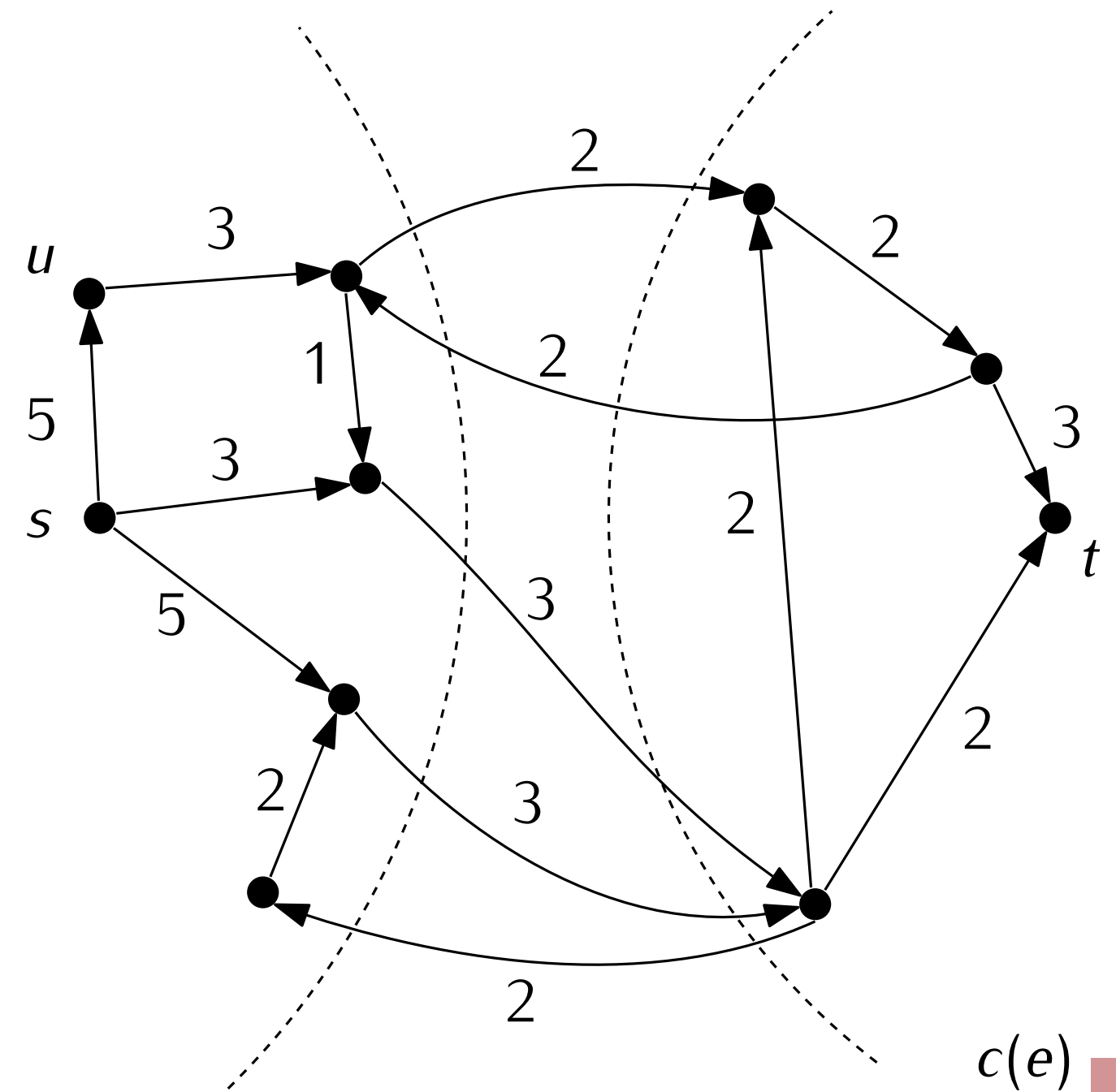
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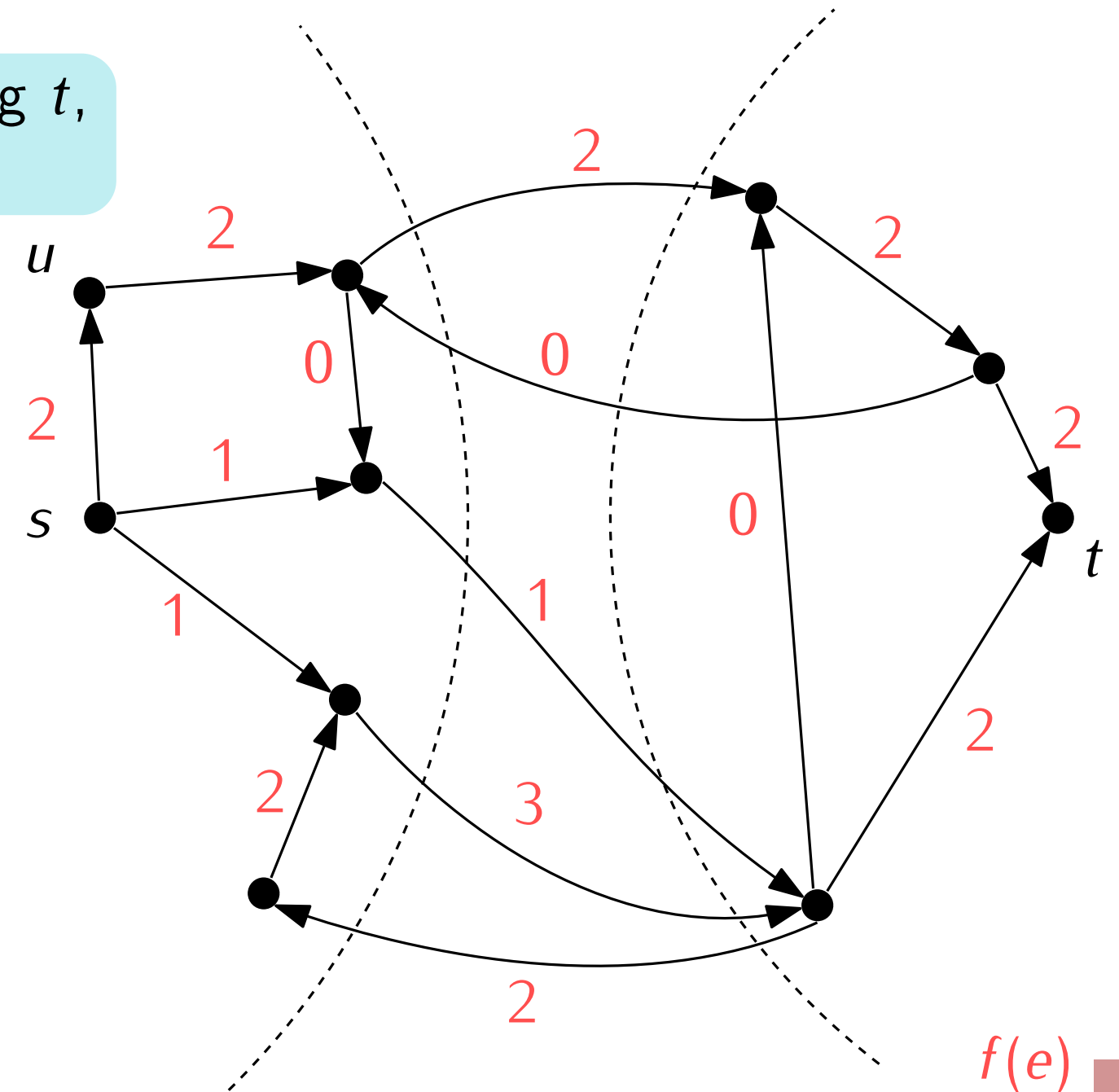
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Lemma. For any $A \subseteq V$ containing s , $B := V \setminus A$ containing t , the value of f is equal to $\sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e)$.

