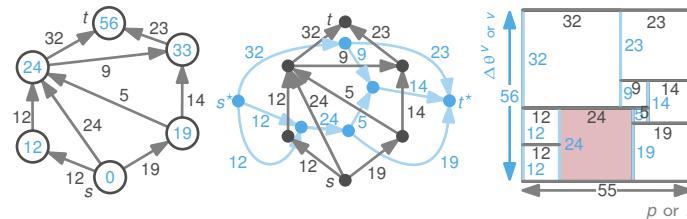
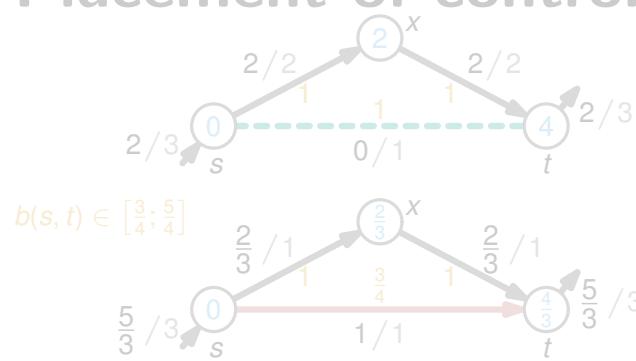


# Problem Statements

## Feasible electrical flows



# Placement of control units



## Cable layout



## Static models

## Alternating Current (AC) model

## Direct Current (DC) model

## Discrete control decisions

## Continuous control decision

## Wind farm cabling with multiple cable types

[Lehmann, Rutter, Wagner, and W.,  
2017]

# Formulations

## Vertex-based Formulation

$$f_{\text{net}}(u) = \sum_{\{u,v\} \in E} f(u, v) = 0$$

$$-\infty \leq f_{\text{net}}(u) \leq -d$$

$$0 \leq f_{\text{net}}(u) \leq \infty$$

$$\theta^v(v) - \theta^v(u) = f(u, v)$$

$$|f(u, v)| \leq \text{cap}(u, v)$$

**KCL**

**KVL**

## Vertex-/Cycle-based (Kirchhoff's) Formulation

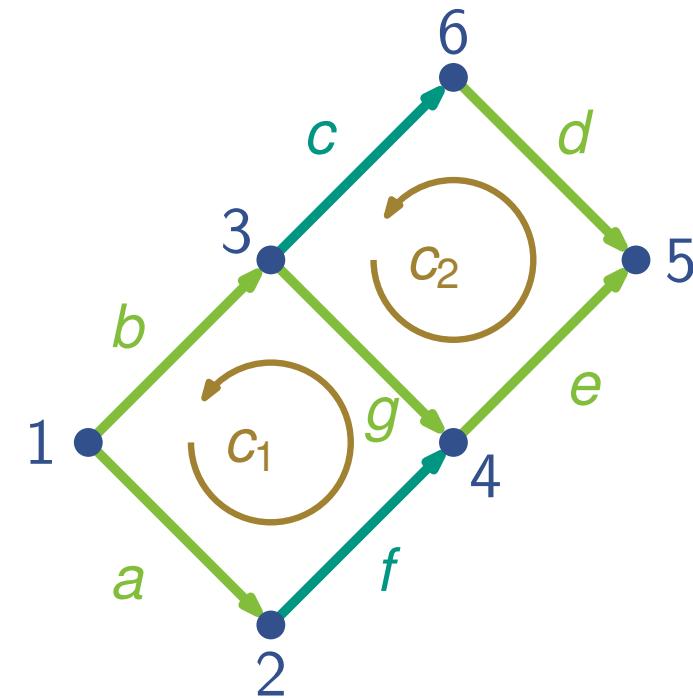
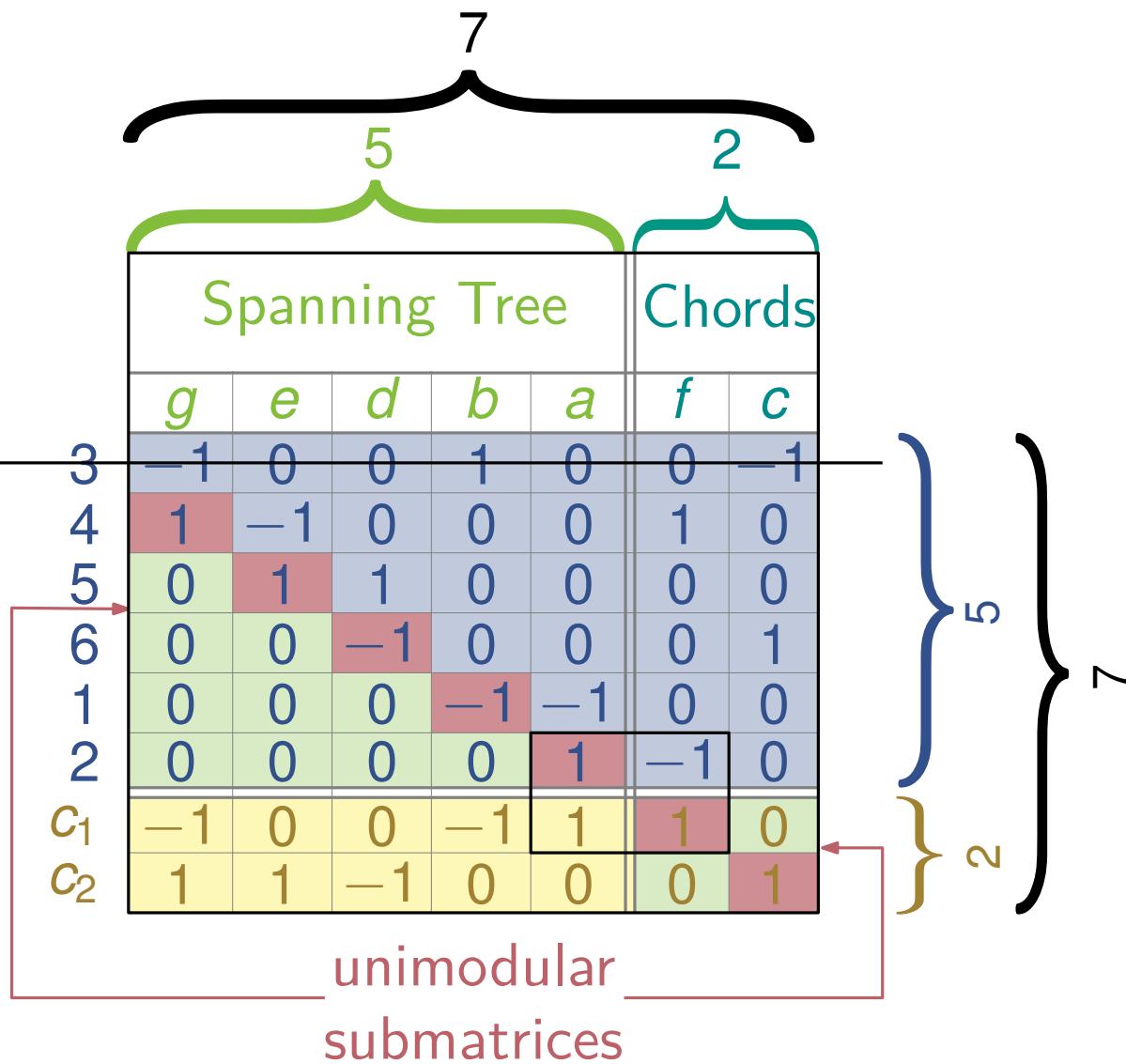
$$\mathbf{I} \vec{f} = \vec{0}$$

$$\mathbf{B} \overrightarrow{\Delta \theta^v} = \vec{0}$$

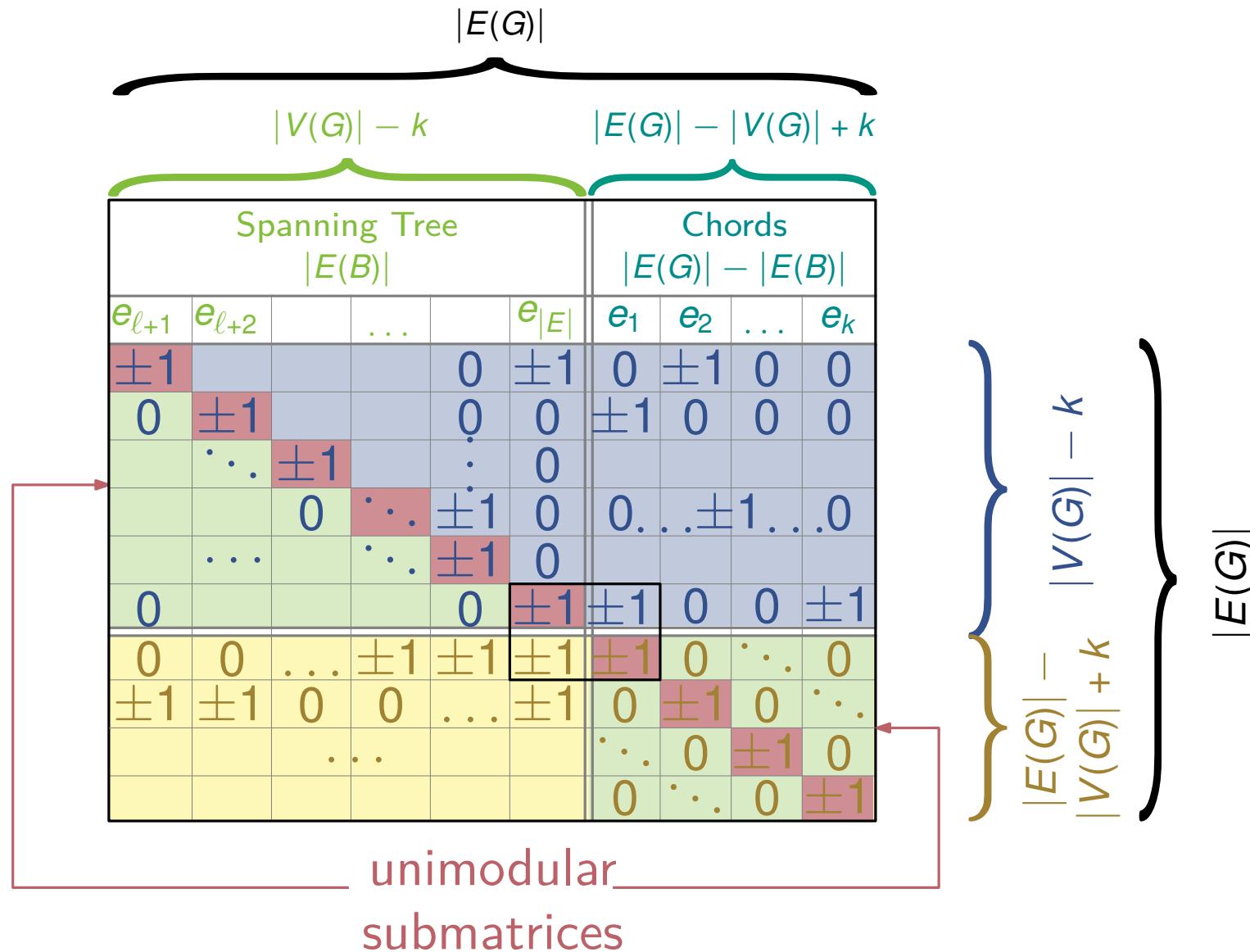
$$\vec{f} \leq \overrightarrow{\text{cap}}$$

**I** – incidence matrix  
**B** – circuite matrix

# Structure of the Incidence and Circuit Matrix



# General Structure



# General Structure



## Lemma 1

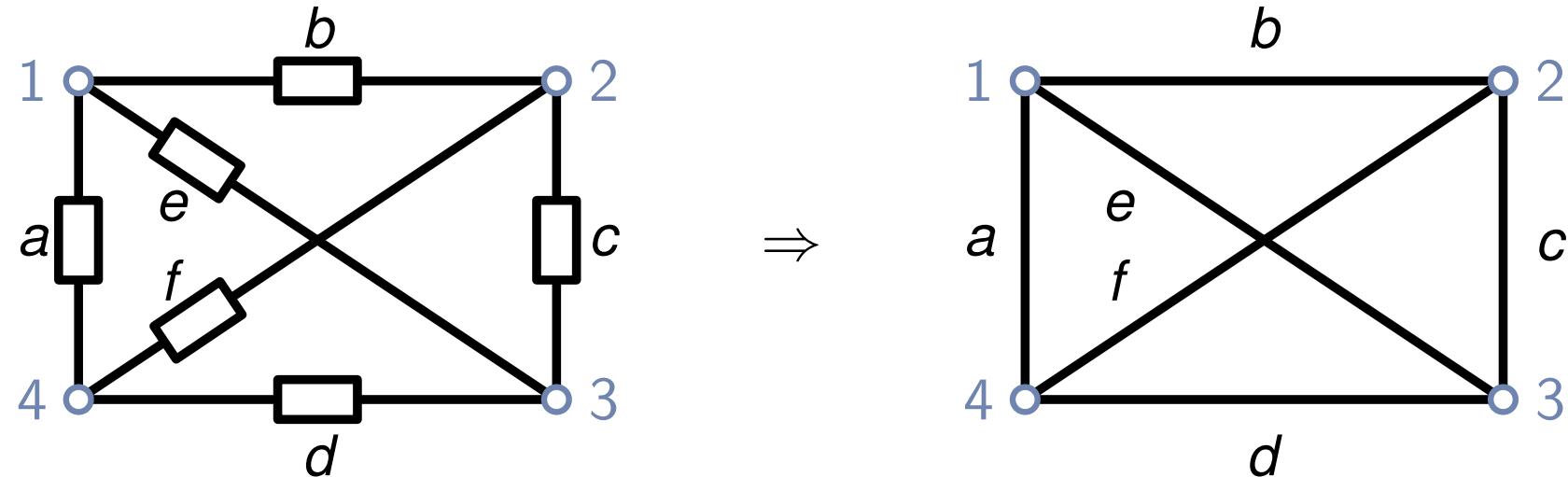
The bases of the incidence matrix **I** and the circuit matrix **B** are TUM. However, the whole system of equations to get an electrically feasible flow using the KCL and KVL is **not** TUM.

|         |         |     |         |         |         |         |         |         |         |
|---------|---------|-----|---------|---------|---------|---------|---------|---------|---------|
| 0       | 0       | ... | $\pm 1$ | $\pm 1$ | $\pm 1$ | $\pm 1$ | 0       | ...     | 0       |
| $\pm 1$ | $\pm 1$ | 0   | 0       | ...     | $\pm 1$ | 0       | $\pm 1$ | 0       | ...     |
|         |         | ... |         |         |         | ...     | 0       | $\pm 1$ | 0       |
|         |         |     |         |         |         | 0       | ...     | 0       | $\pm 1$ |
|         |         |     |         |         |         |         |         |         |         |

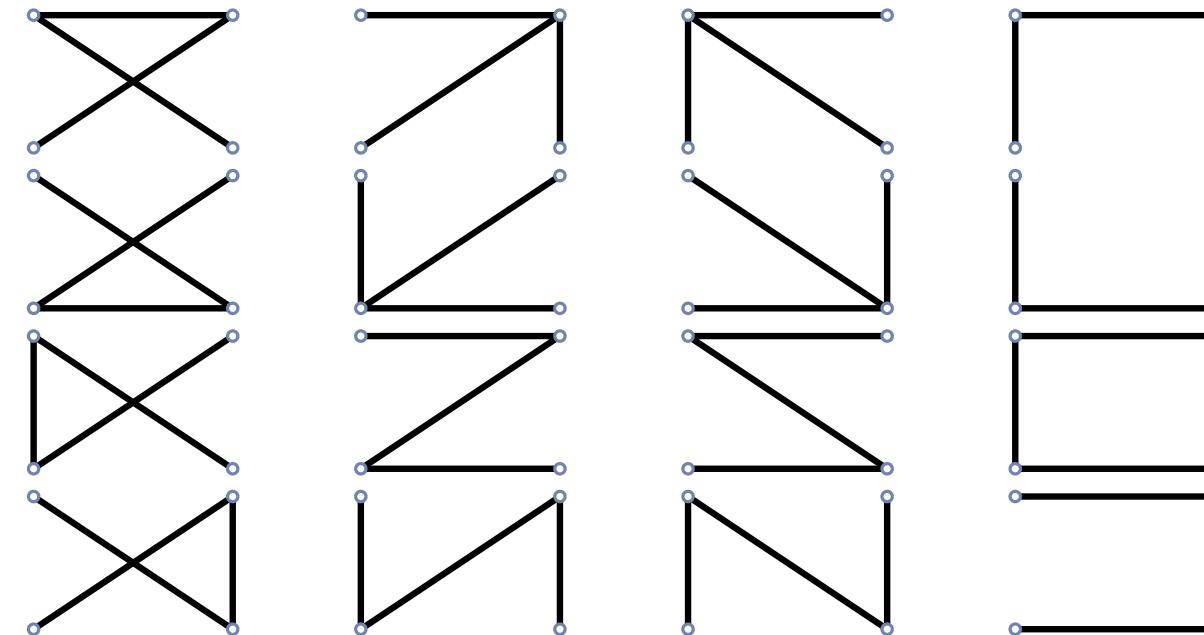
unimodular  
submatrices

$$\left. \begin{array}{l} |E(G)| - \\ |V(G)| + k \end{array} \right\}$$

# Circuits and Spanning Trees



All spanning trees  $T$



# An Ancient Algorithm for the Power Flow

- A first algorithm that represents a structural result

## Lemma 2 [p.36, Lemma 1; Shapiro, 1987]

Let every edge of  $G$  have a resistor of 1 ohm. Let  $N$  denote the number of spanning trees and let  $N(s, a \rightarrow b, t)$  be the spanning trees that contain the edge  $(a, b)$  in that particular direction. Put a 1-ampere current between  $s$  and  $t$  and let  $i(a, b) = (N(s, a \rightarrow b, t) - N(s, b \rightarrow a, t))/N$ . Then  $i(a, b)$  is the current in the edge  $ab$  oriented from  $a$  to  $b$ .

- Apply Binet-Cauchy theorem on matrix  $\mathbf{Y}_n = \mathbf{I} \mathbf{Y}_e \mathbf{I}^T$
- $\Delta_n = \det(\mathbf{Y}_n) = \det(\mathbf{I} \mathbf{Y}_n \mathbf{I}^T) = \sum_{T \in \mathcal{T}} (\text{Tree-Admittance Product of } T)$
- Tree-Admittance Product  $\sum_{(u,w) \in E(T)} \mathbf{Y}_{u,w}$

# Primal and Dual Graphs

## Theorem 3 [p.522, Theorem 23; Whitney, 1935]

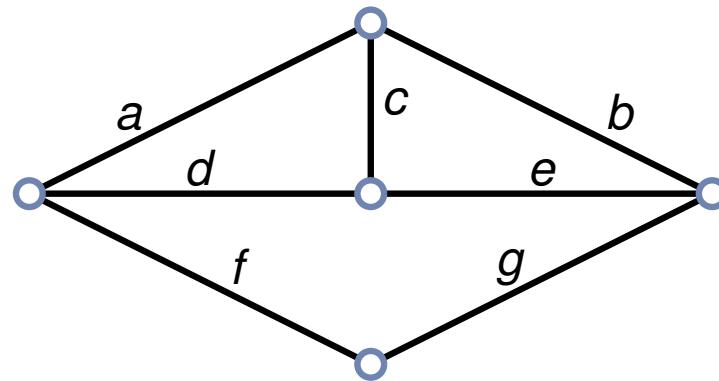
Let  $\mathcal{E}$  be a planar embedding of a primal graph  $G$  with  $G(\mathcal{E})$  being isomorphic to  $G$ . The graphs  $G$  and  $G^*$  are duals if and only if there is a bijection  $\mu_{\text{dual}}: E(G) \rightarrow E(G^*)$  between their edges such that bases in one correspond to base complements in the other.