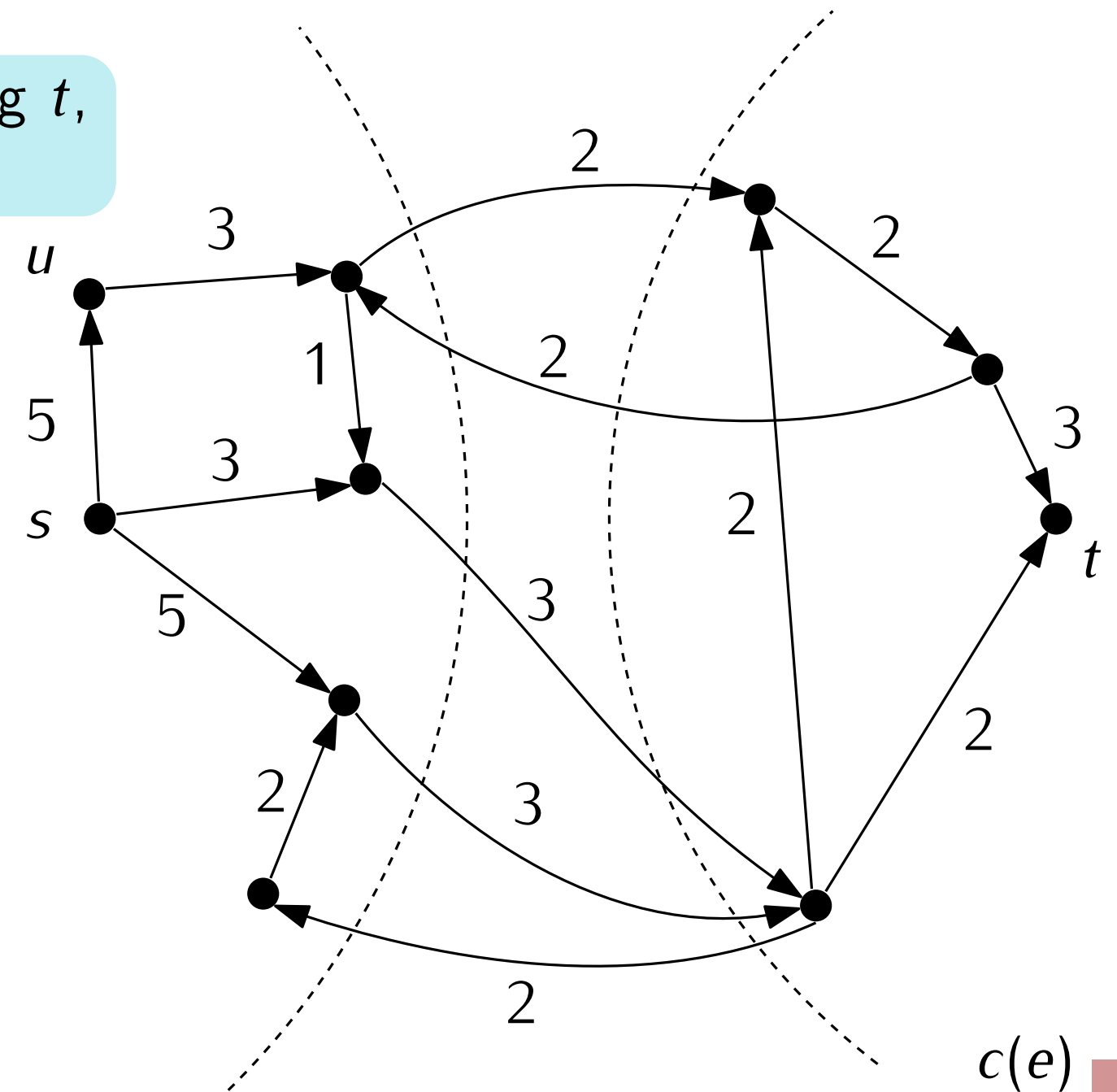


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Lemma. For any $A \subseteq V$ containing s , $B := V \setminus A$ containing t , the value of f is equal to $\sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e)$.

\leadsto the maximal value of a flow is at most $\sum_{e \text{ out of } A} c(e)$.



What are values that can in principle be achieved?