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primal graph  $G$

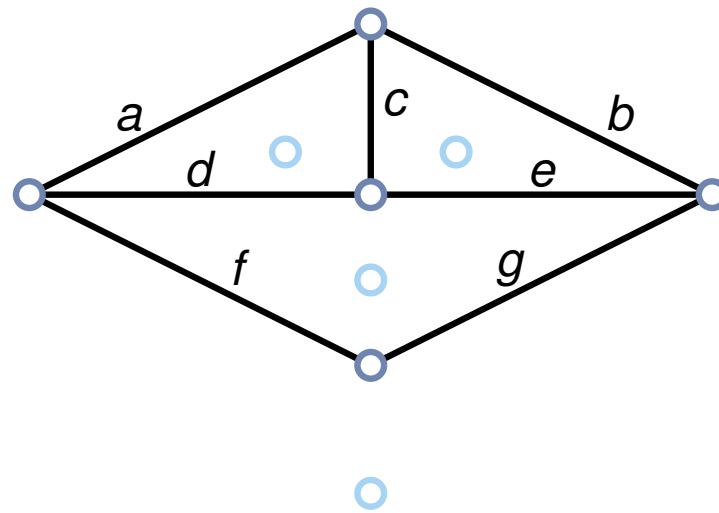


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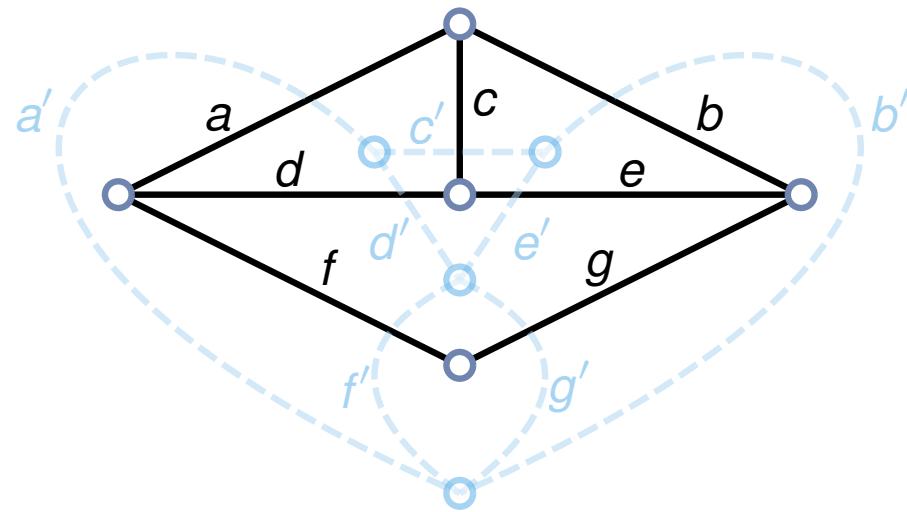


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dual graph  $G^*$

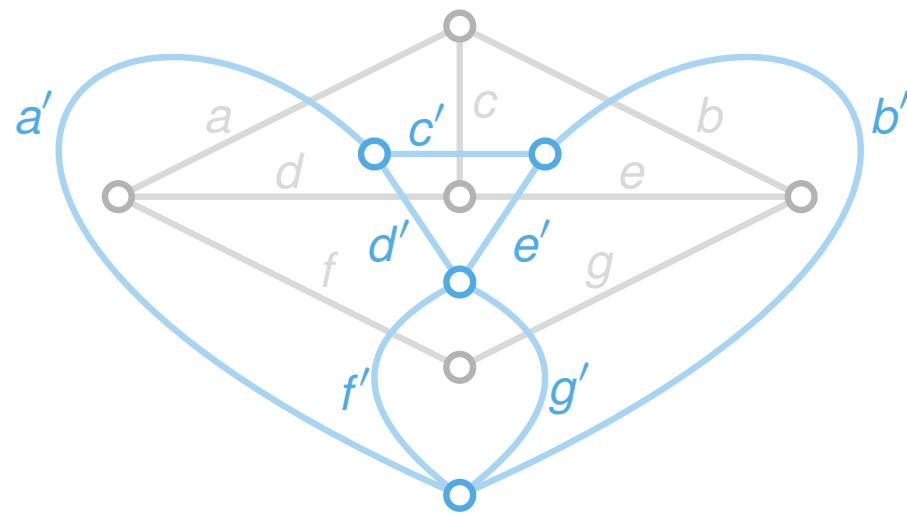


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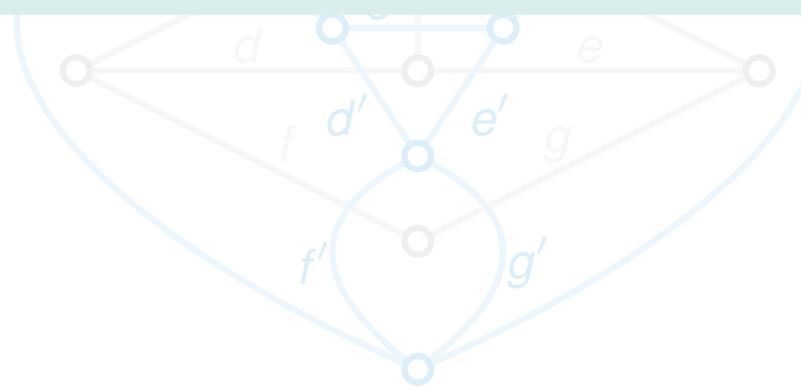
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**Corollary 4** [p.85, Corollary 4-24; Seshu and Reed, 1961]

If  $G$  and  $G^*$  are dual graphs, the incidence matrix of either graphs is a circuit matrix of the other (with the proper rank, and each row representing a cycle); that is

$$I_1 = B_2 \text{ and } I_2 = B_1.$$



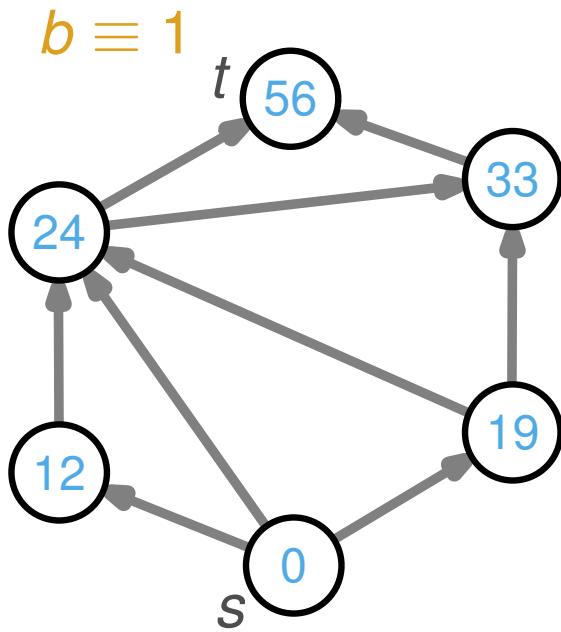
# Power Flow Problem Reformulation

## PLANAR $s$ - $t$ PF AND MPF

**Instance:** A plane  $s$ - $t$ -graph  $G$ , its dual graph  $G^*$  and a bijection  $\mu_{\text{dual}}: E(G) \rightarrow E(G^*)$ .

**Objective:** Find feasible flows  $f_G, f_{G^*}: E \rightarrow \mathbb{R}_{\geq 0}$  in  $G$  and  $G^*$  such that for every edge  $e \in E(G)$  we have

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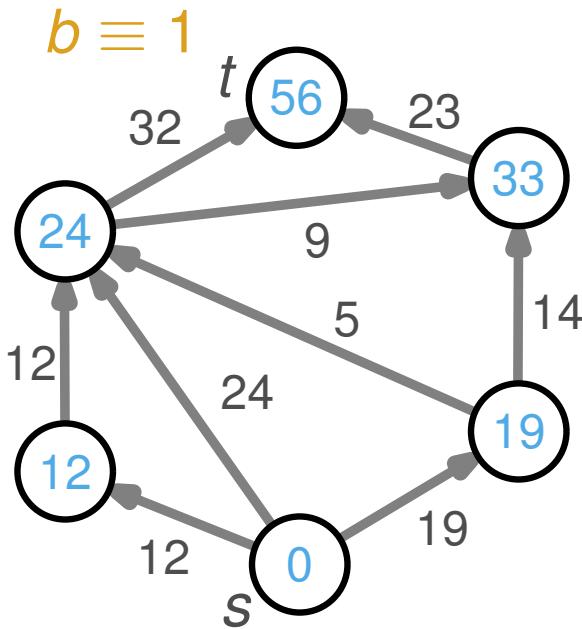
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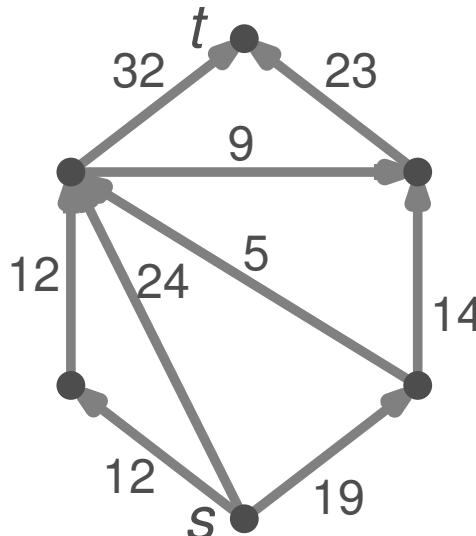
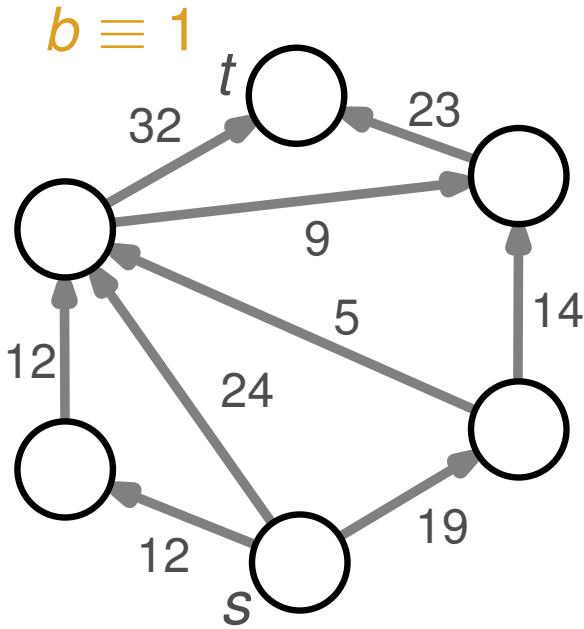
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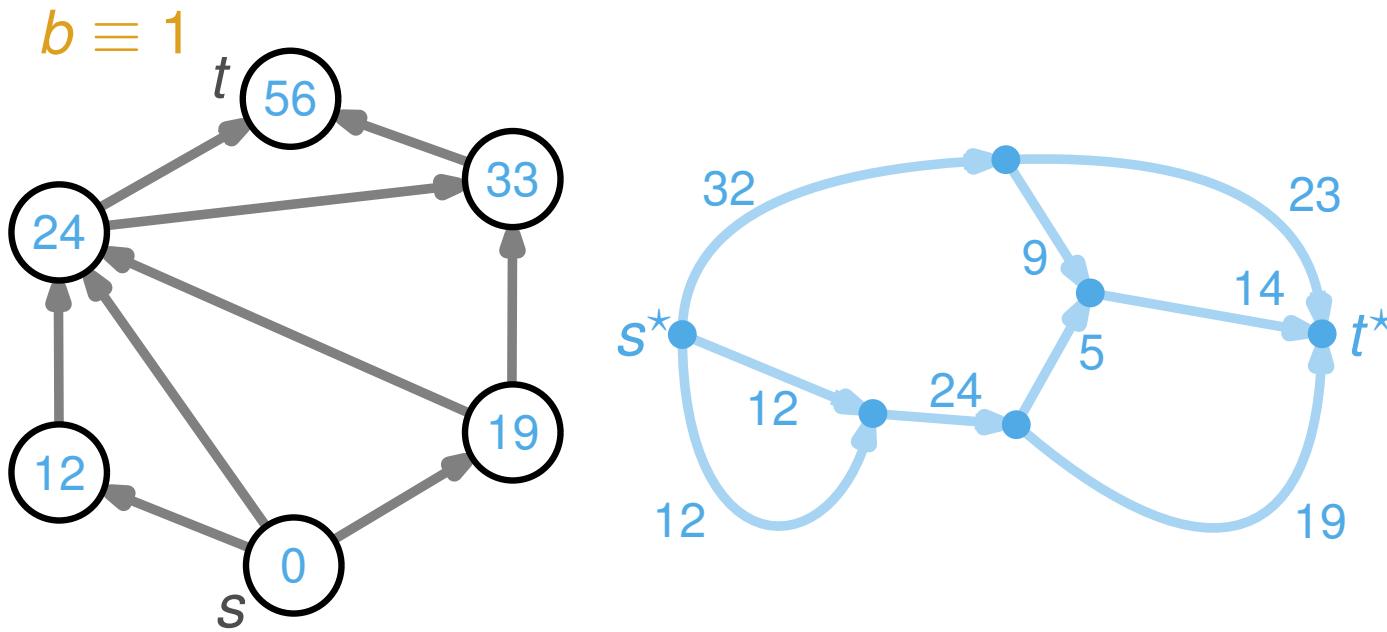
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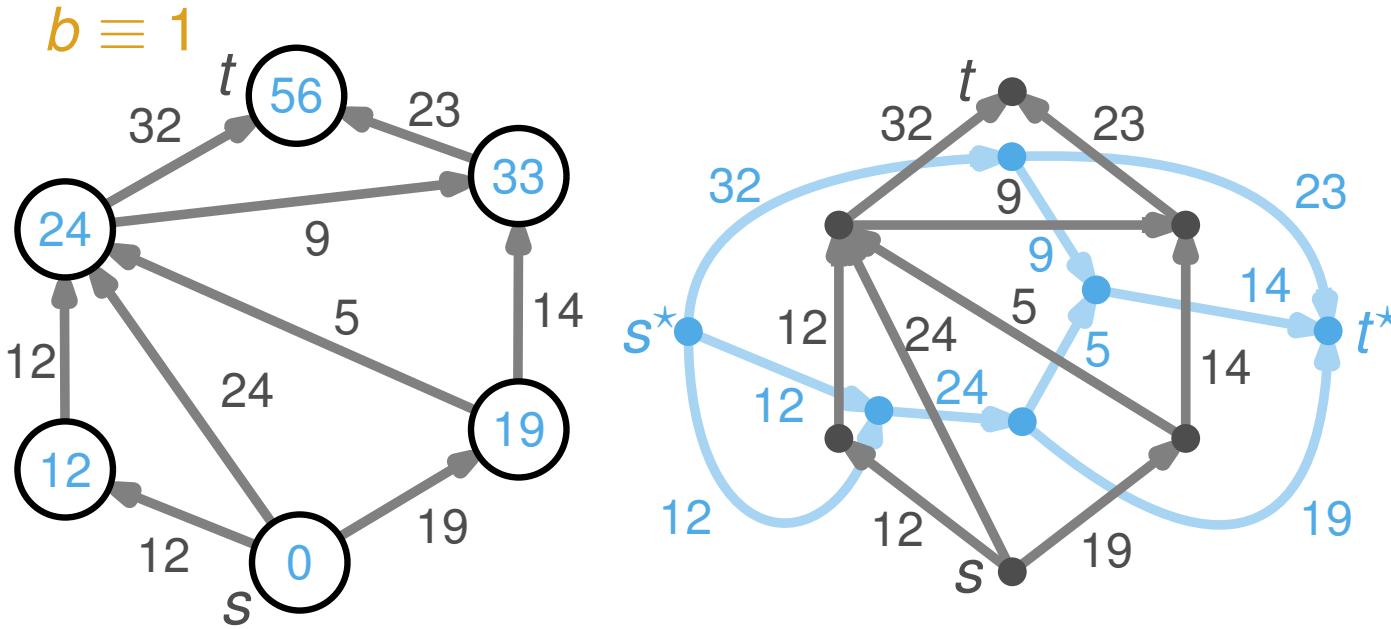
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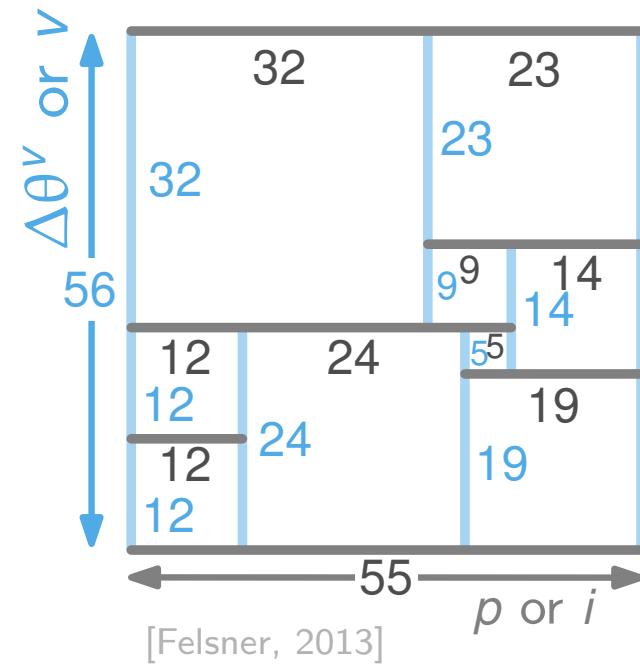
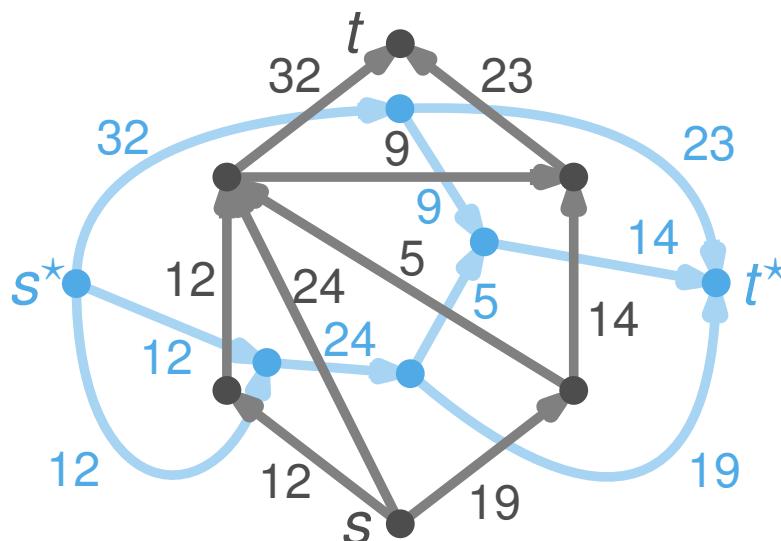
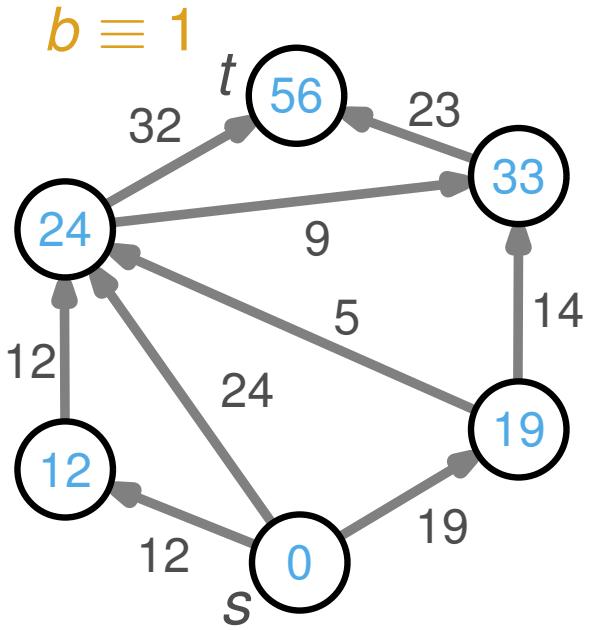
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[Felsner, 2013]

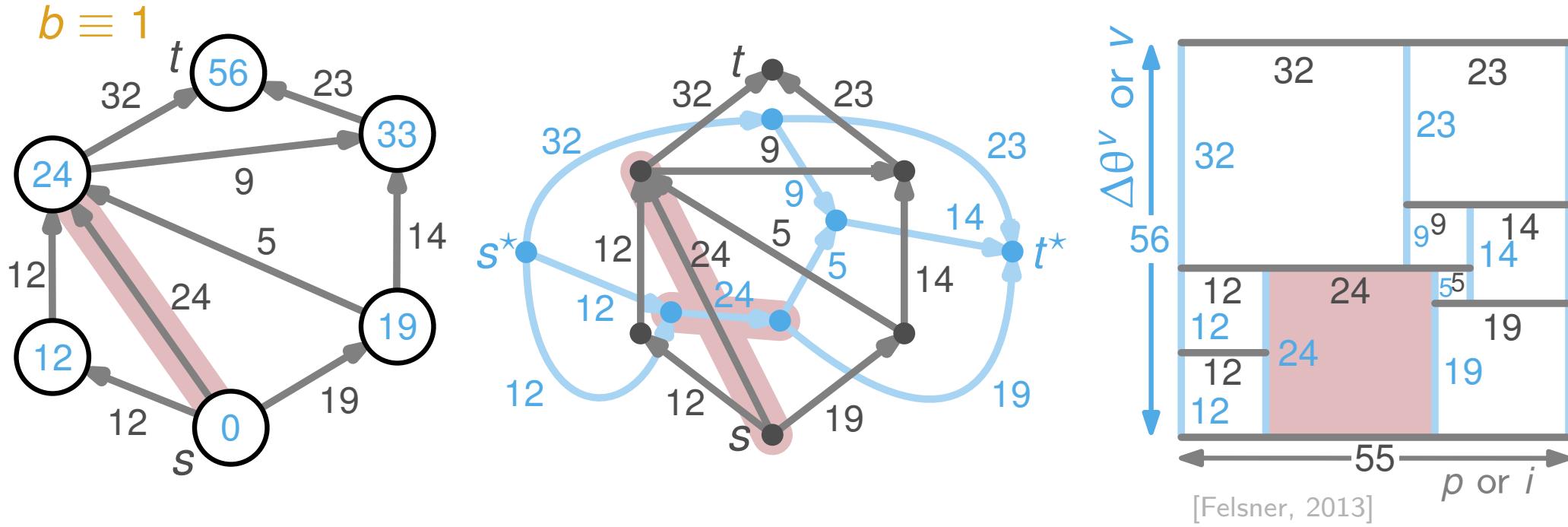
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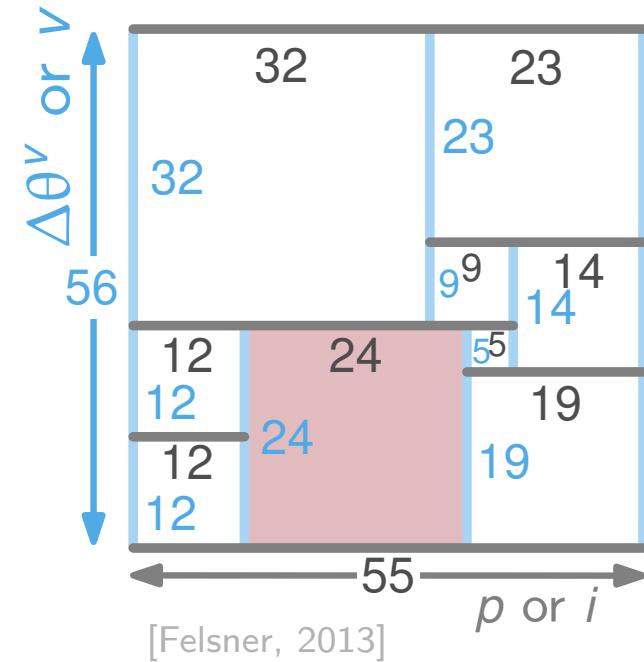
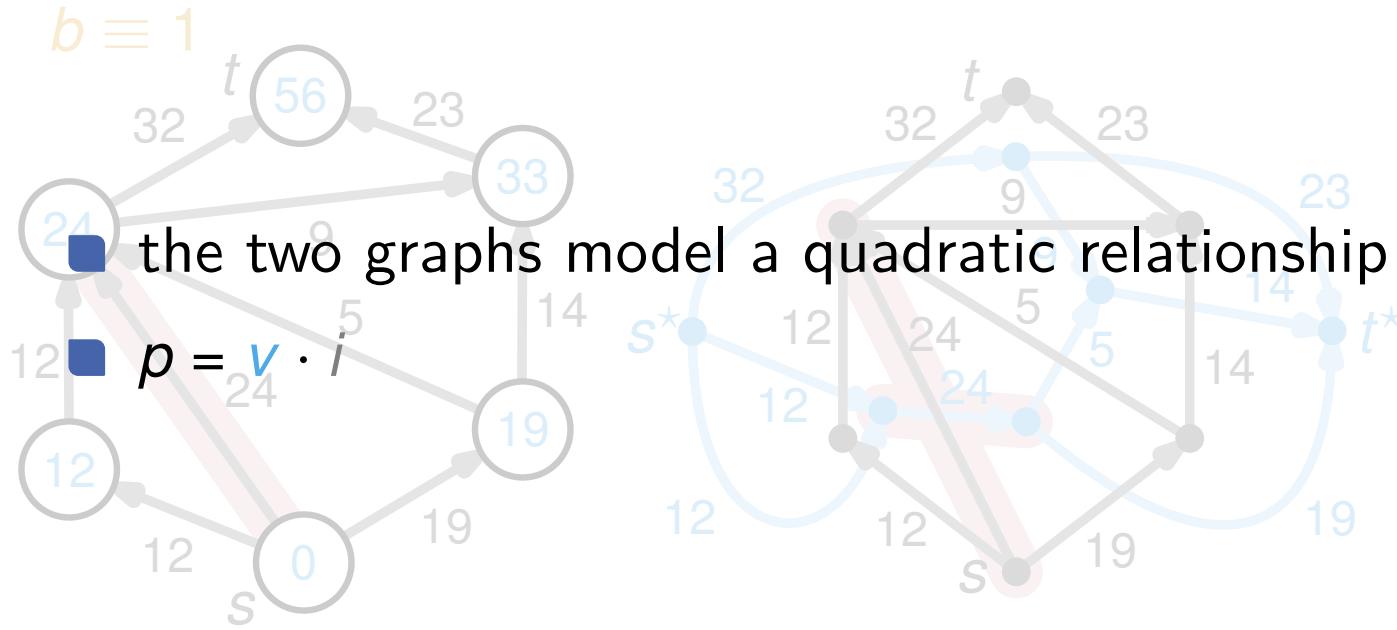
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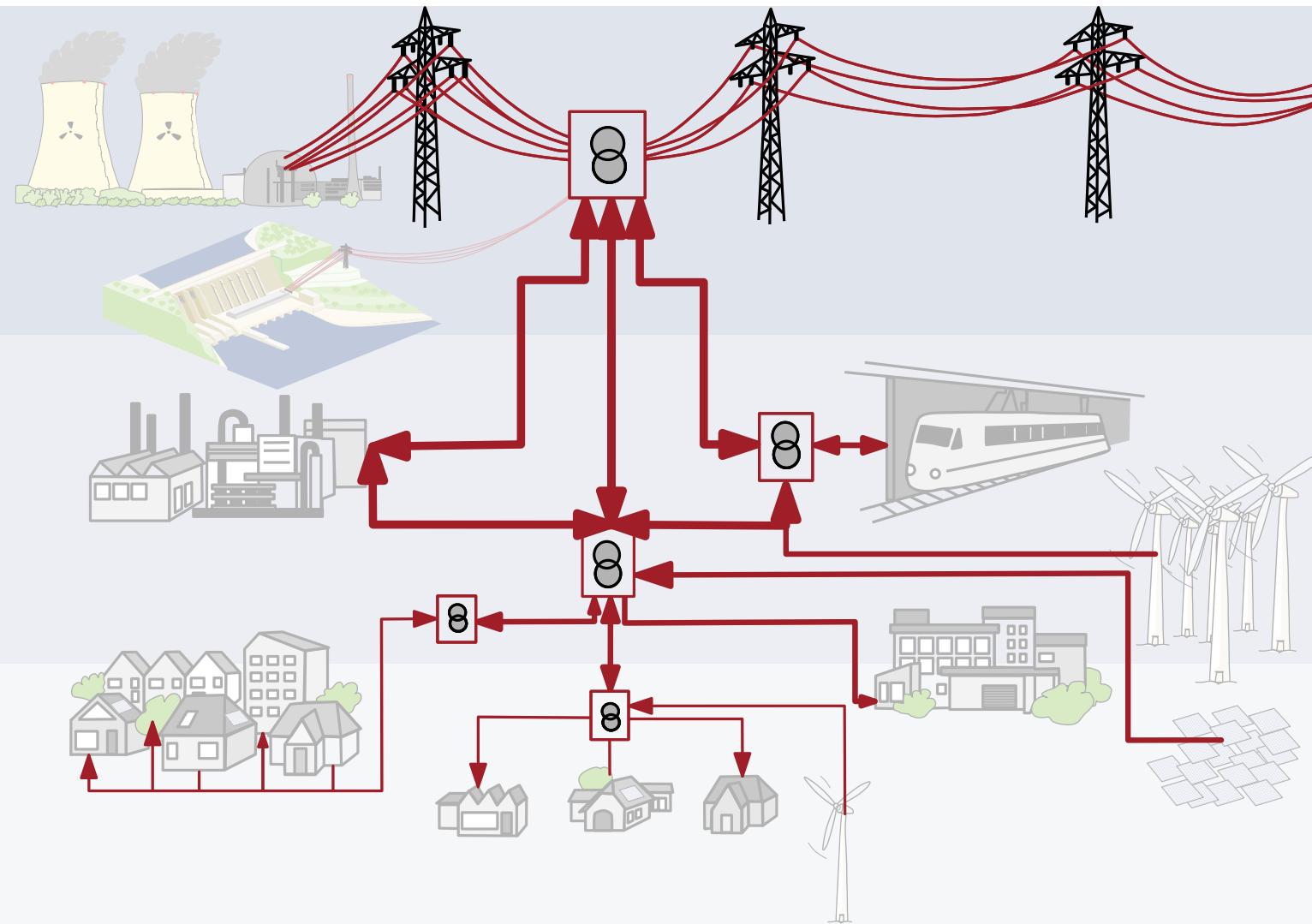
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# Towards an Algorithm for the Power Flow

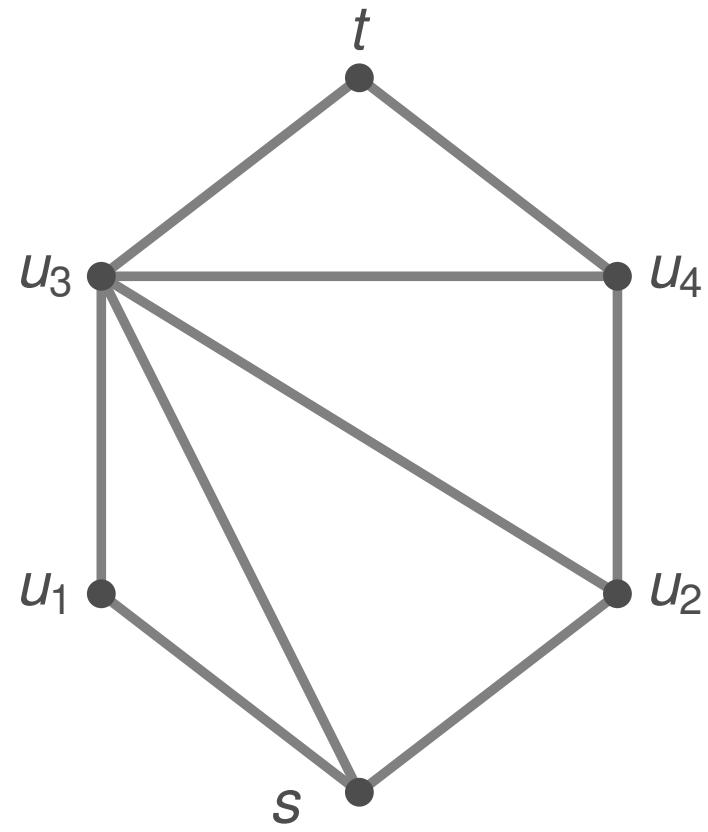
Producer



Power Grid

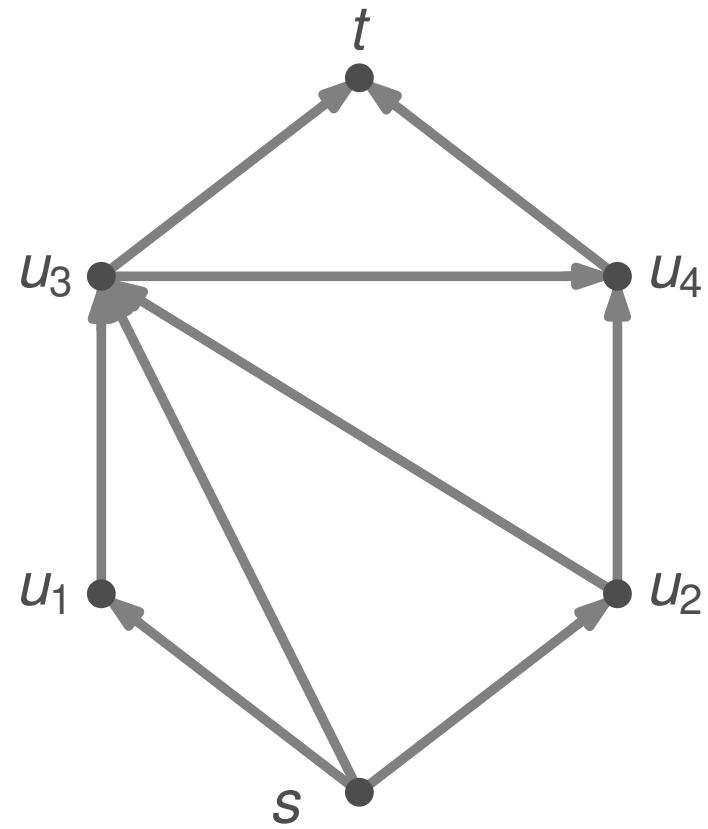
Prosumer

# Algorithmic Sketch



# Algorithmic Sketch

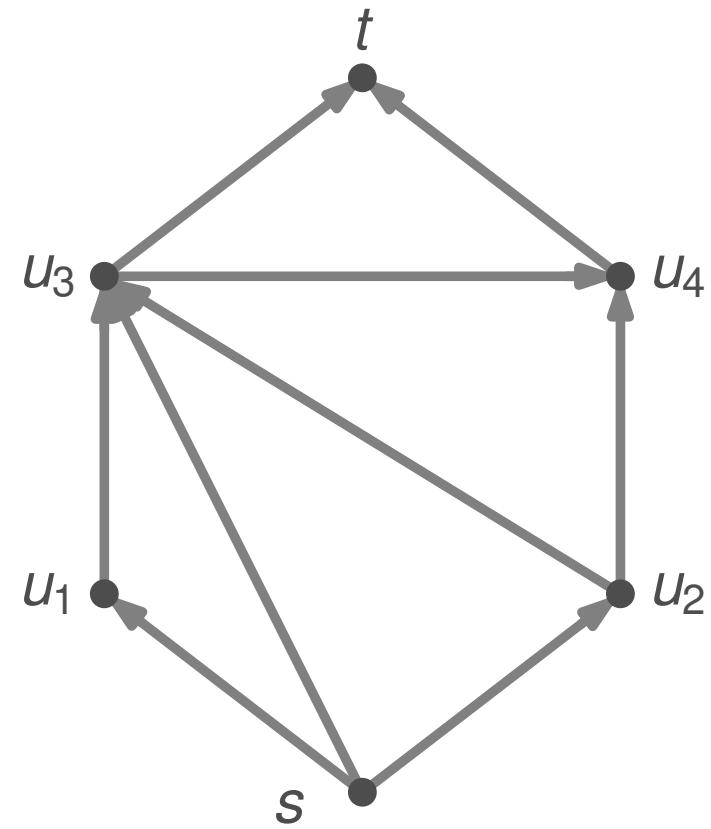
```
 $G = \text{bipolarSubgraphOf}(G, s, t)$ 
```



# Algorithmic Sketch

$G = \text{bipolarSubgraphOf}(G, s, t)$

$\mathcal{E} = \text{planarEmbeddingOf}(G)$



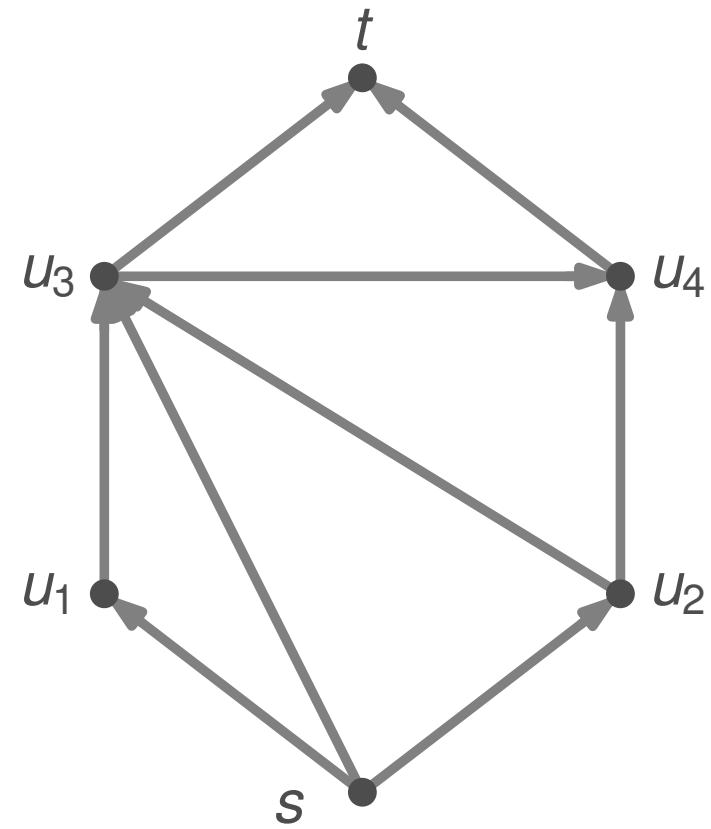
▷ PQ-Tree; Invariant  $G(\mathcal{E}) \cong G$

# Algorithmic Sketch

$G = \text{bipolarSubgraphOf}(G, s, t)$

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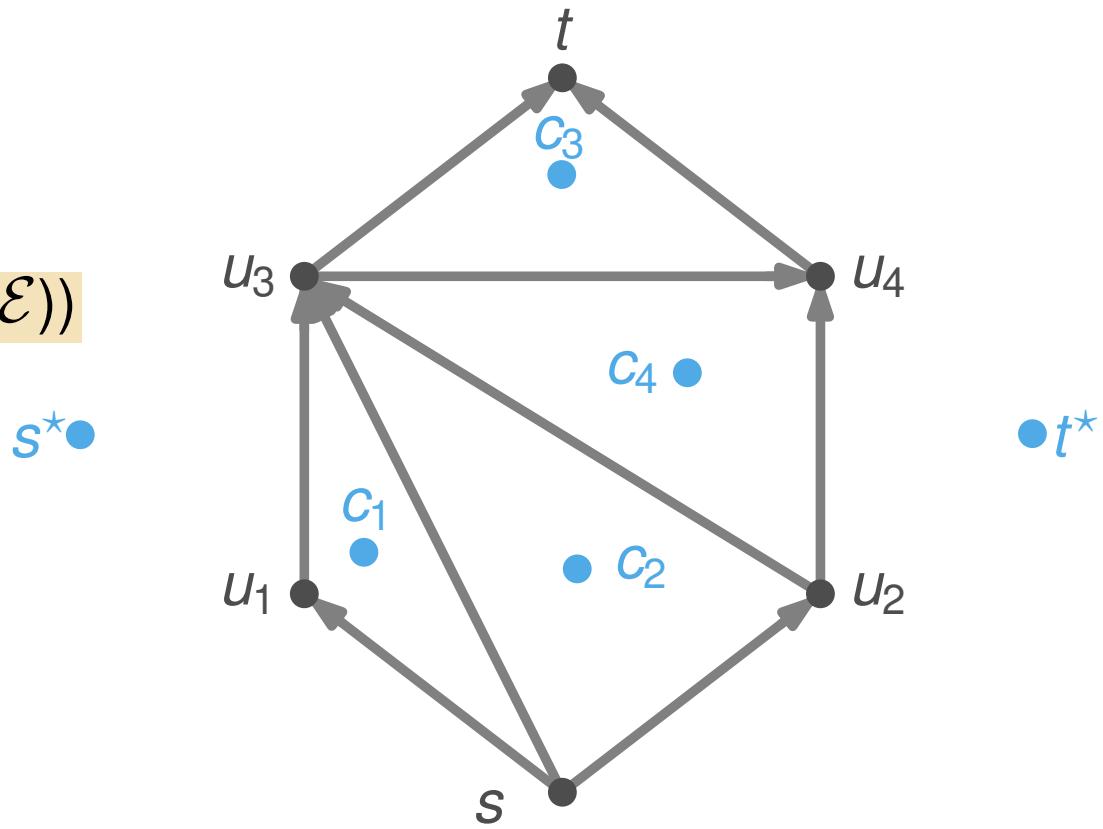


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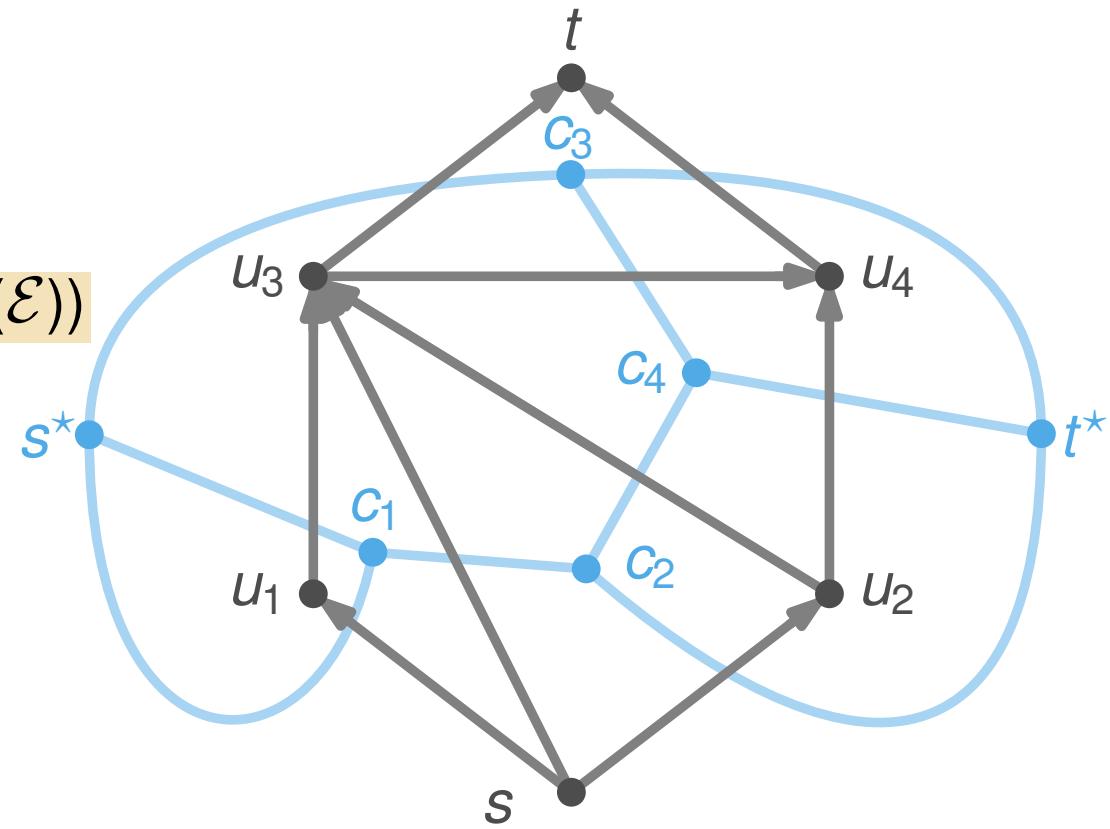


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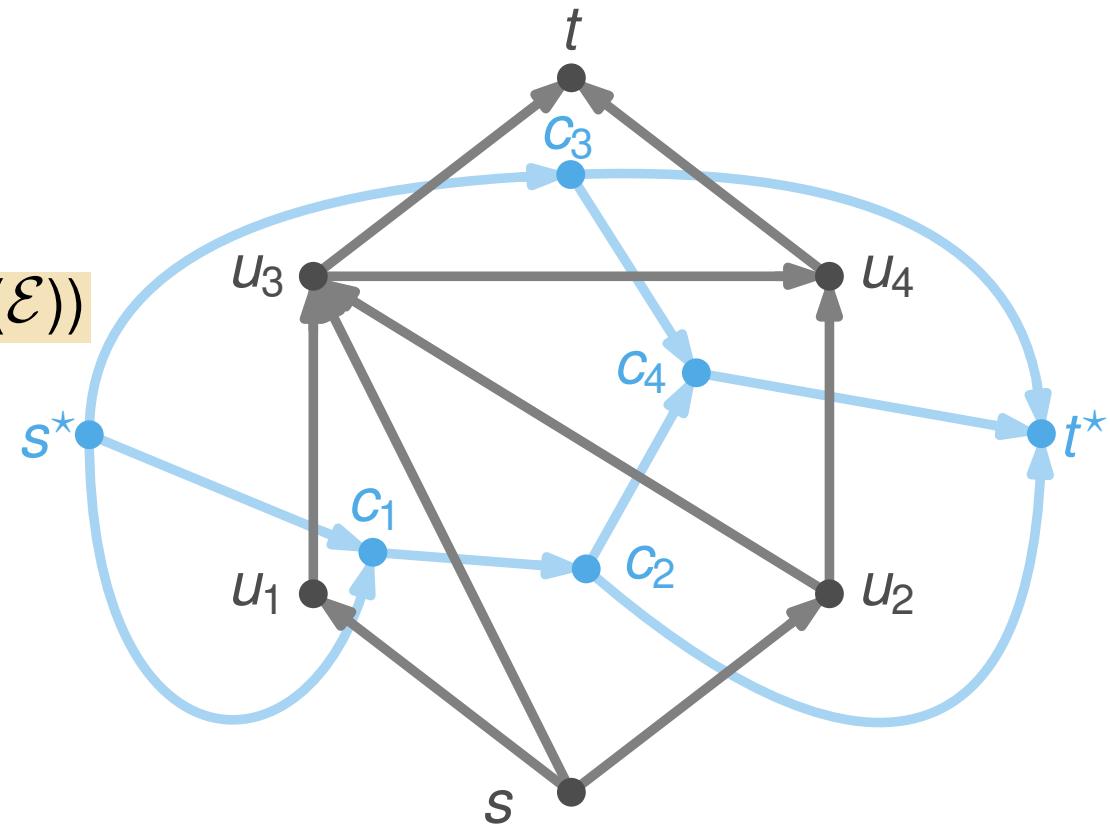