- Value is $\leq \sum_{e \text{ out of } s} c(e) = 13$
- Value is $\leq \sum_{e \text{ out of } \{s,u\}} c(e) = 11$
- Value is ≤ 8

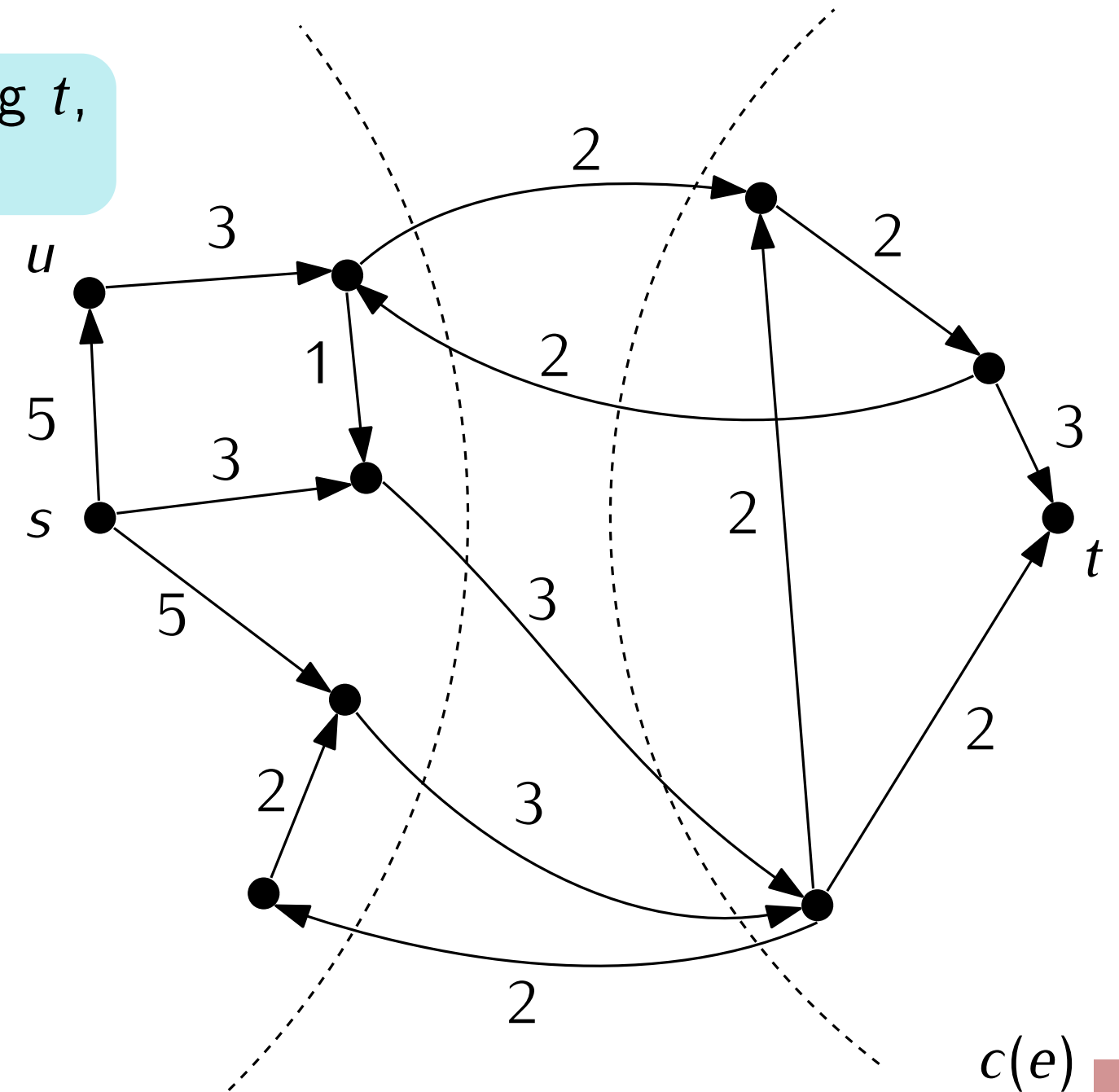
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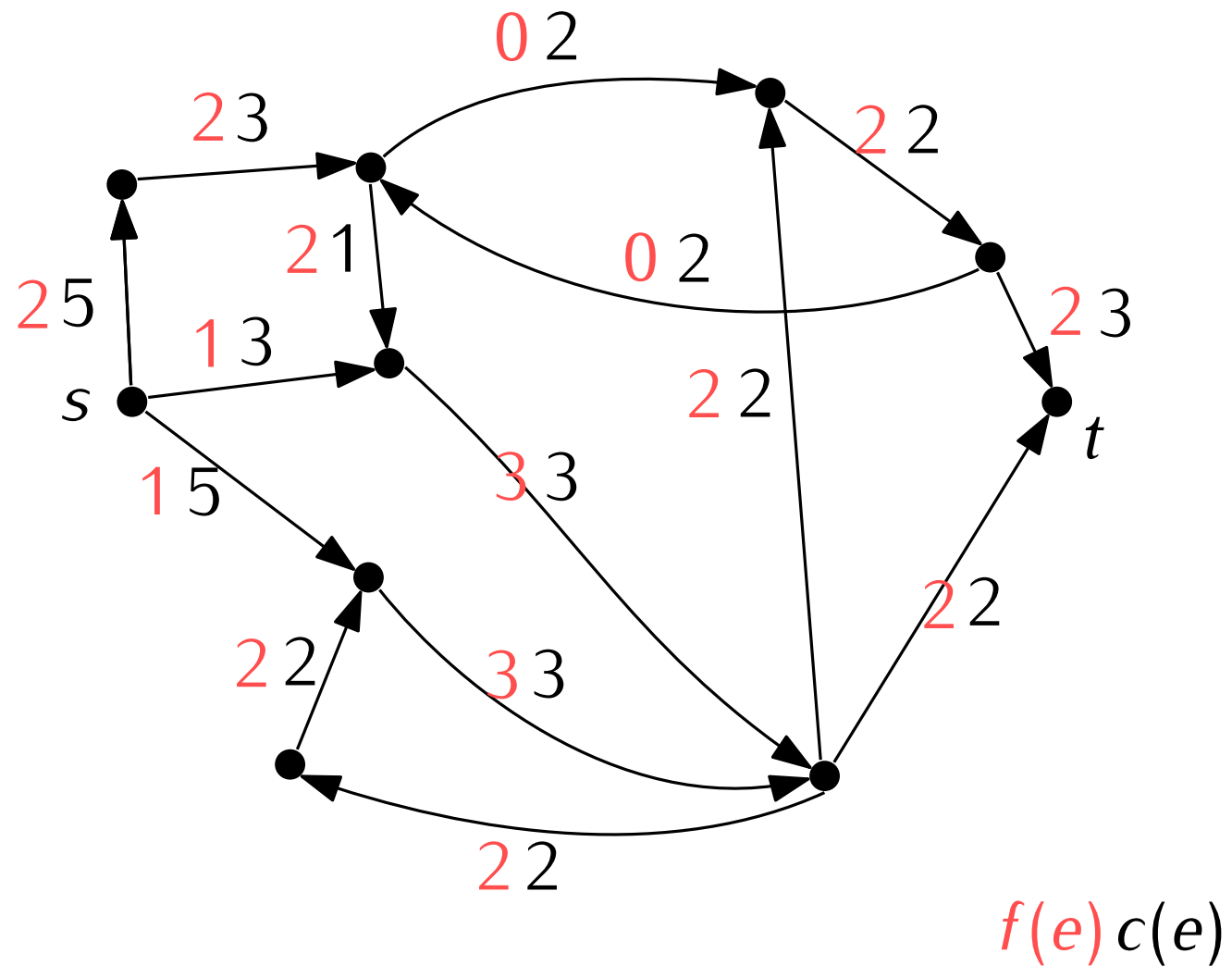
Since this holds for **all** s - t cuts (A, B) :