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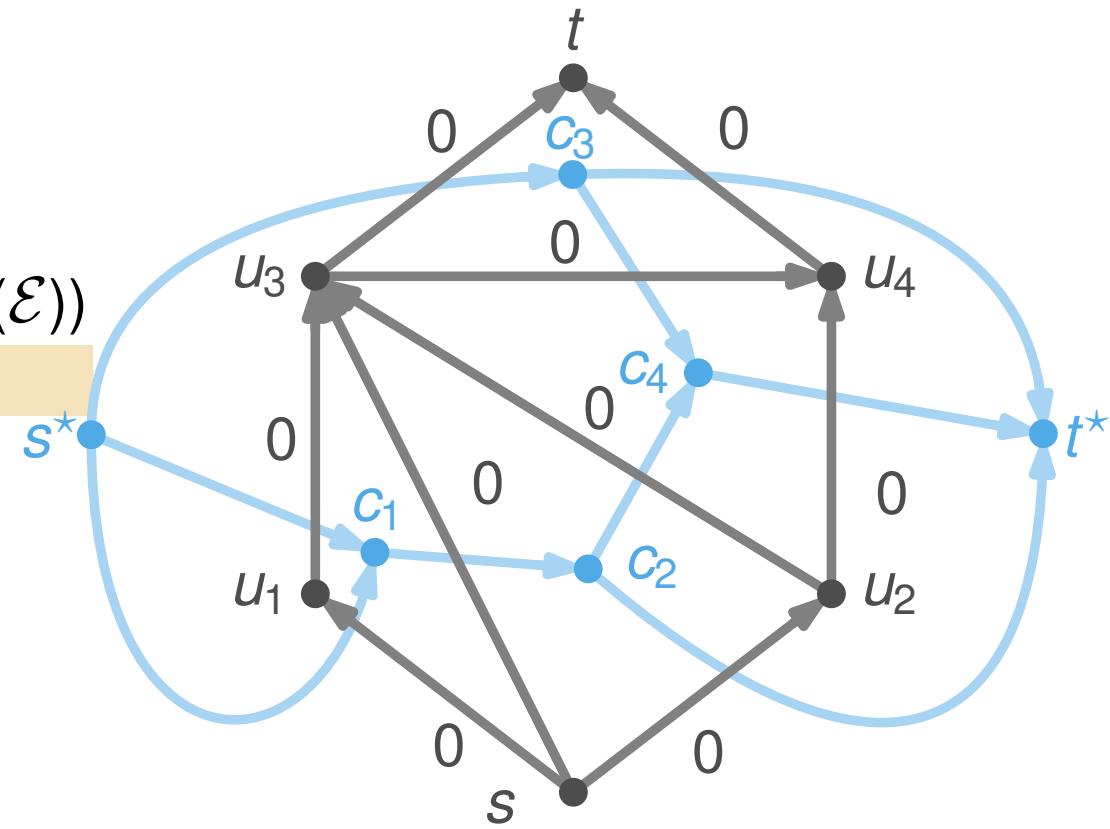
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▷ Augment flow along an incident edge at source  $s$

# Algorithmic Sketch

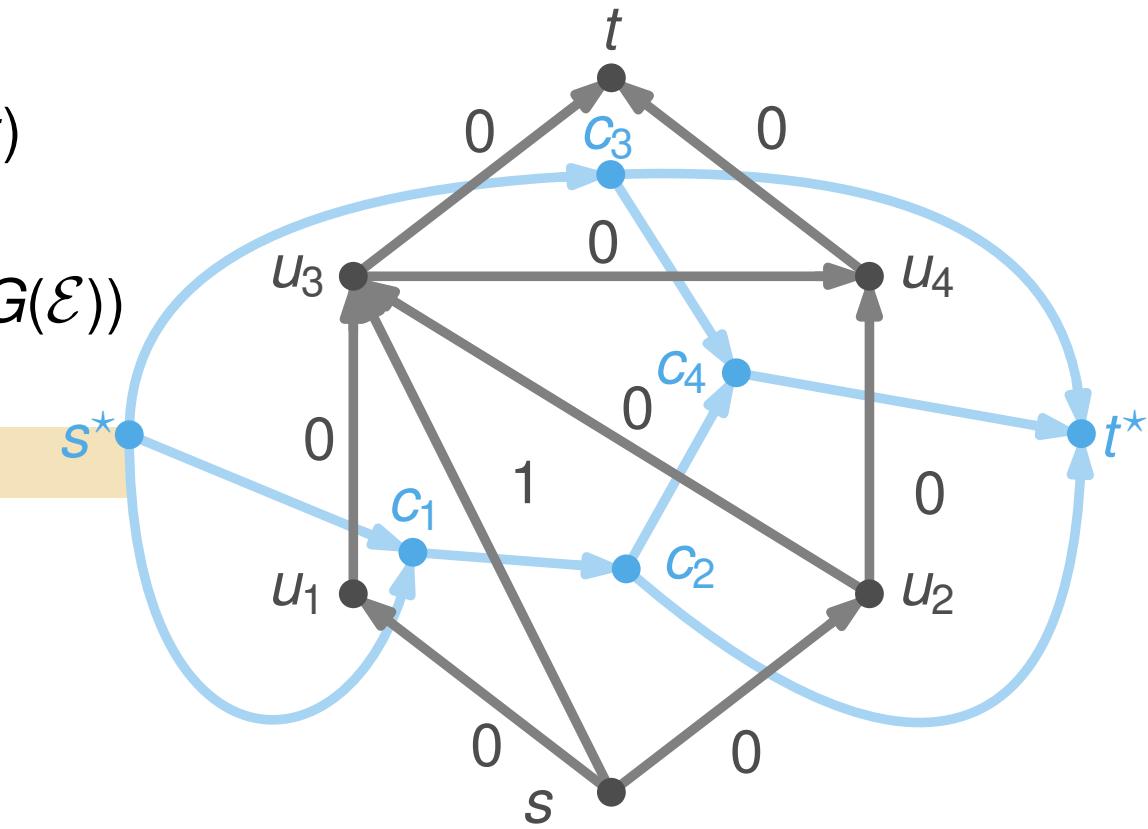
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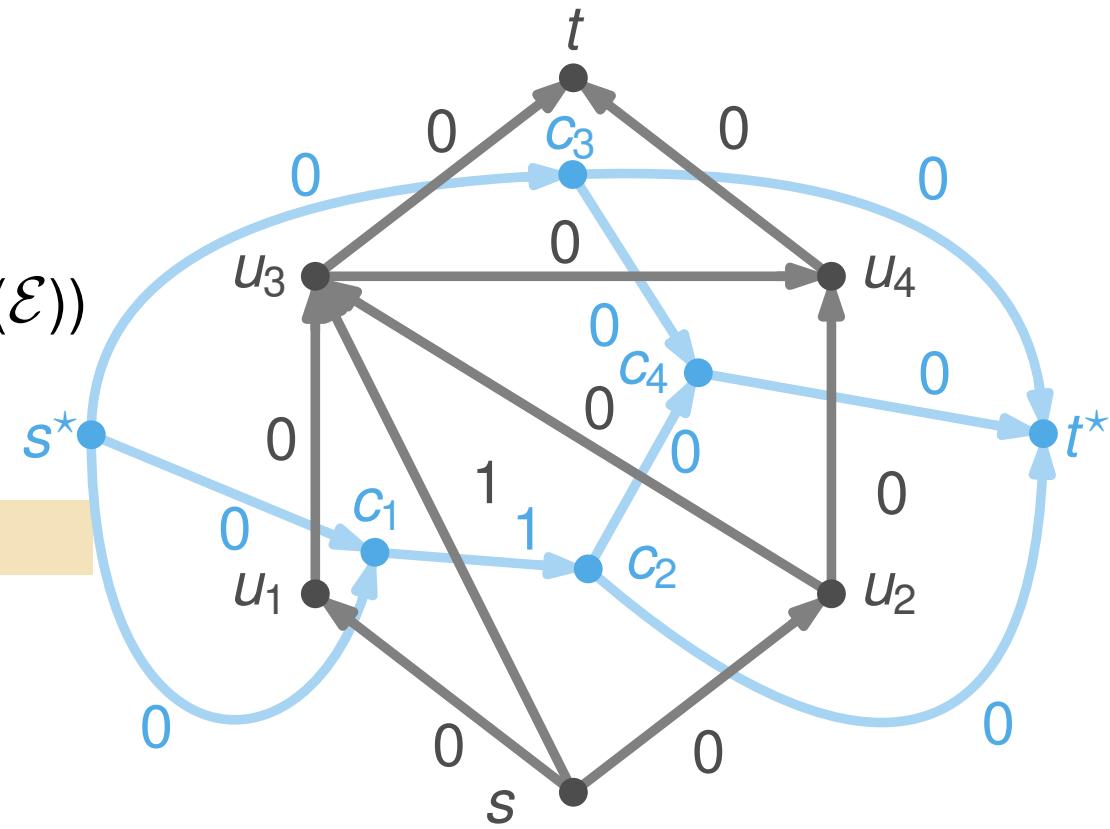
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▷ Bijective function

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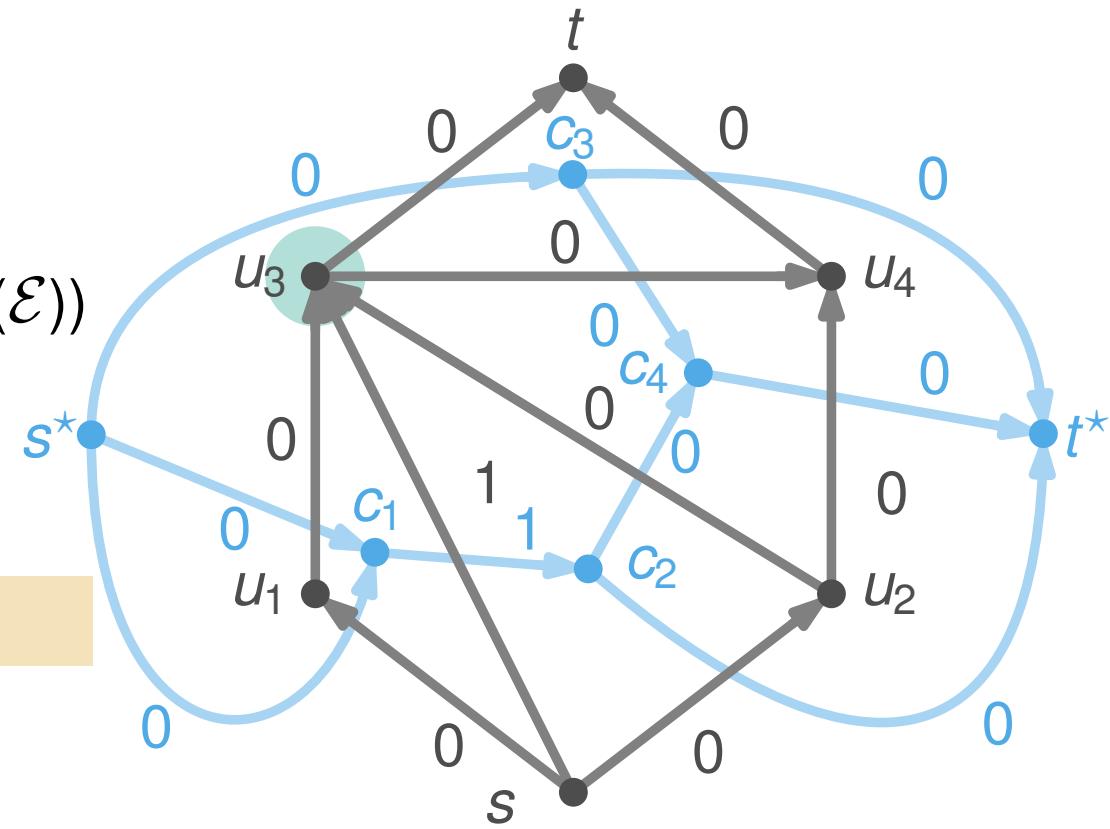
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$X = \bigcup_{u \in V(G): f_{\text{net}}(u) \neq 0} \{u\}$



▷ Check KCL property in graph  $G$

# Algorithmic Sketch

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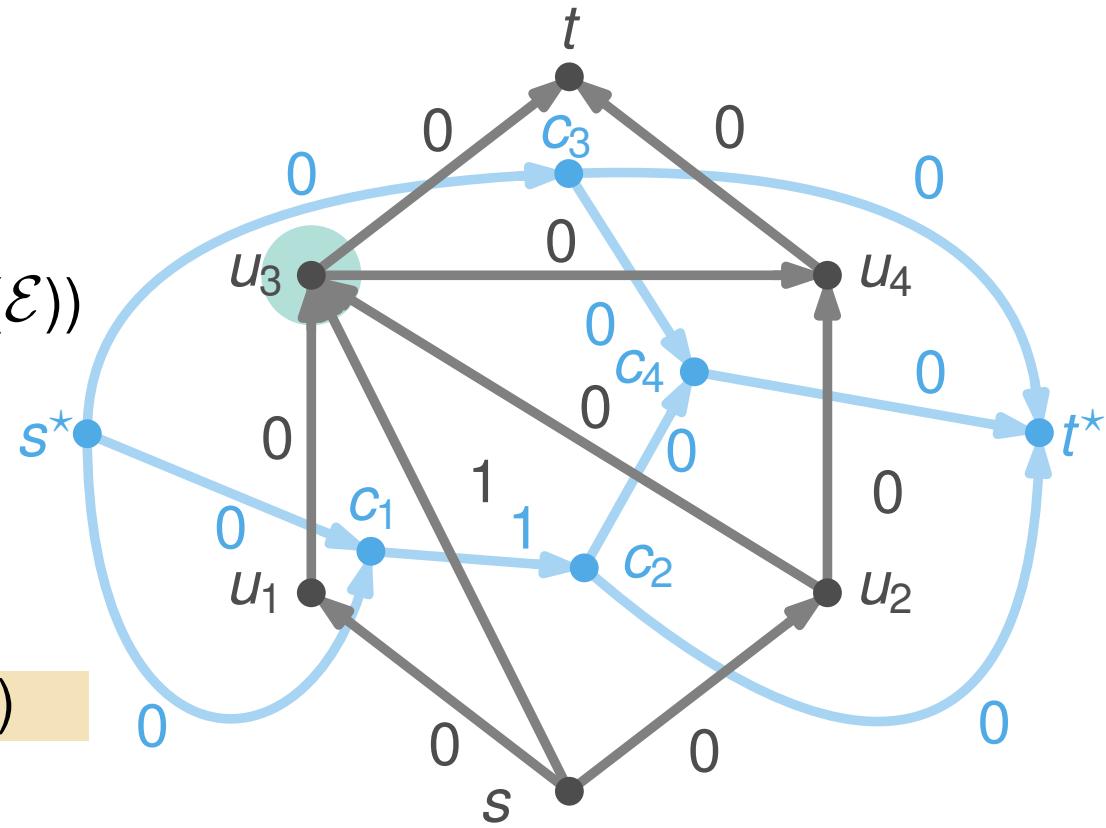
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$f = \text{resolveConflict}(G, s, t, f, X)$