Theorem. For all G , we have

$$\text{maxflow}(G) \leq \min_{(A,B) \text{ s-t cut}} \sum_{e \text{ out of } A} c(e)$$

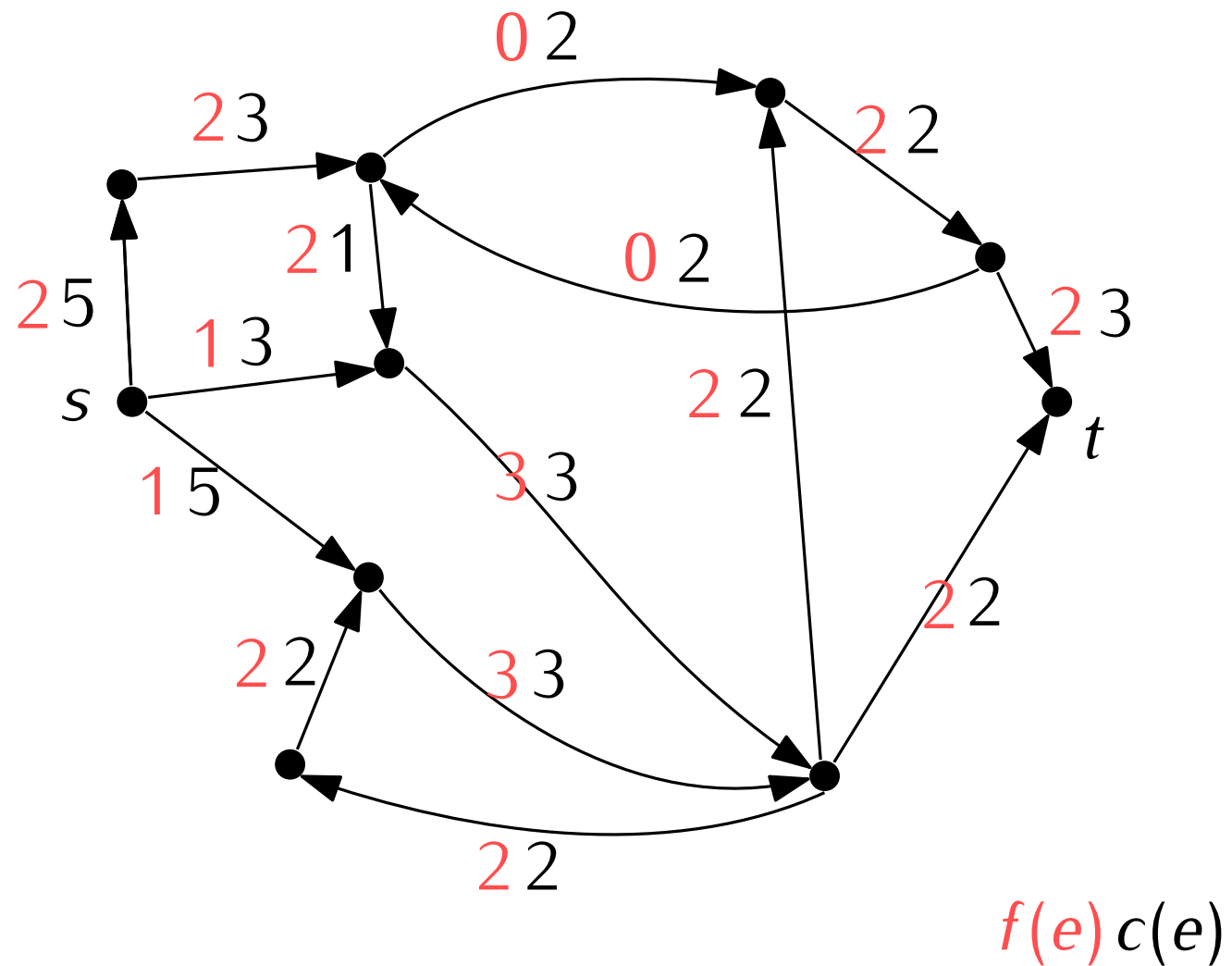


How can we know that a flow f is maximal?