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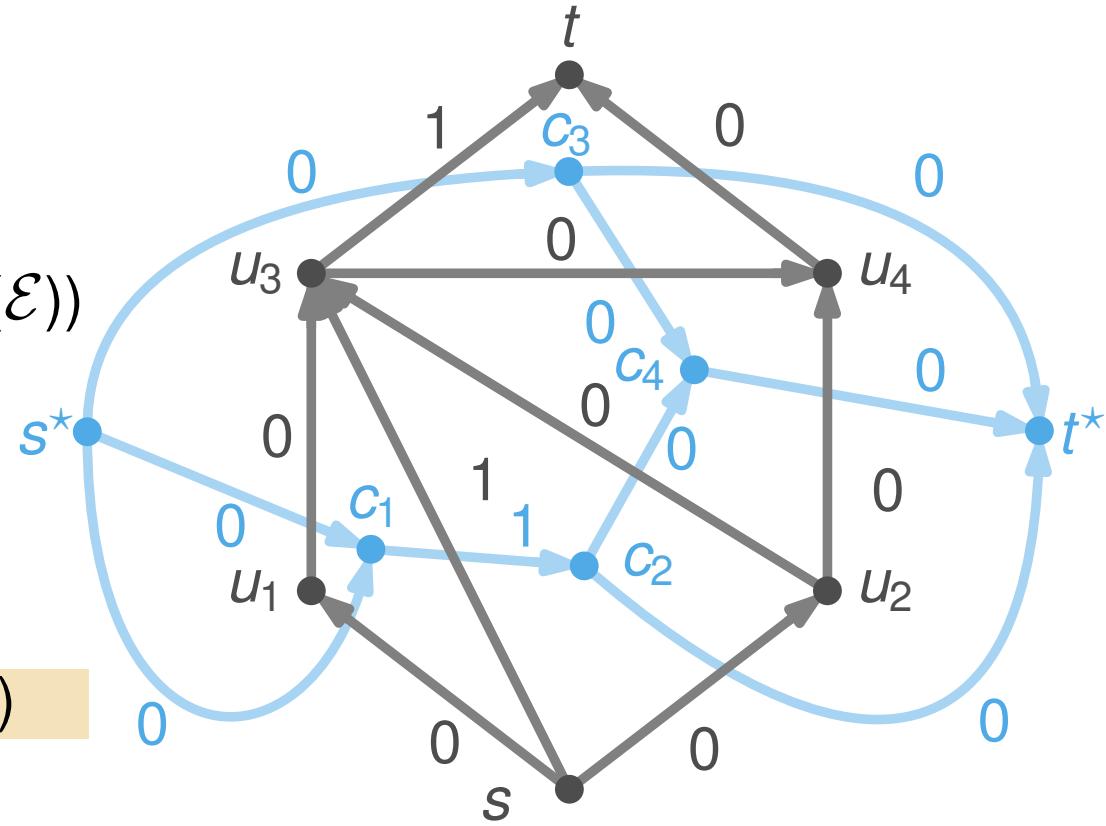
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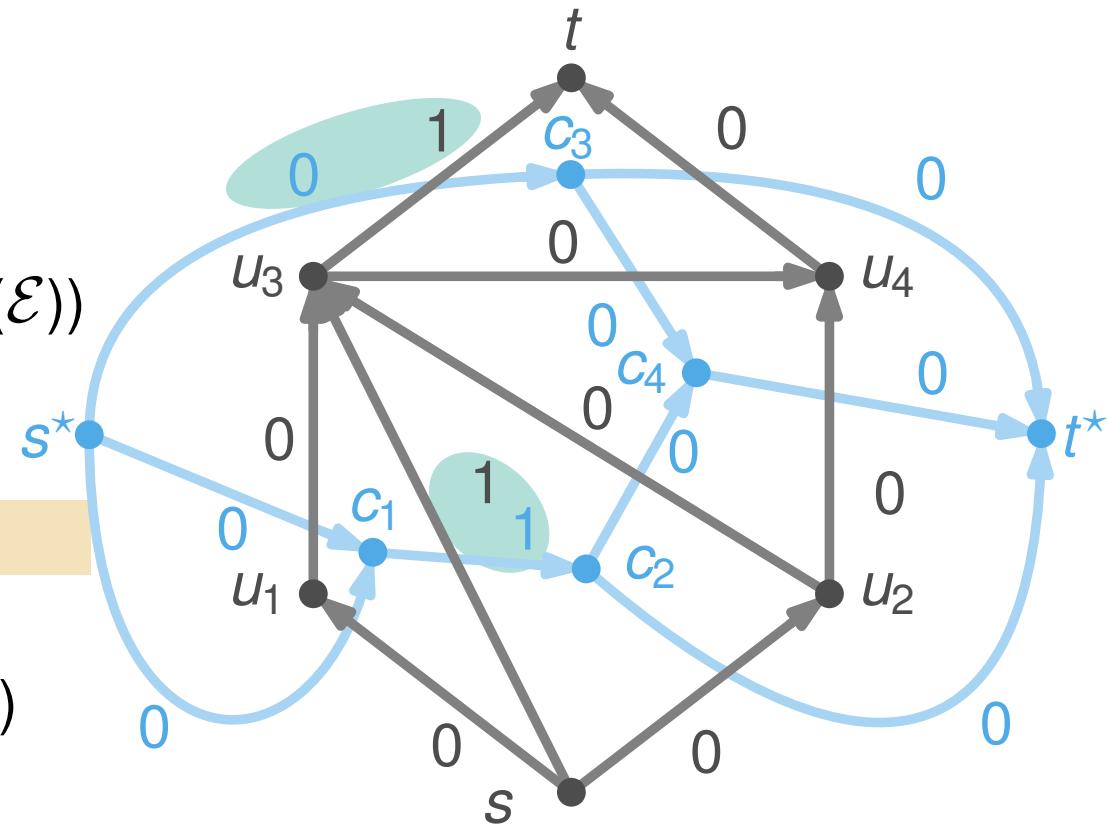
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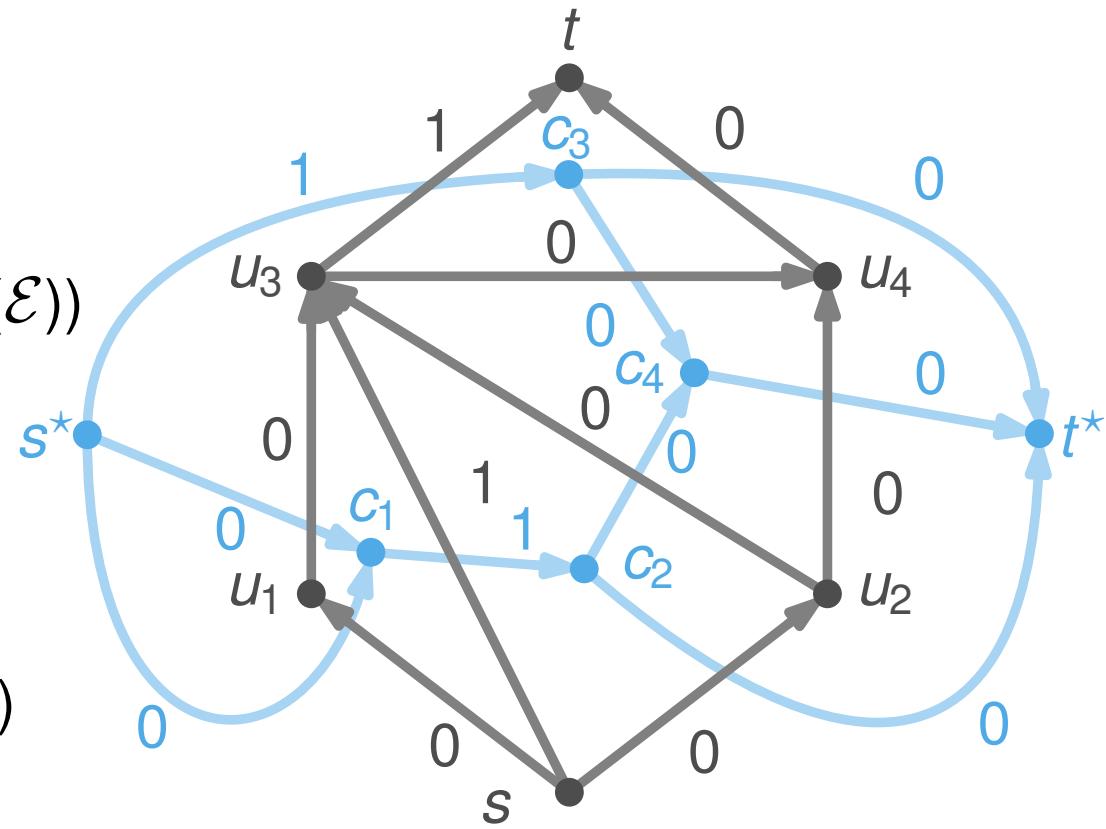
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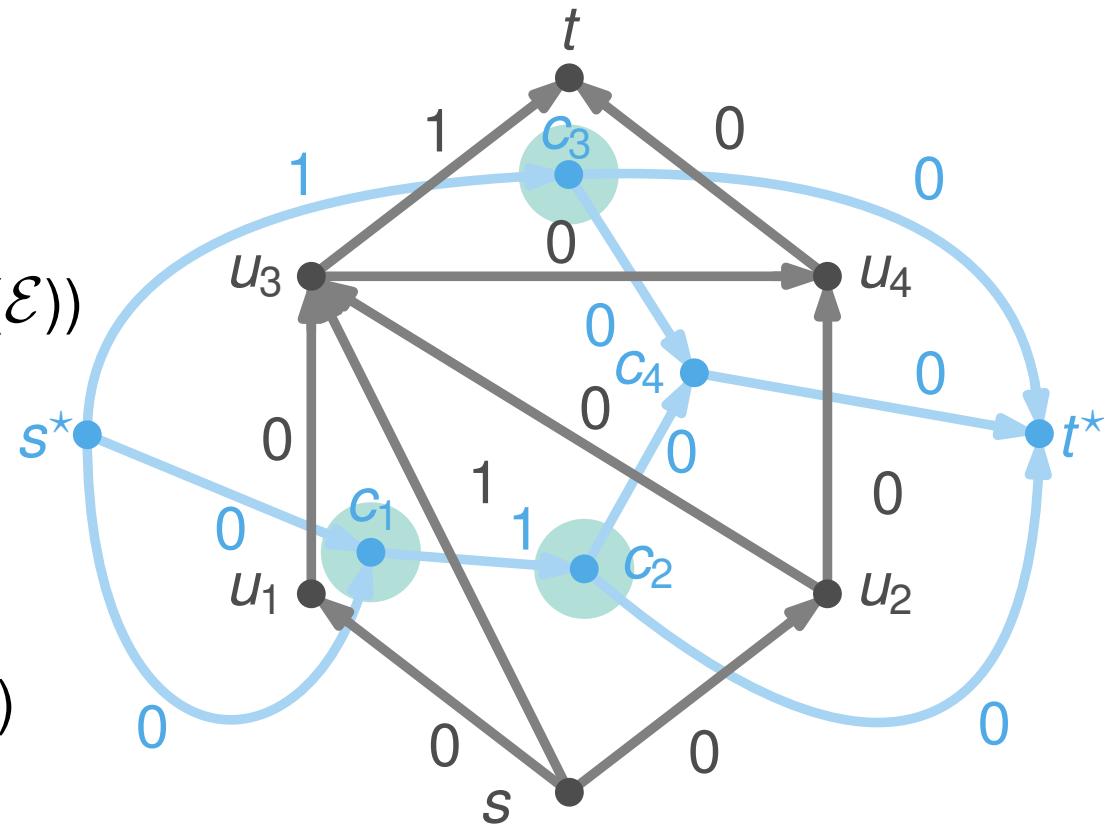
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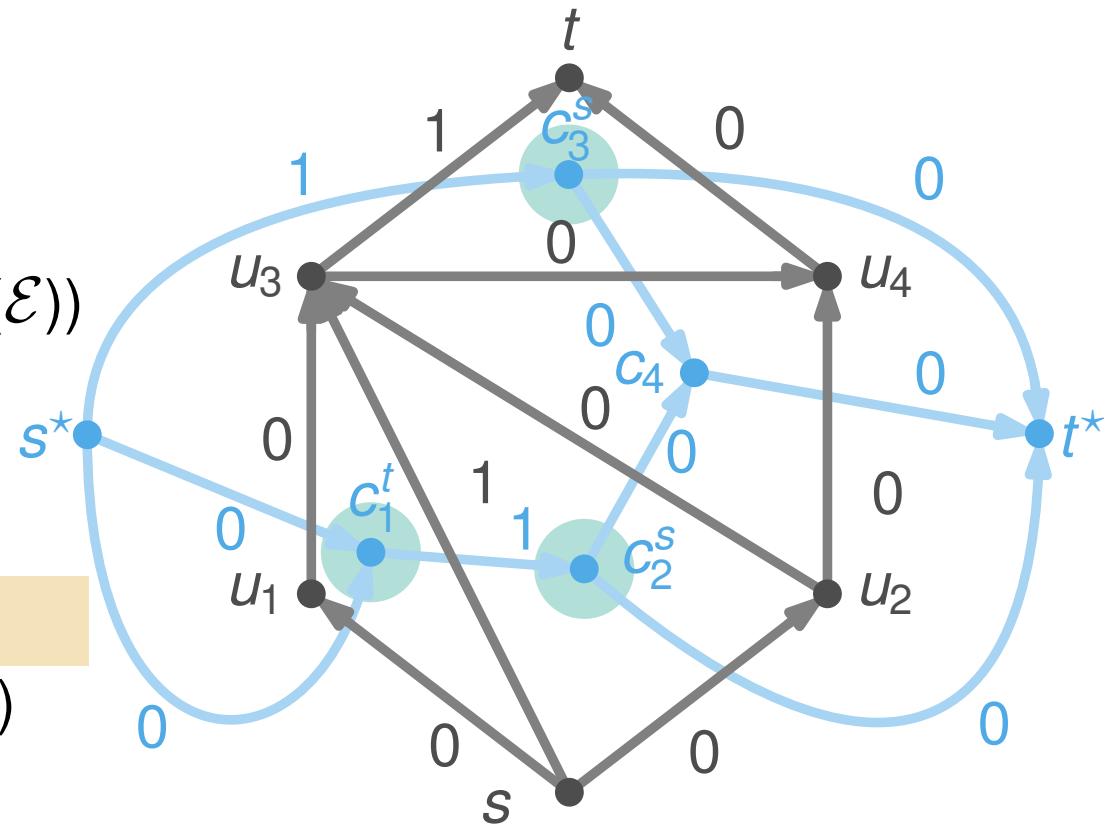
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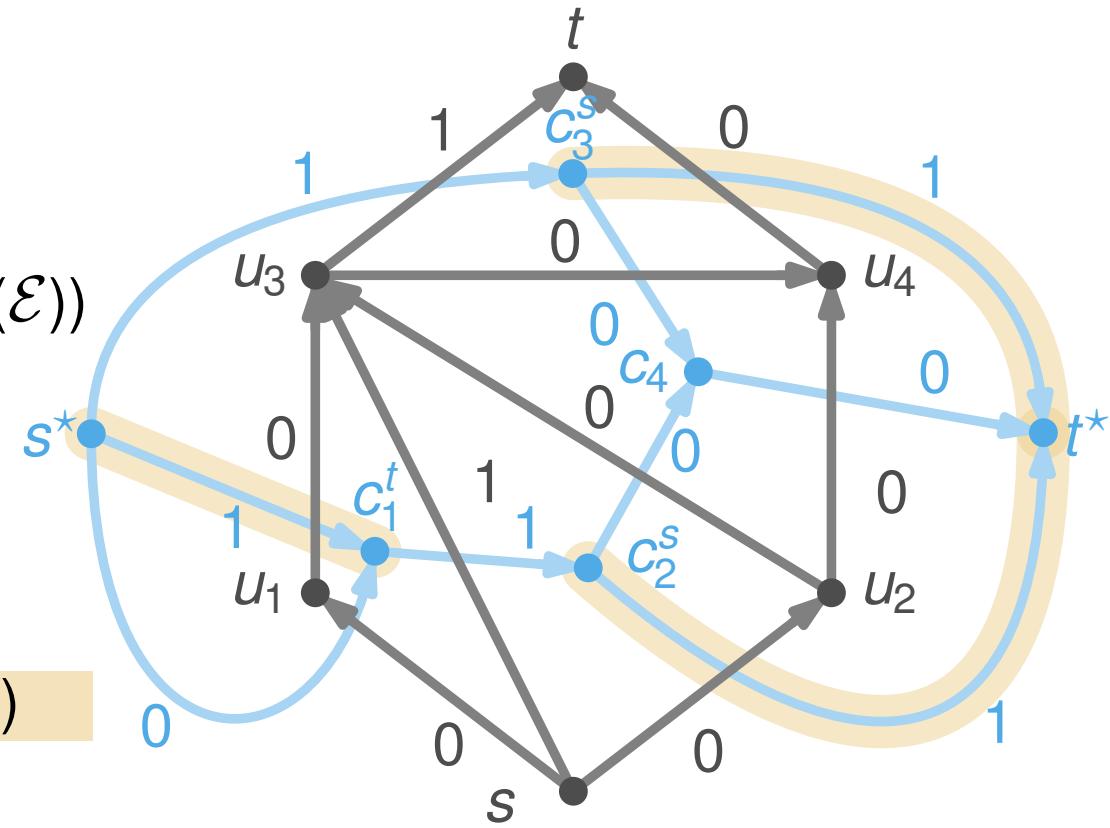
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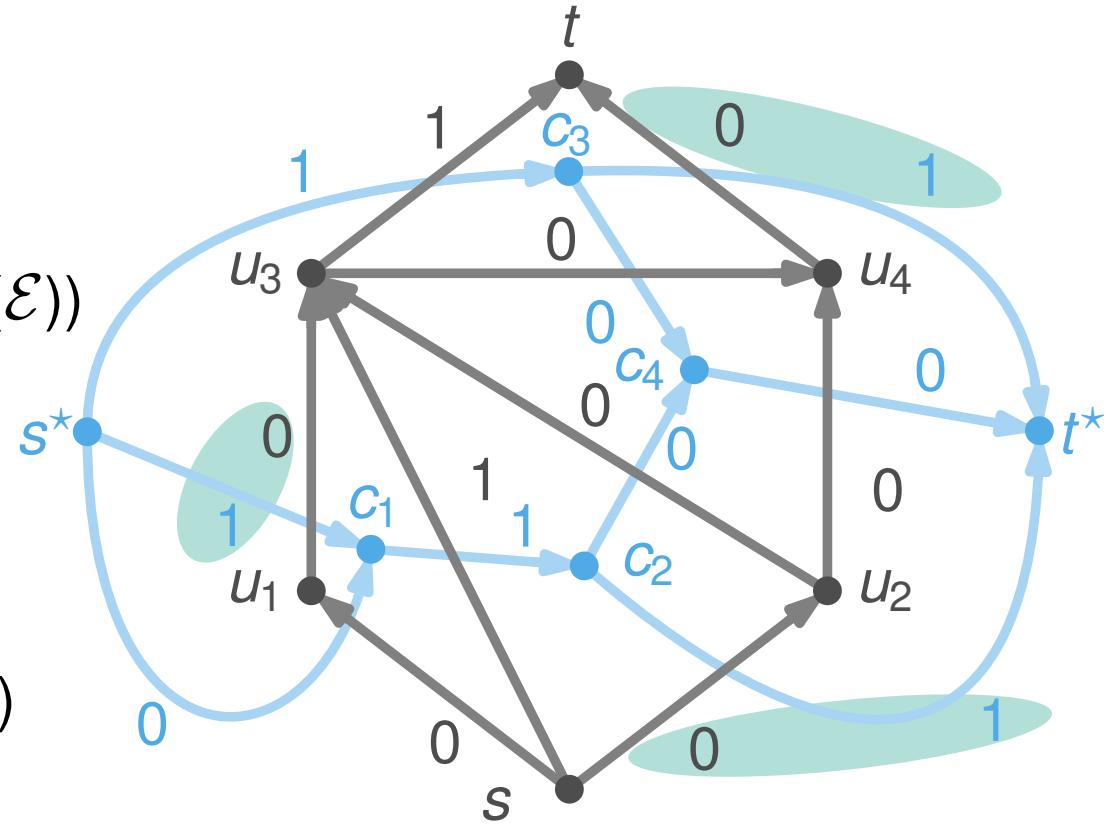
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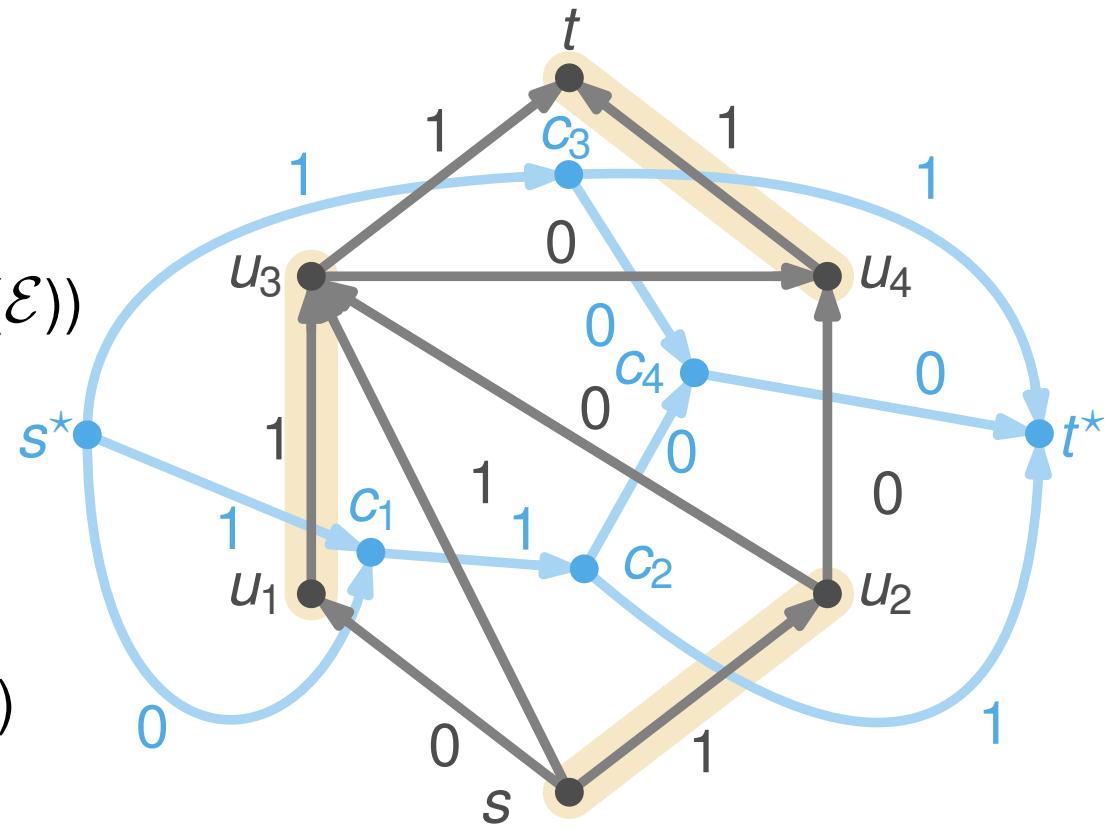