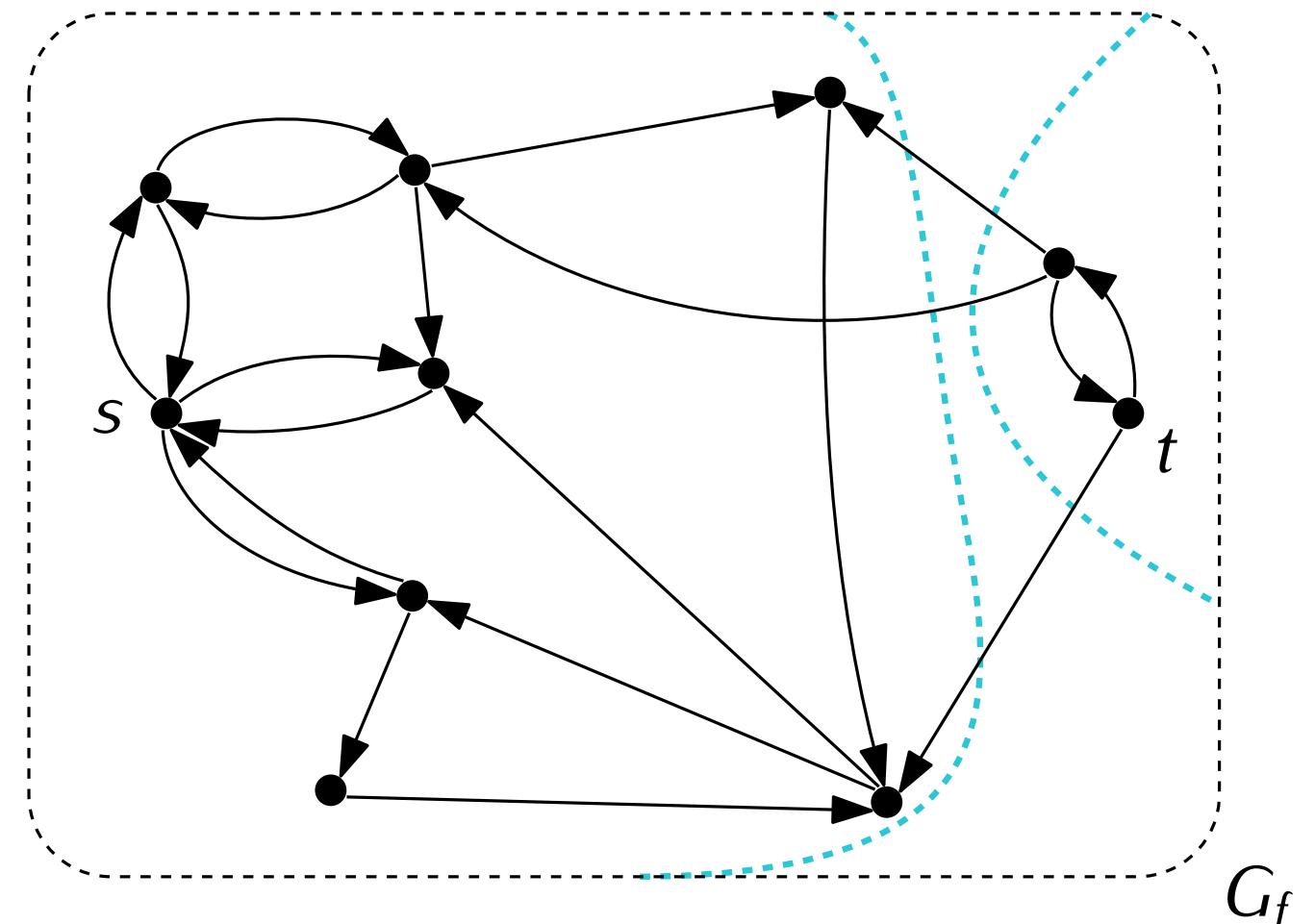
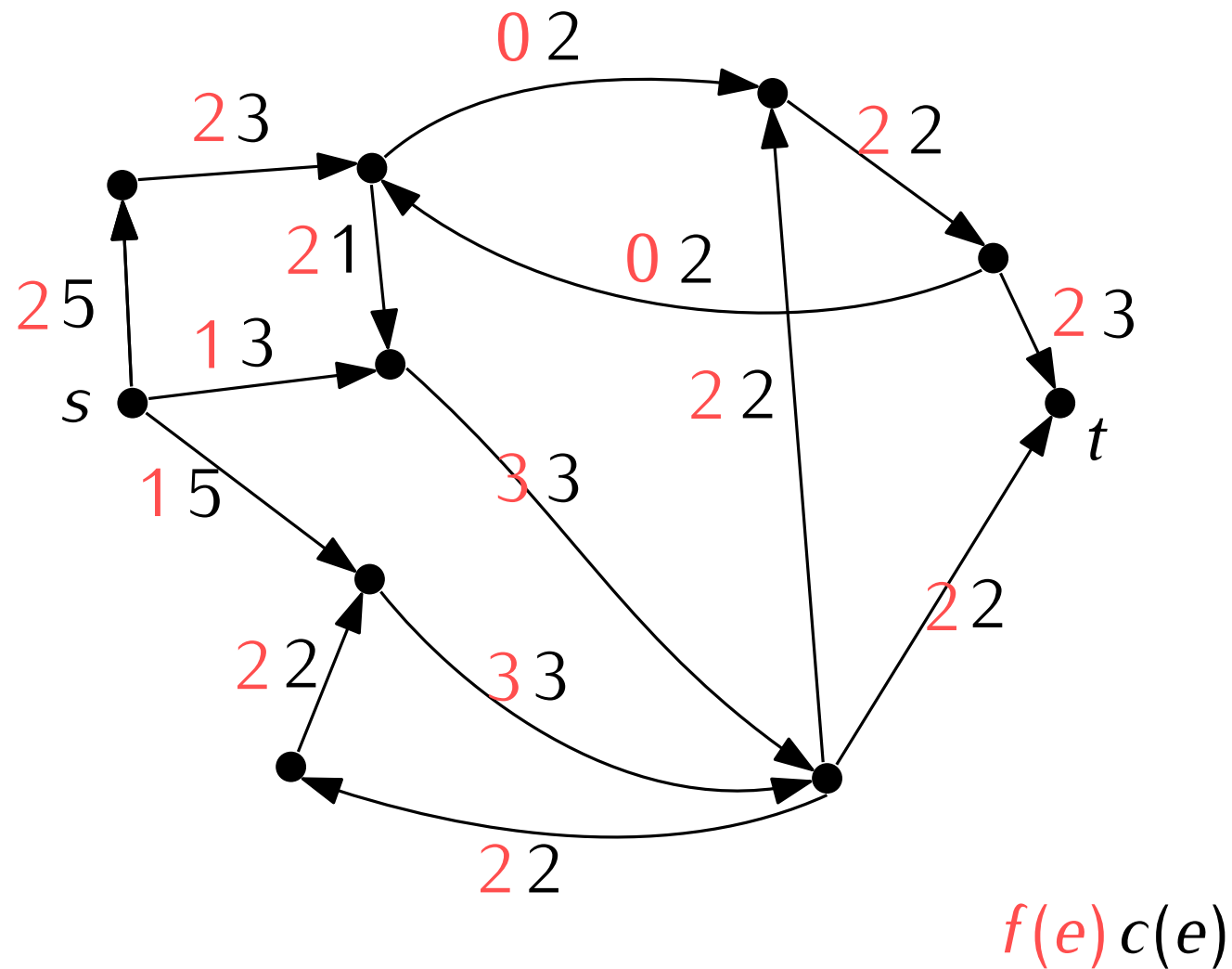
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- It is enough to give a cut (A, B) such that $\sum_{e \text{ out of } A} c(e) \leq \text{value of } f$