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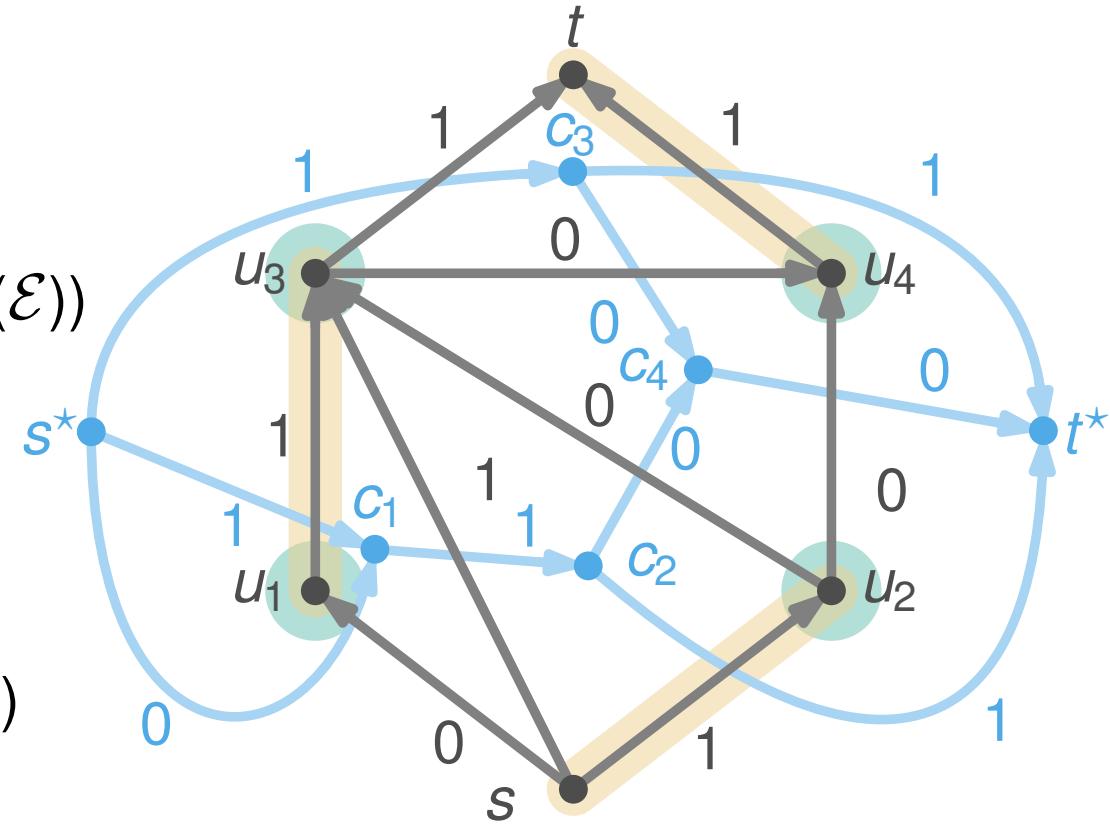
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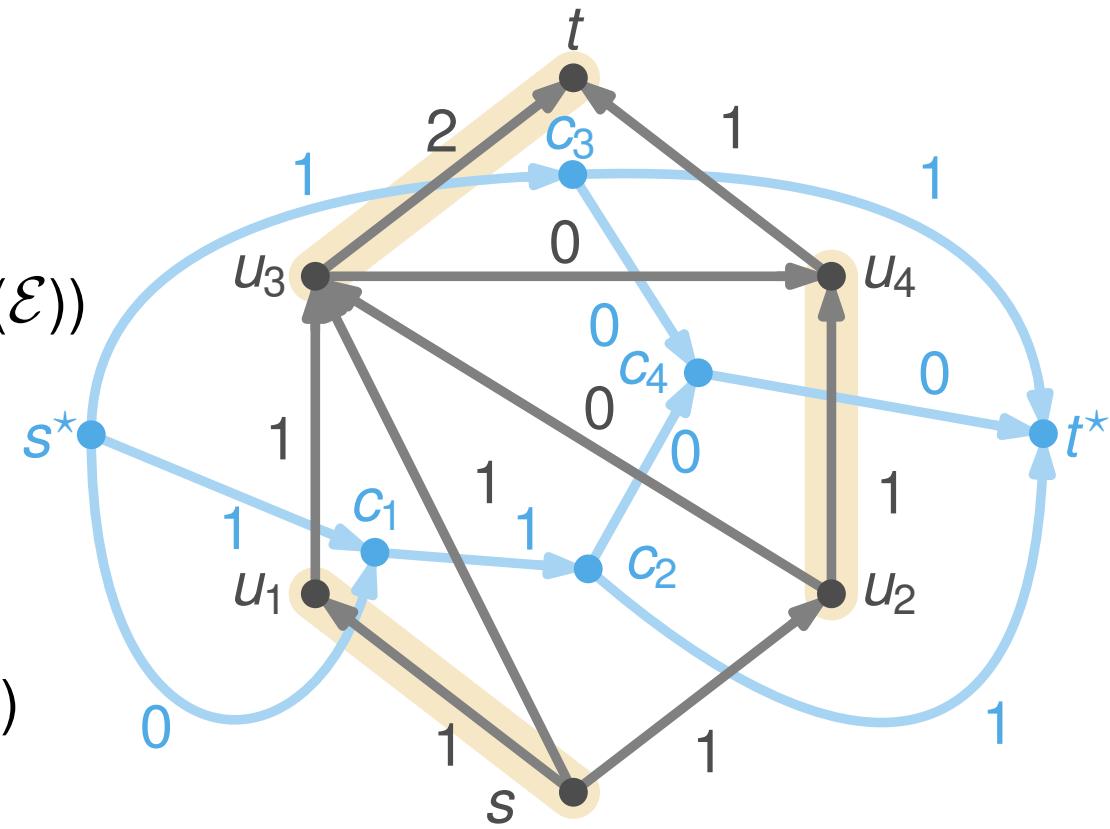
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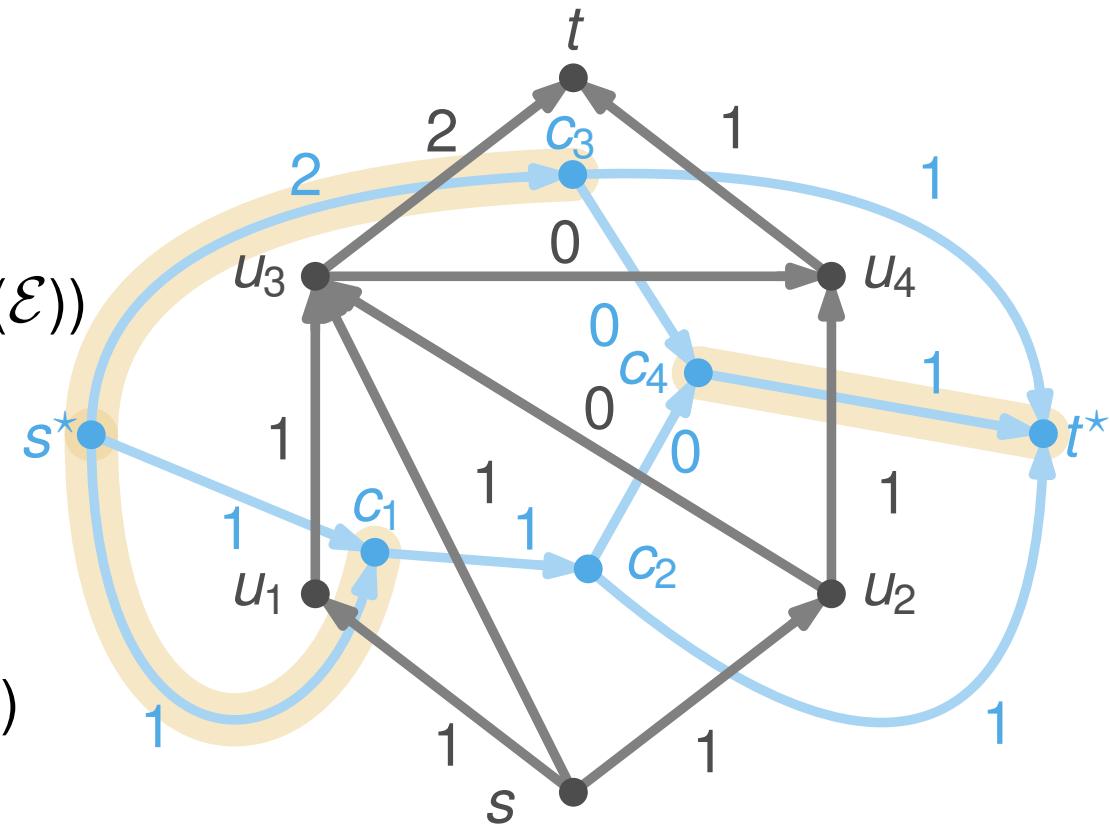
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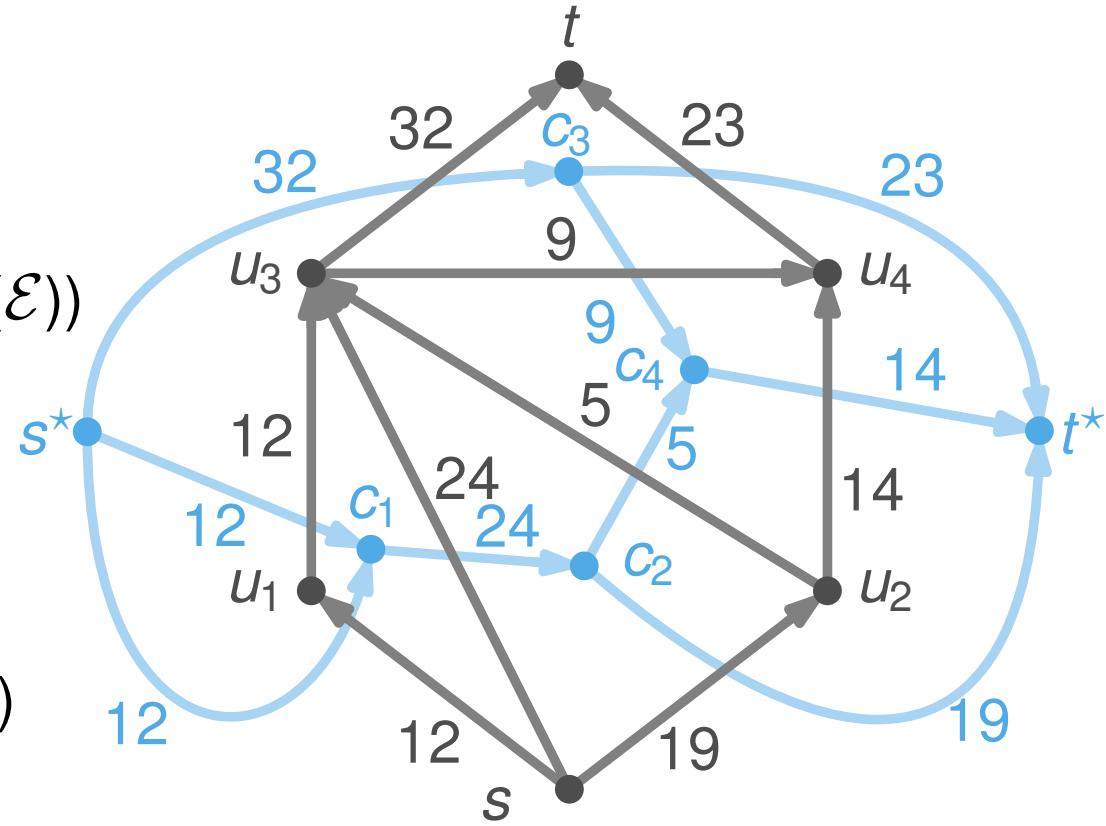
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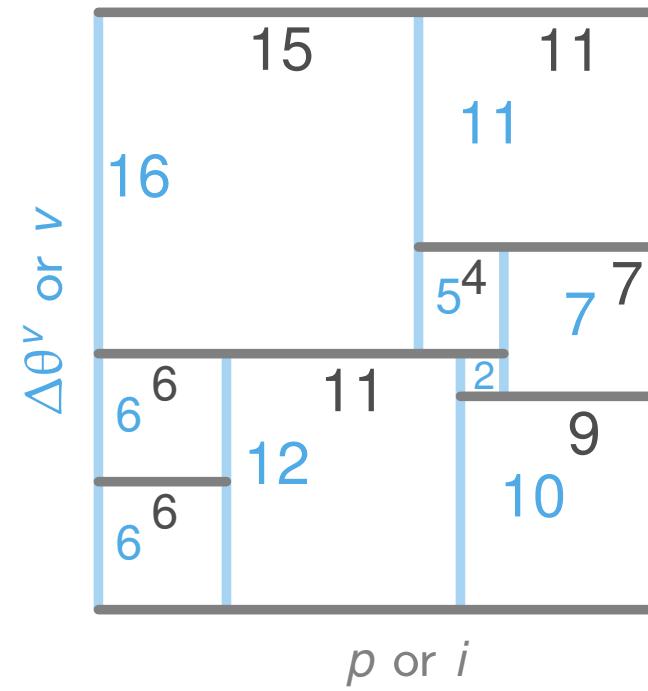
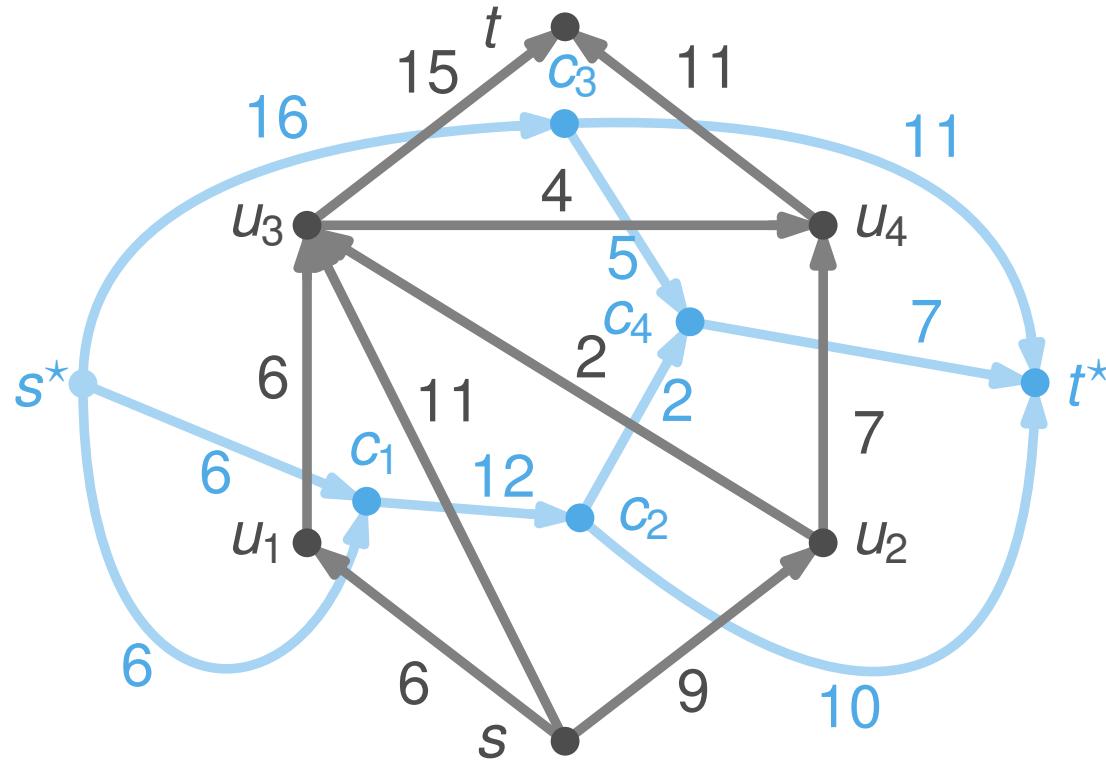
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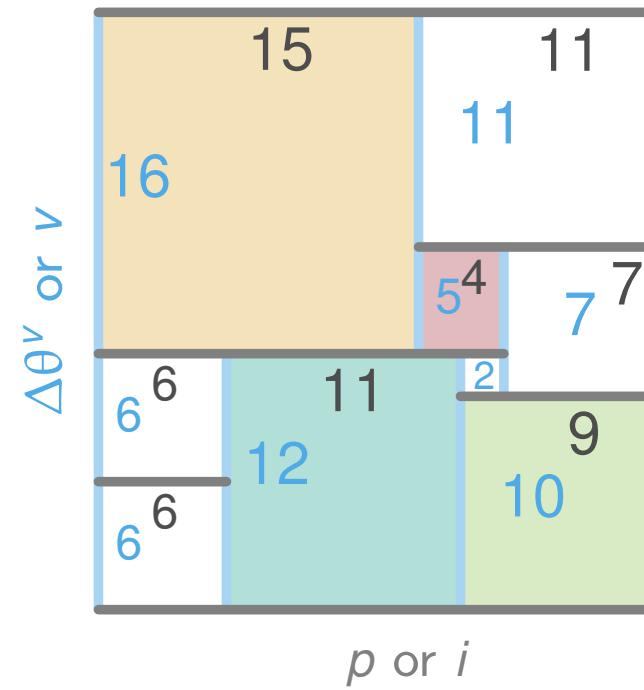
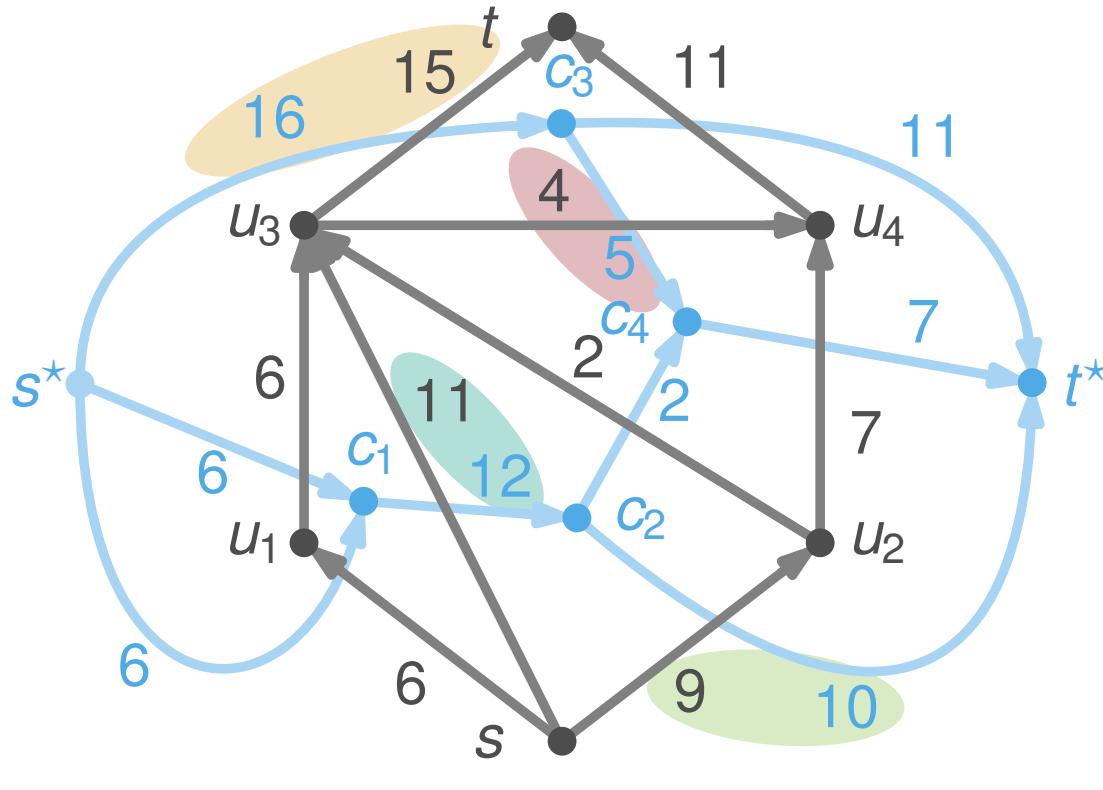
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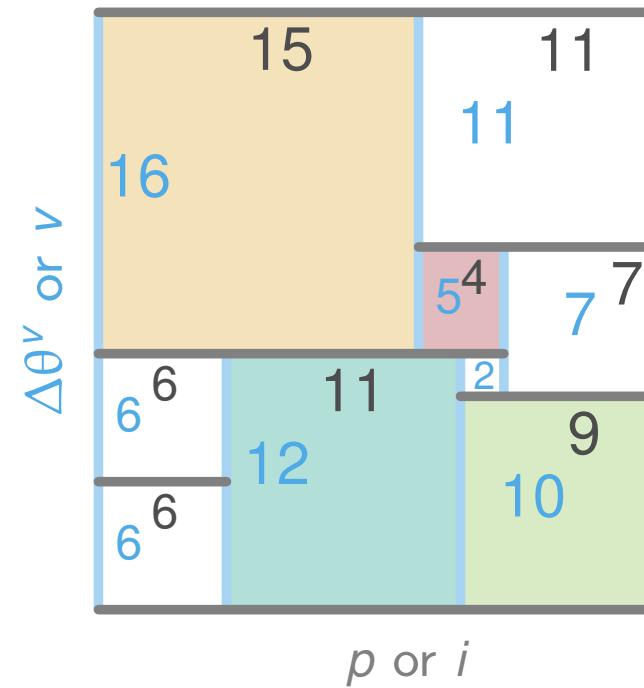
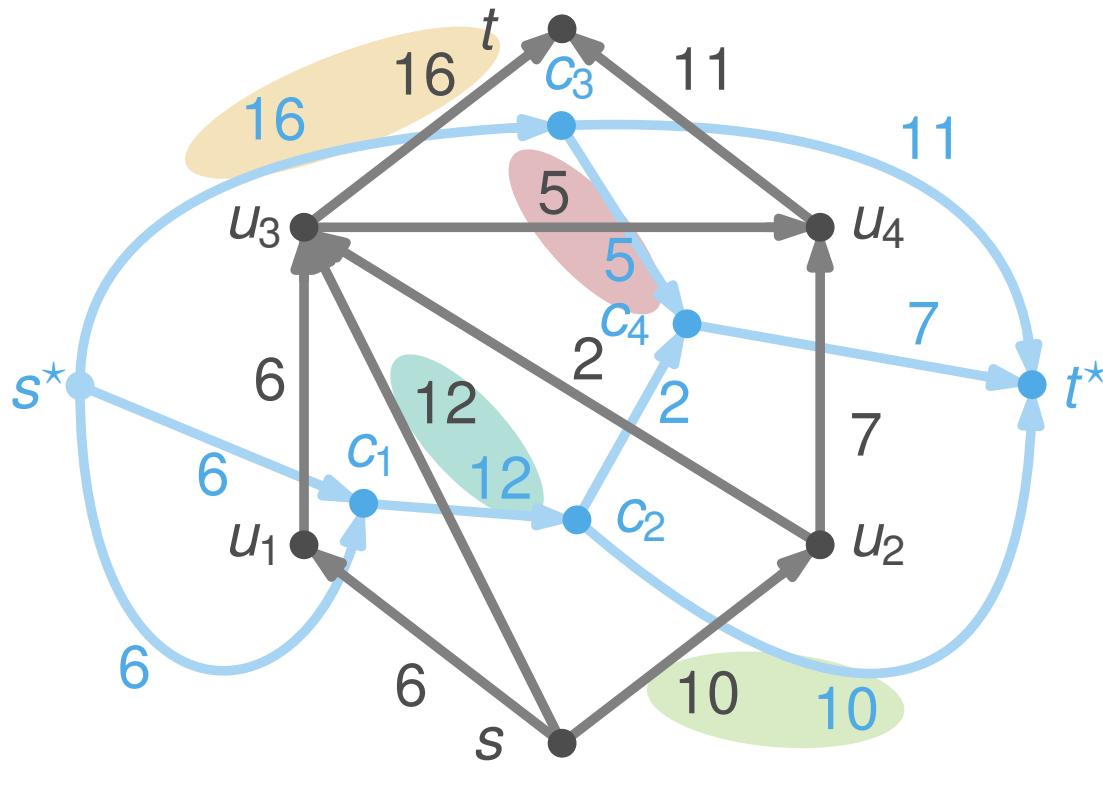
# Geometric Interpretation of a KCL Conflict



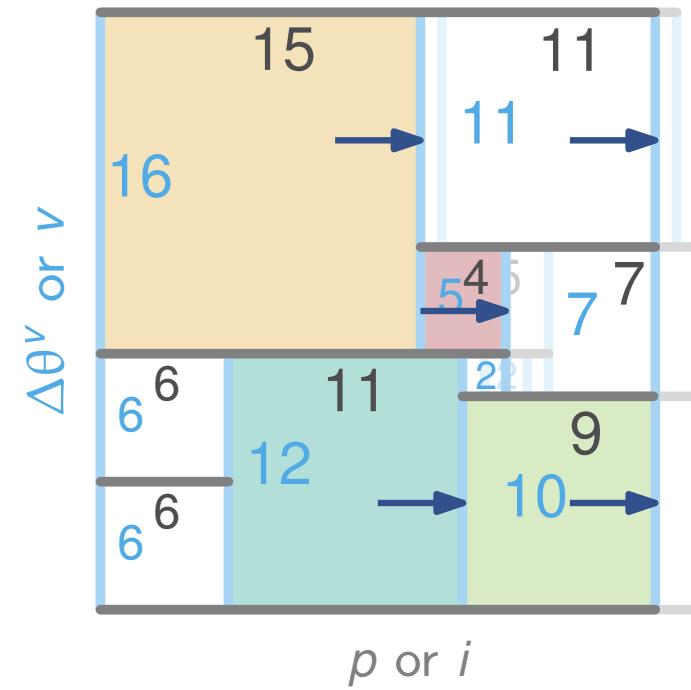
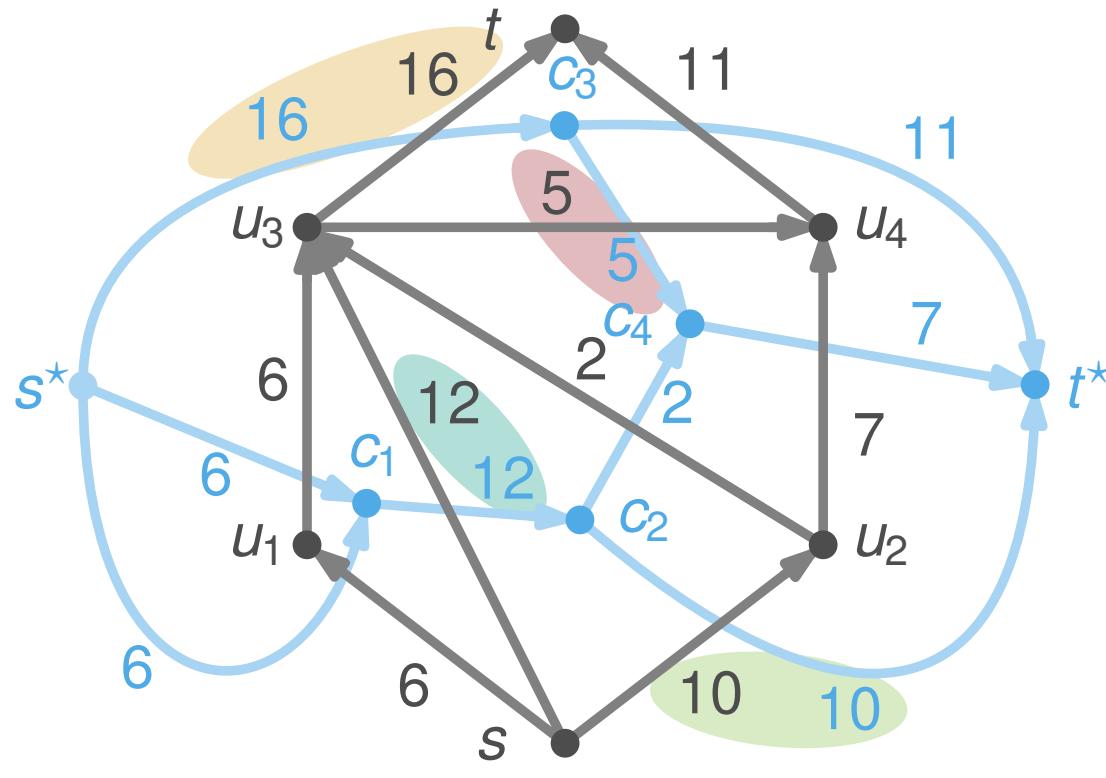
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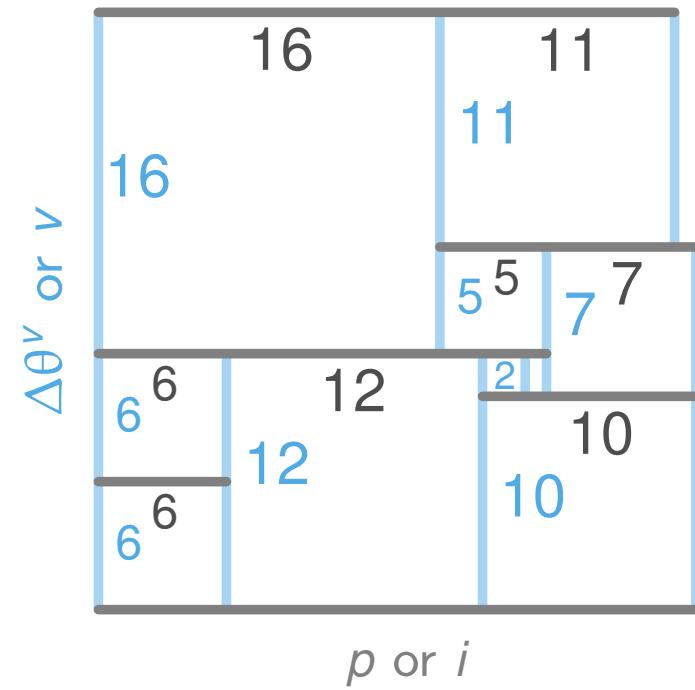
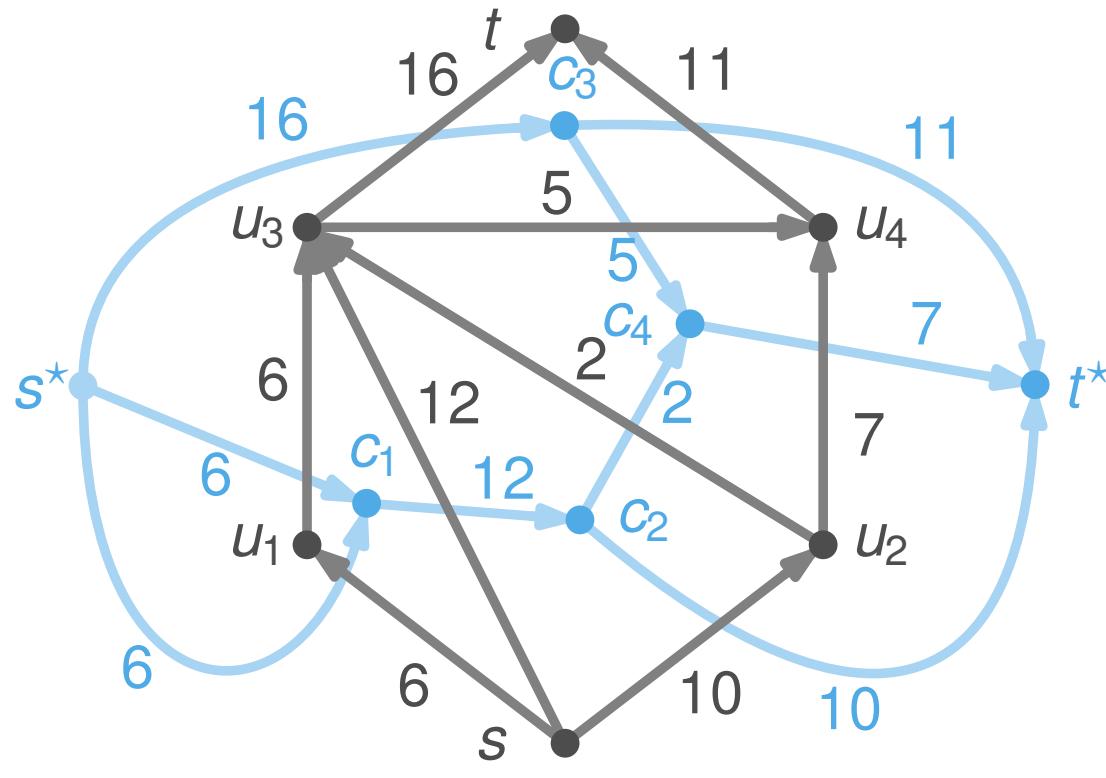
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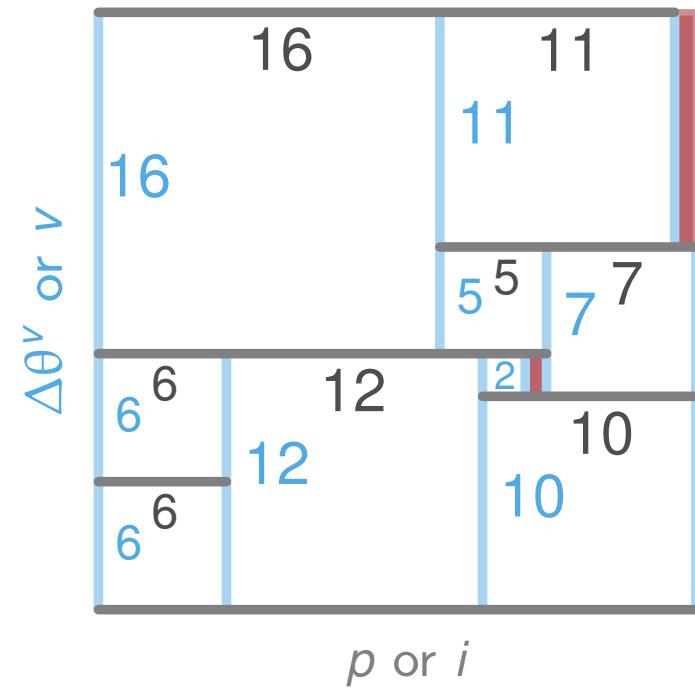
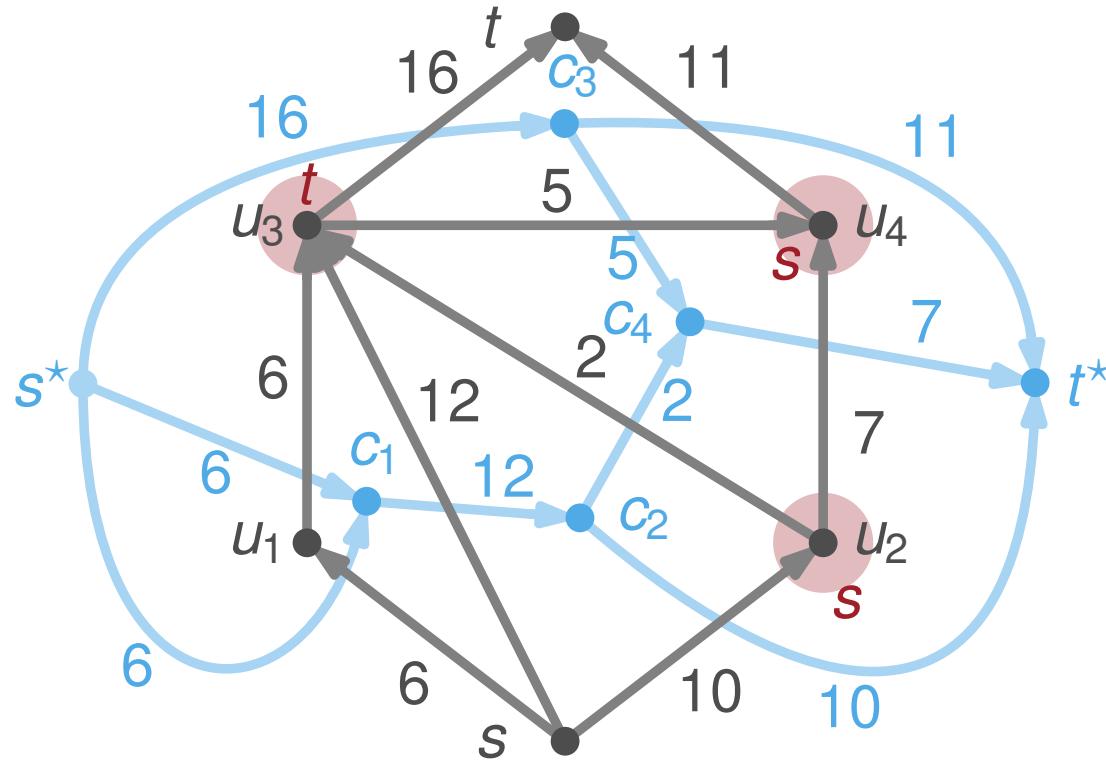
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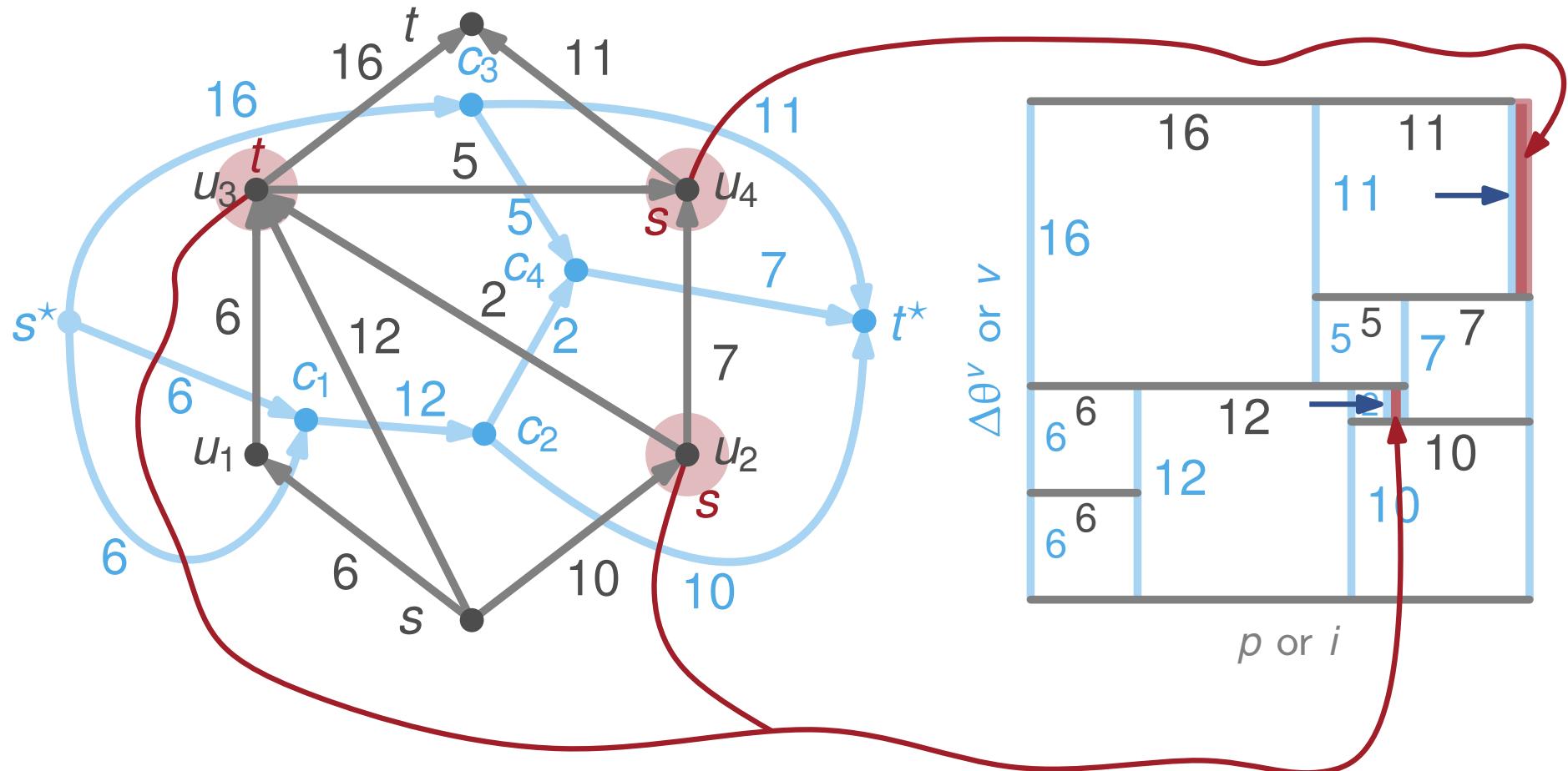
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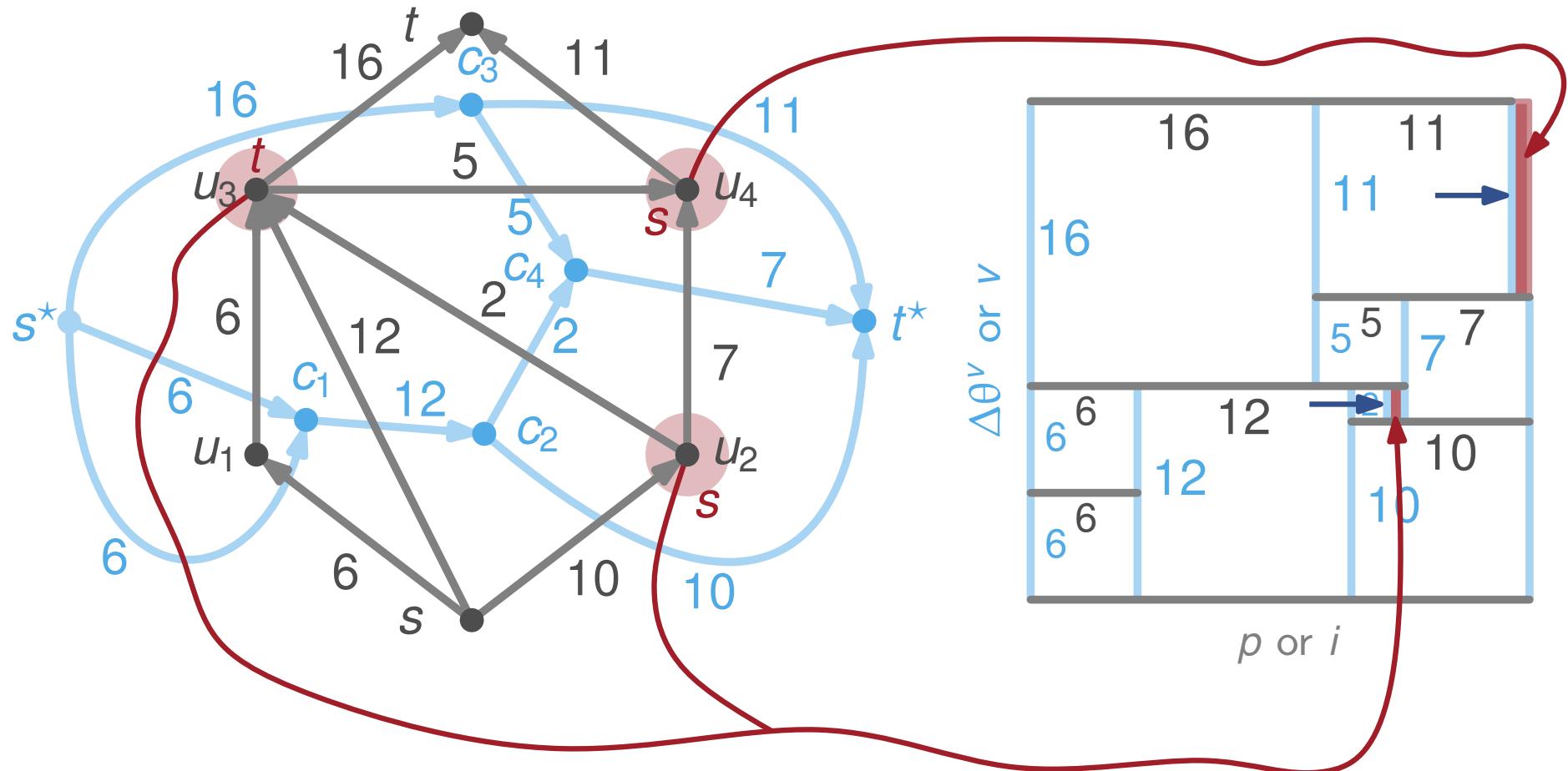
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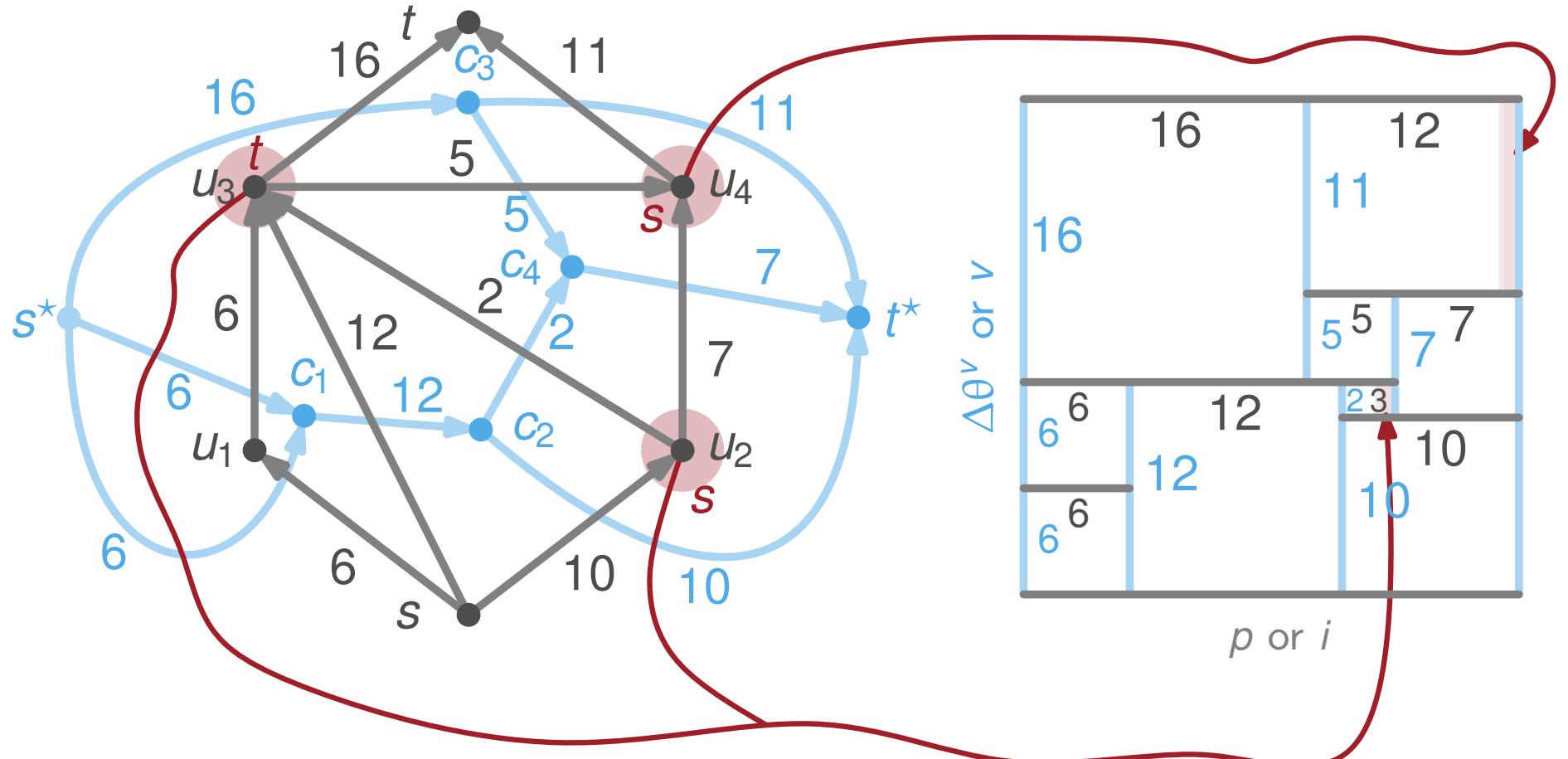
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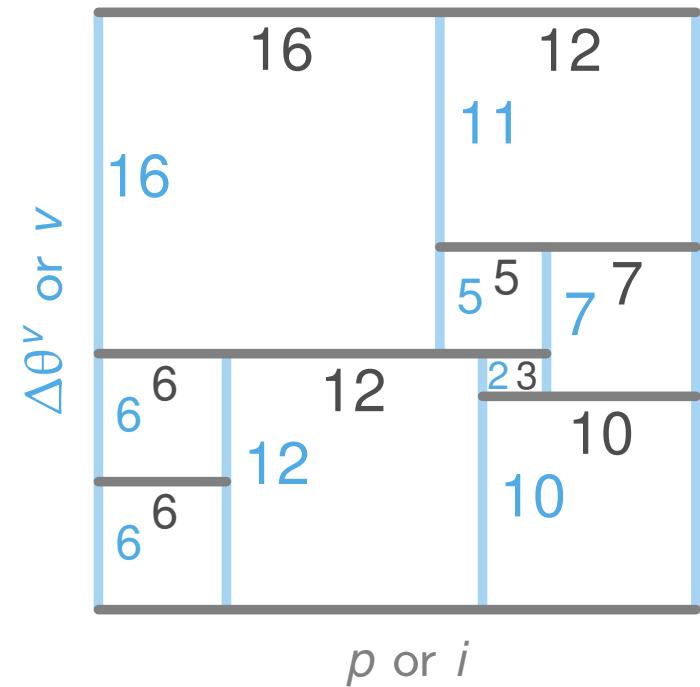