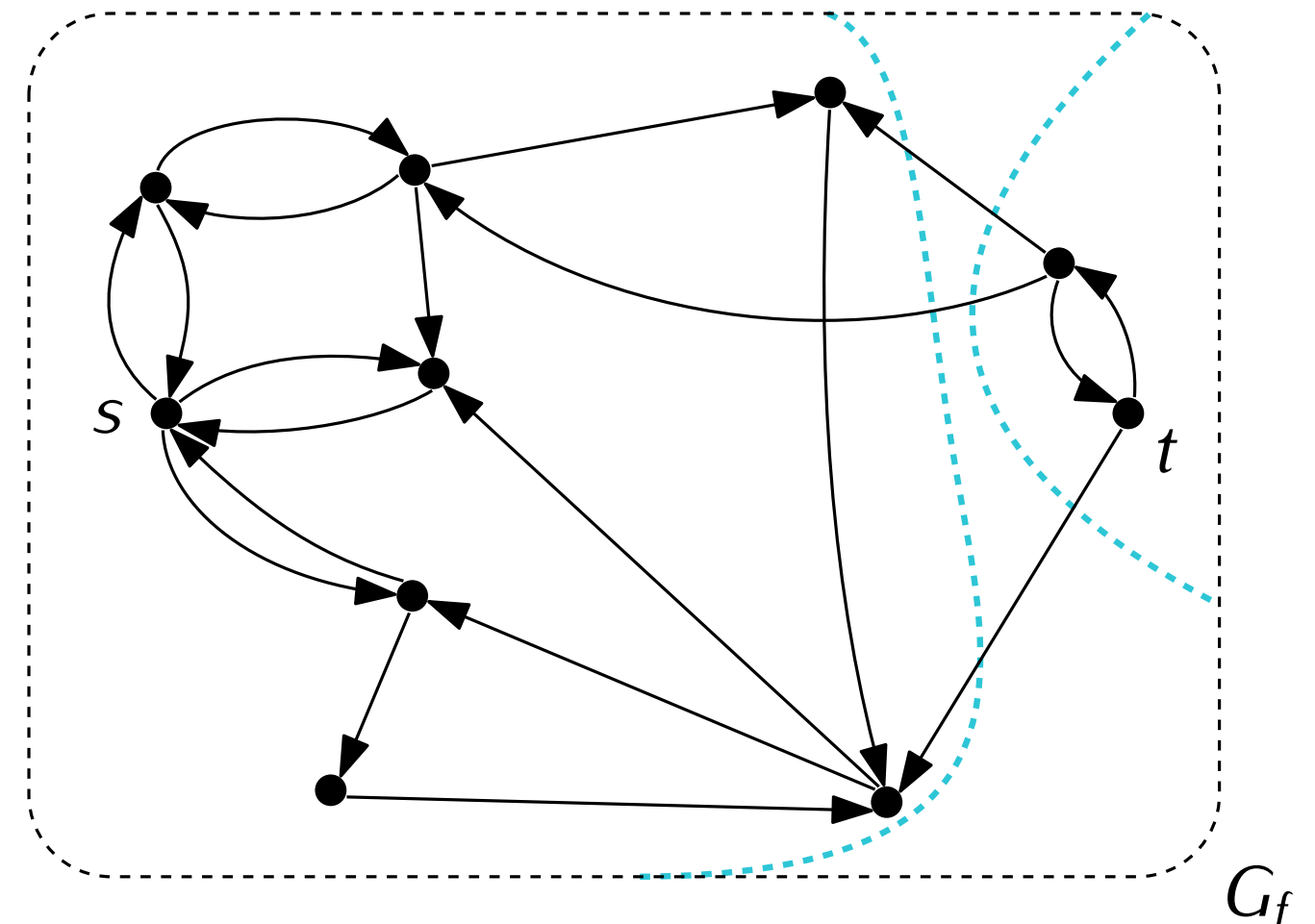
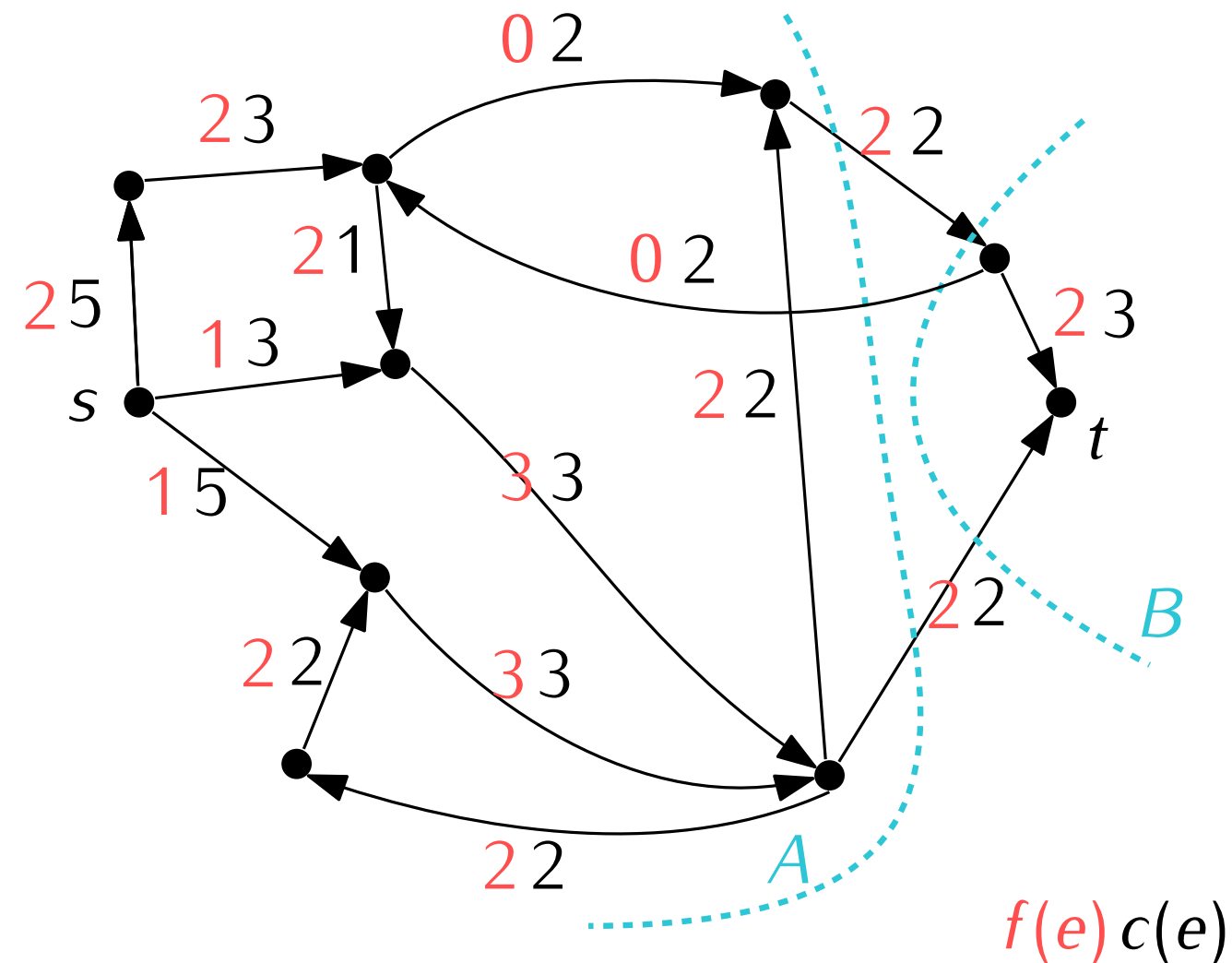
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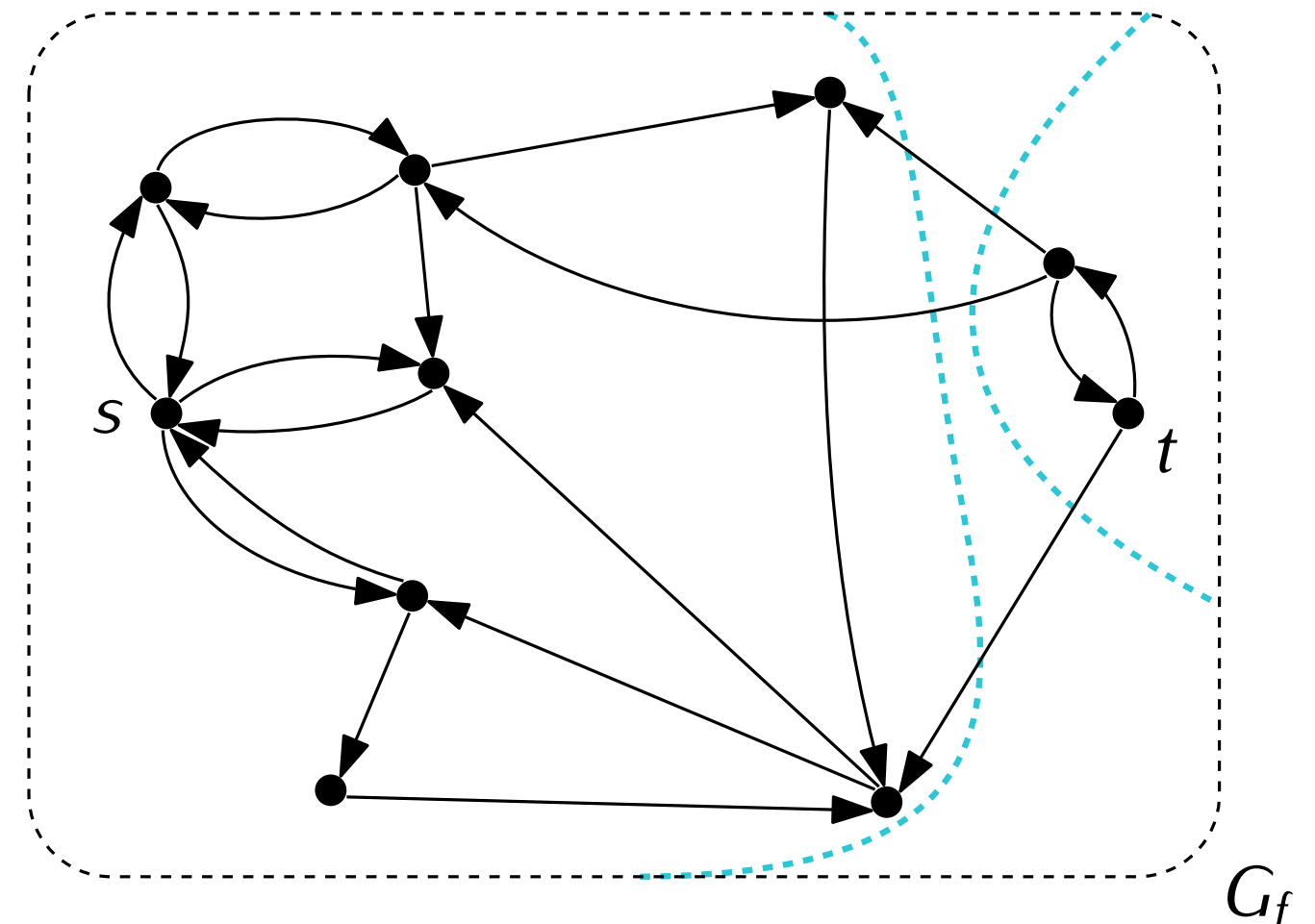
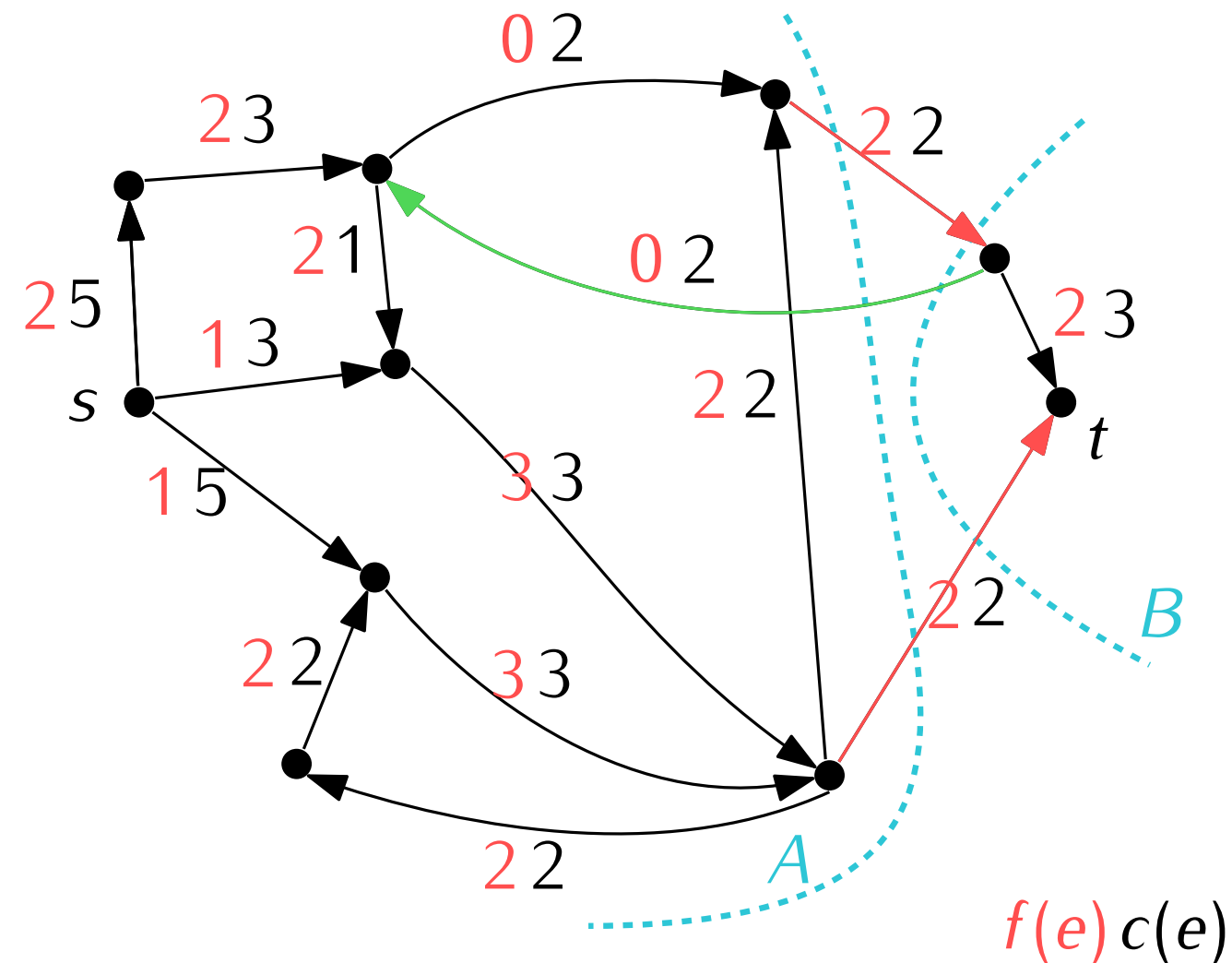
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We get: $\text{value of } f = \sum_{e \text{ out of } A} f(e) - \sum_{e \text{ into } A} f(e) = \sum_{e \text{ out of } A} c(e) \geq \text{maxflow}(G)$

Our algorithm in fact proves an important result:

Theorem. Every flow network has a maximal flow f such that $f(e) \in \mathbb{N}$ for all $e \in E$.

Interesting applications of the algorithm for max-flow and the integrality theorem:

- Efficient algorithm for maximum bipartite matching
- Hall's perfect matching theorem
- Efficient algorithm for finding edge-disjoint paths and Menger's theorem

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Other algorithms are possible:

$$\begin{array}{ll} \text{maximize} & \sum_{e \text{ out of } s} c(e) \\ \text{subject to} & \left\{ \begin{array}{ll} \sum_{e \text{ out of } v} f(e) - \sum_{e \text{ into } v} f(e) & = 0 \quad \text{for all } v \in V \\ c(e) - f(e) & \geq 0 \quad \text{for all } e \in E \\ f(e) & \geq 0 \quad \text{for all } e \in E \end{array} \right. \end{array}$$

Such **linear programs** can be solved in polynomial time: area of **linear and convex optimization**