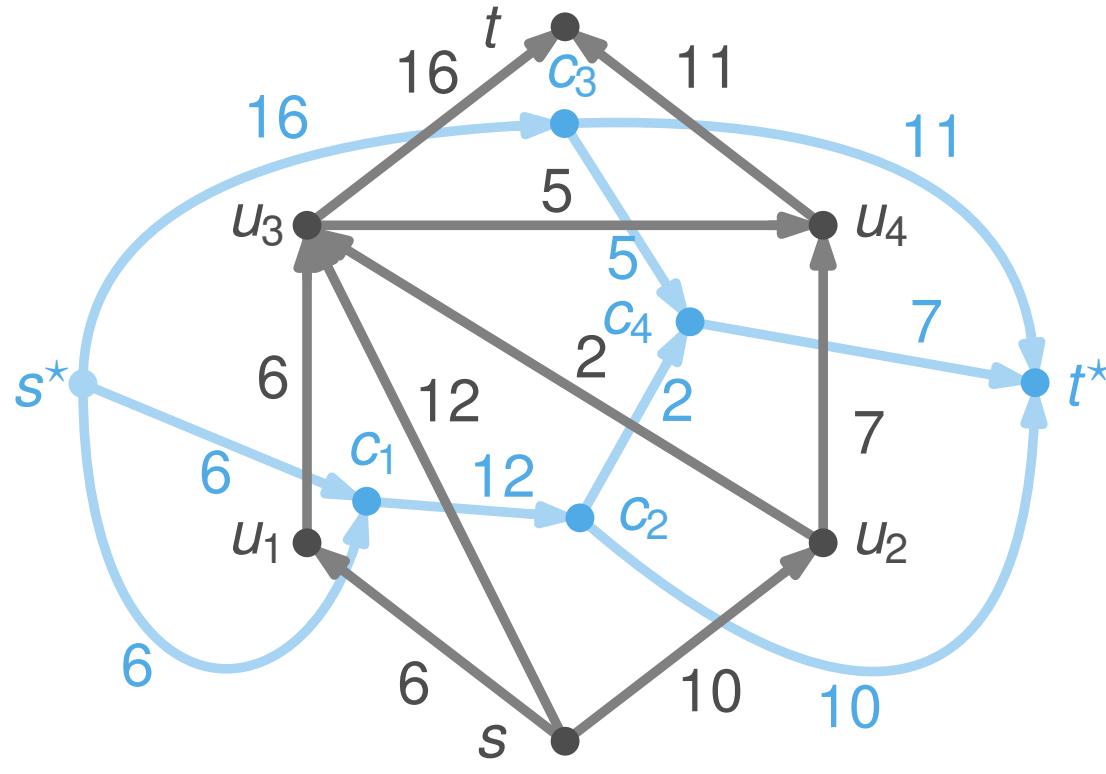
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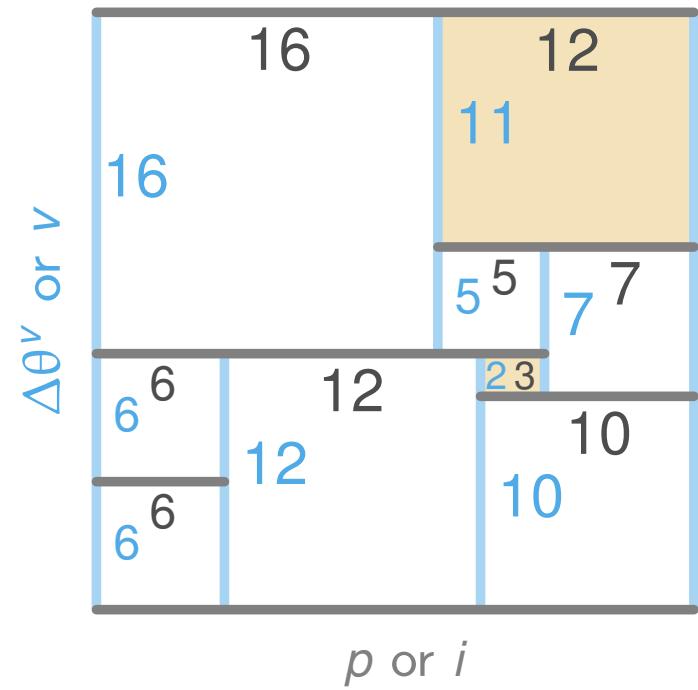
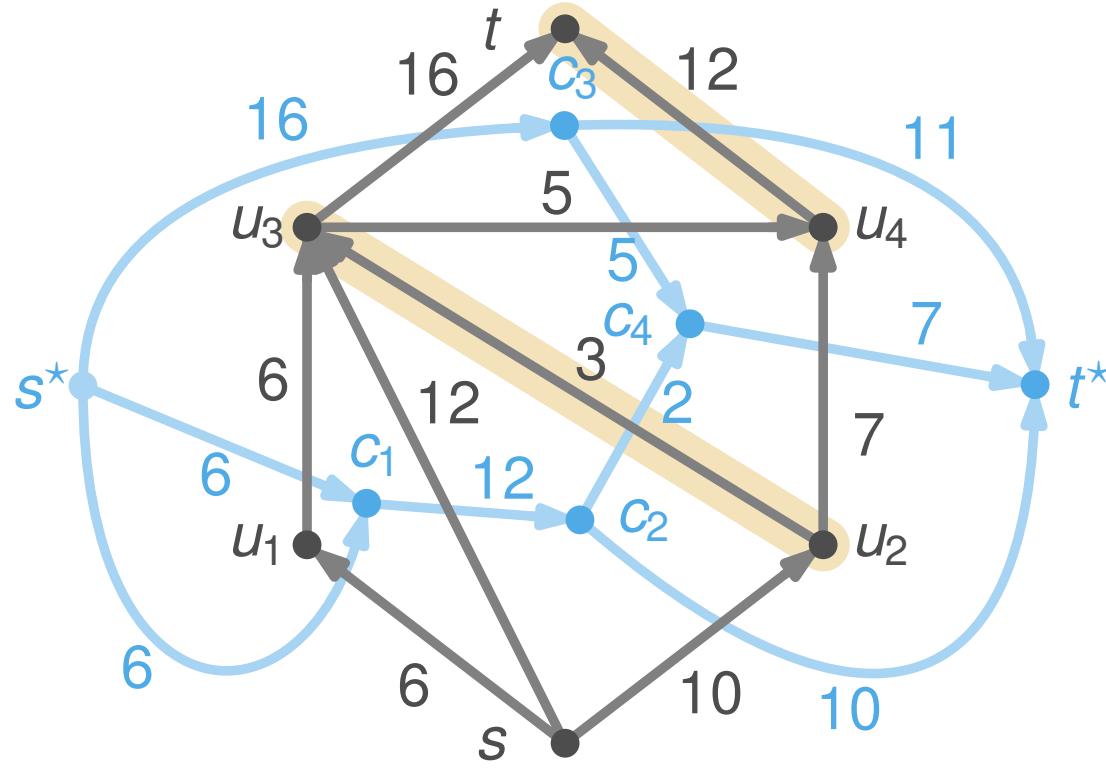
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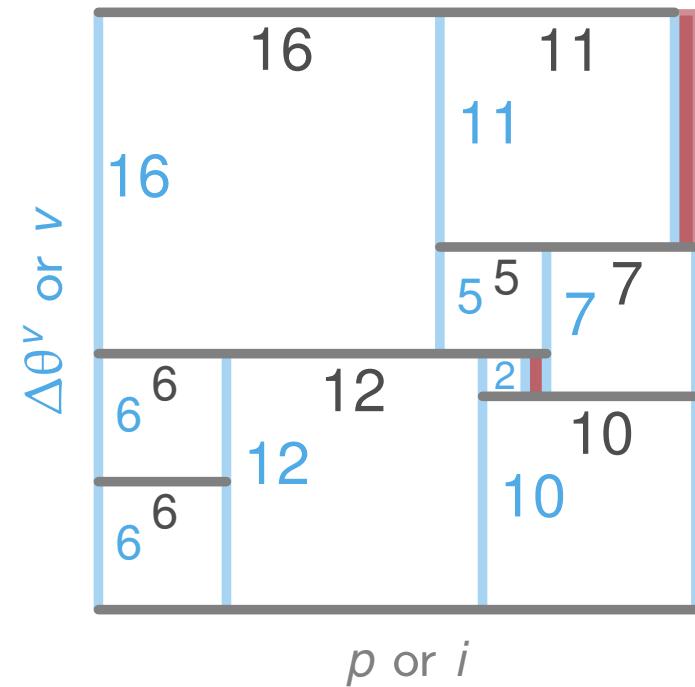
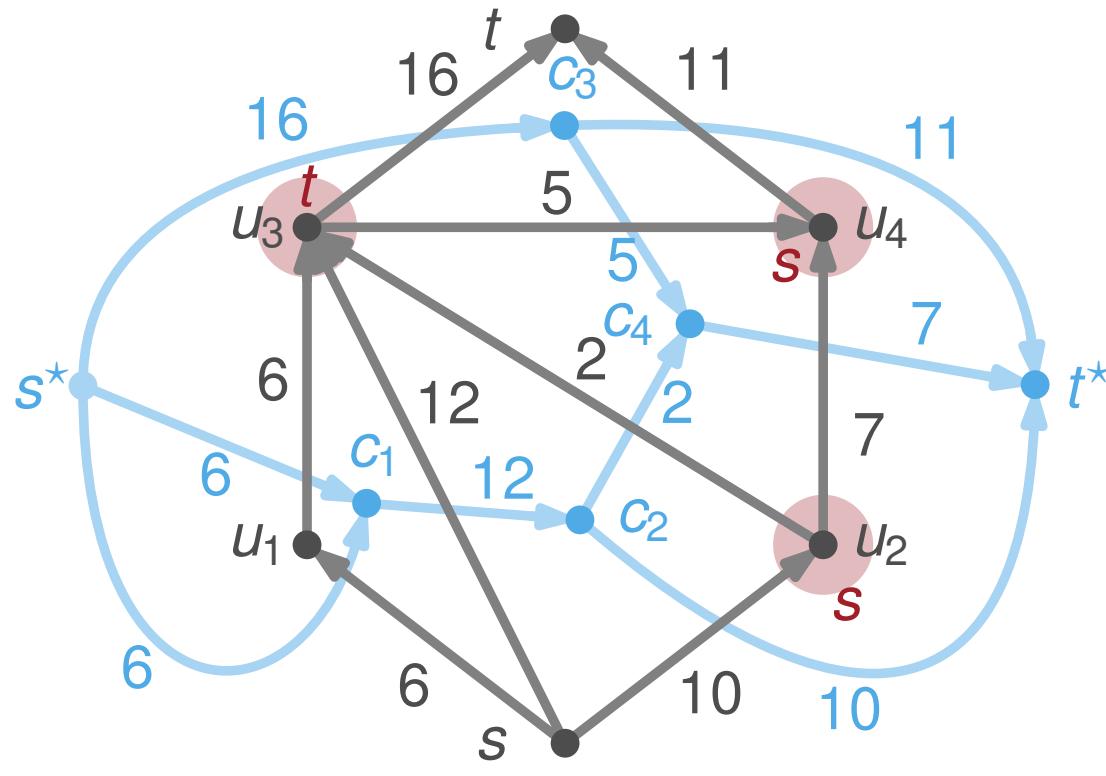
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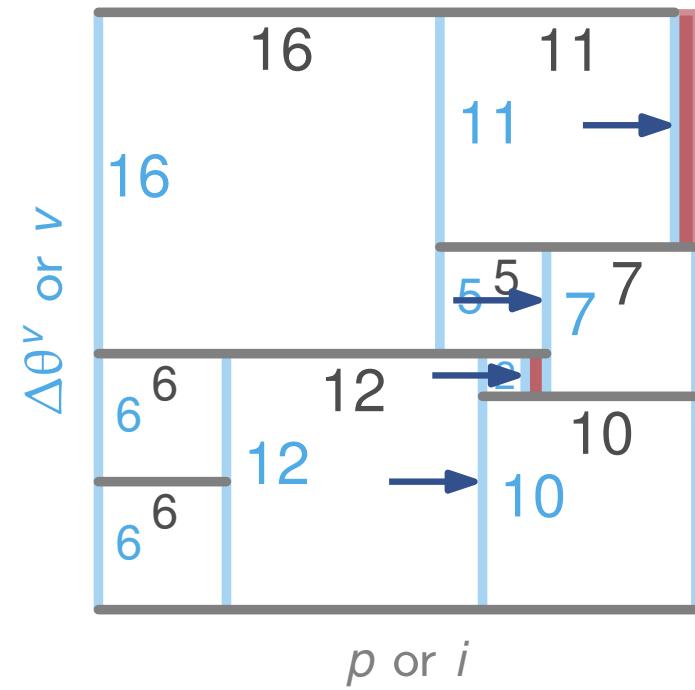
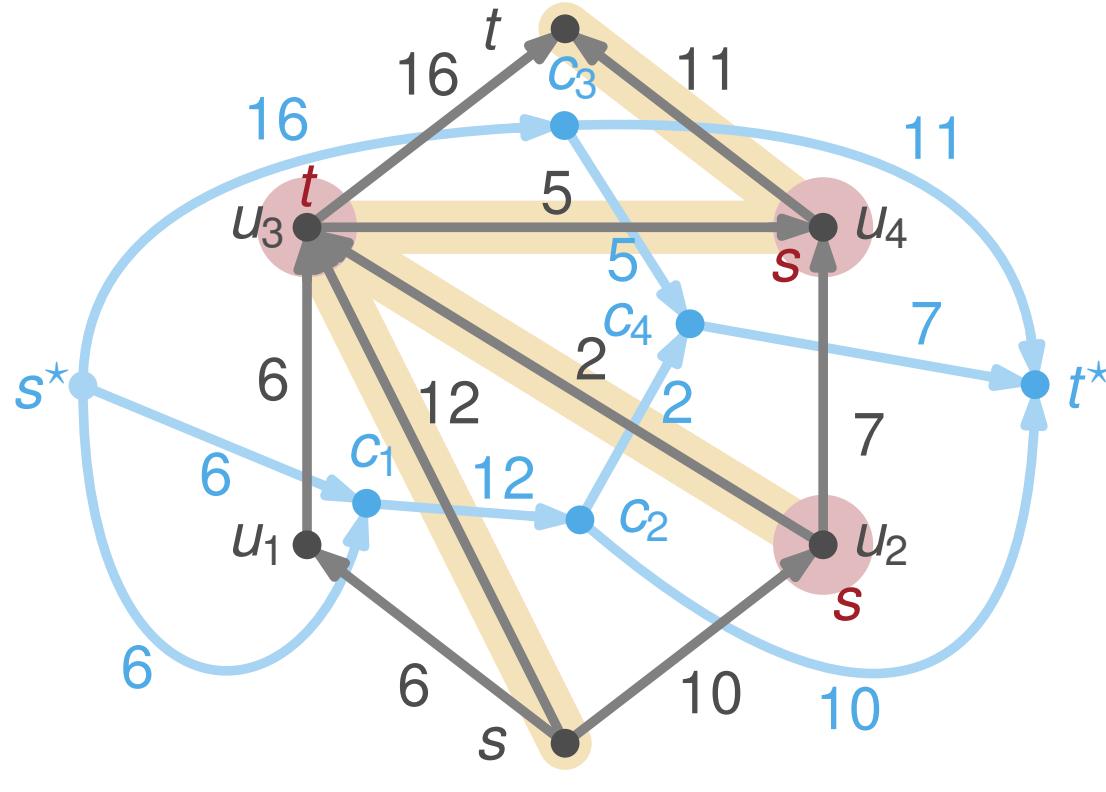
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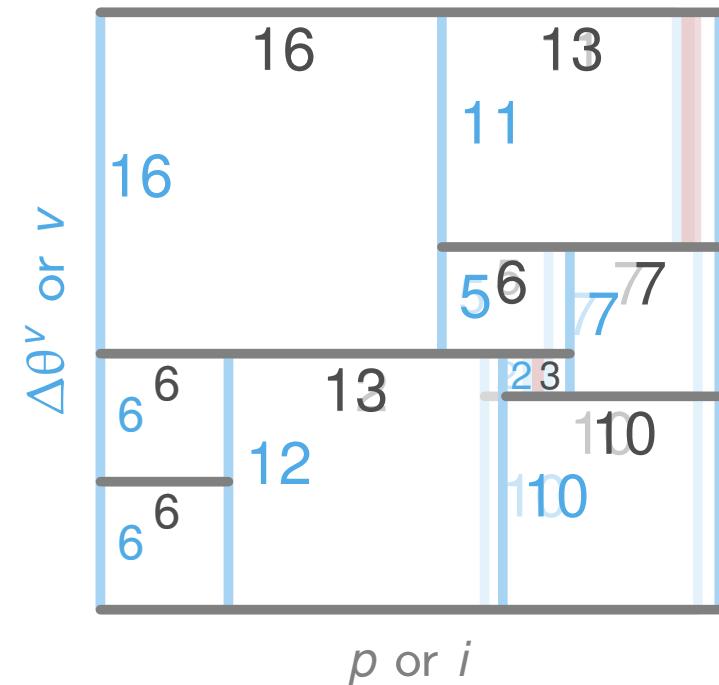
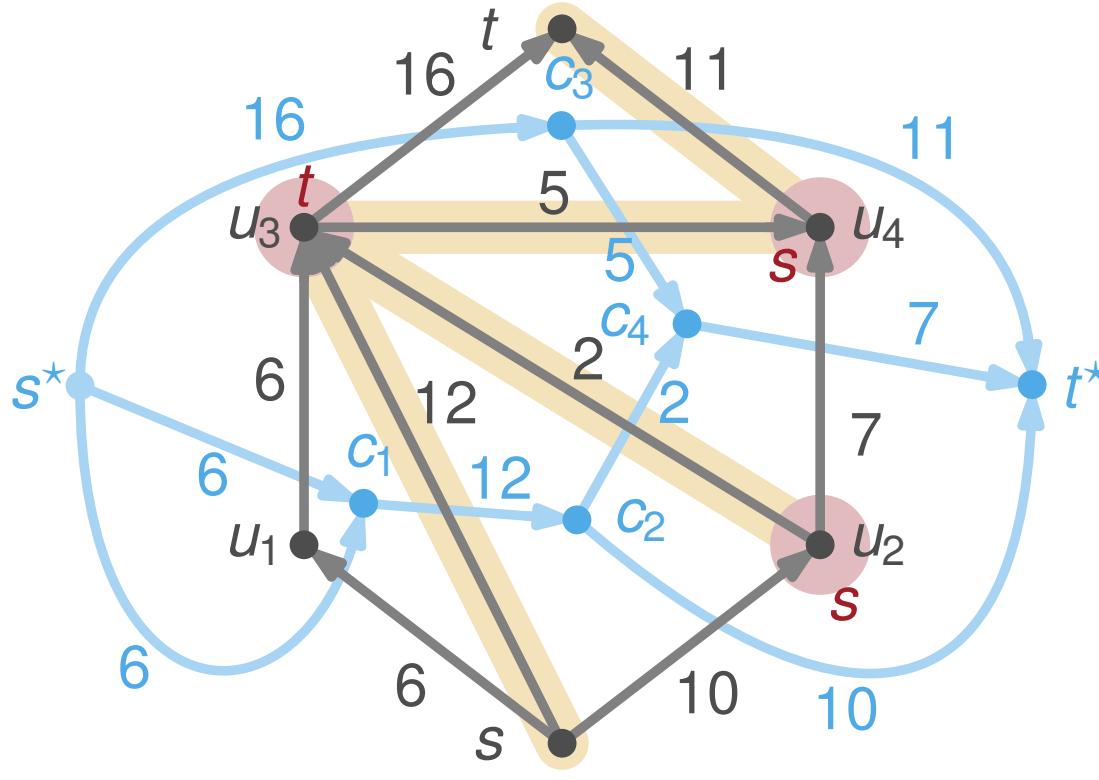
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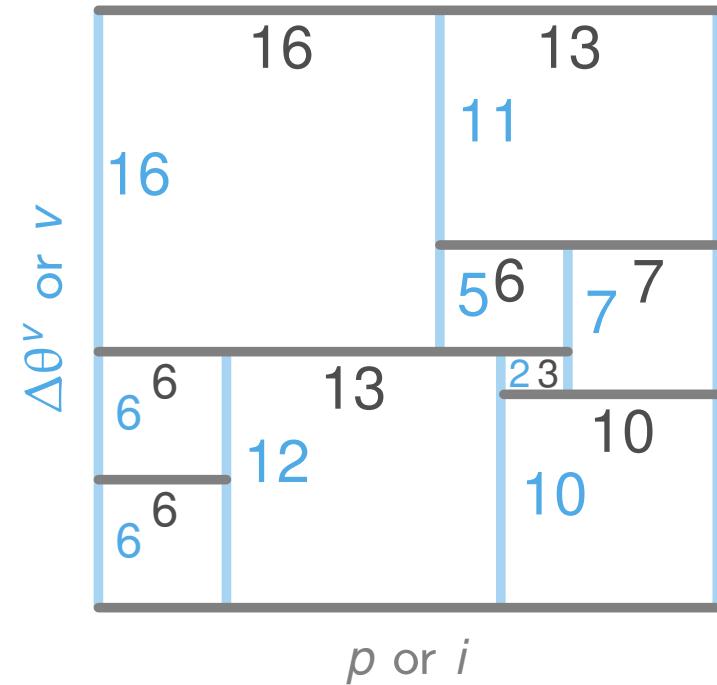
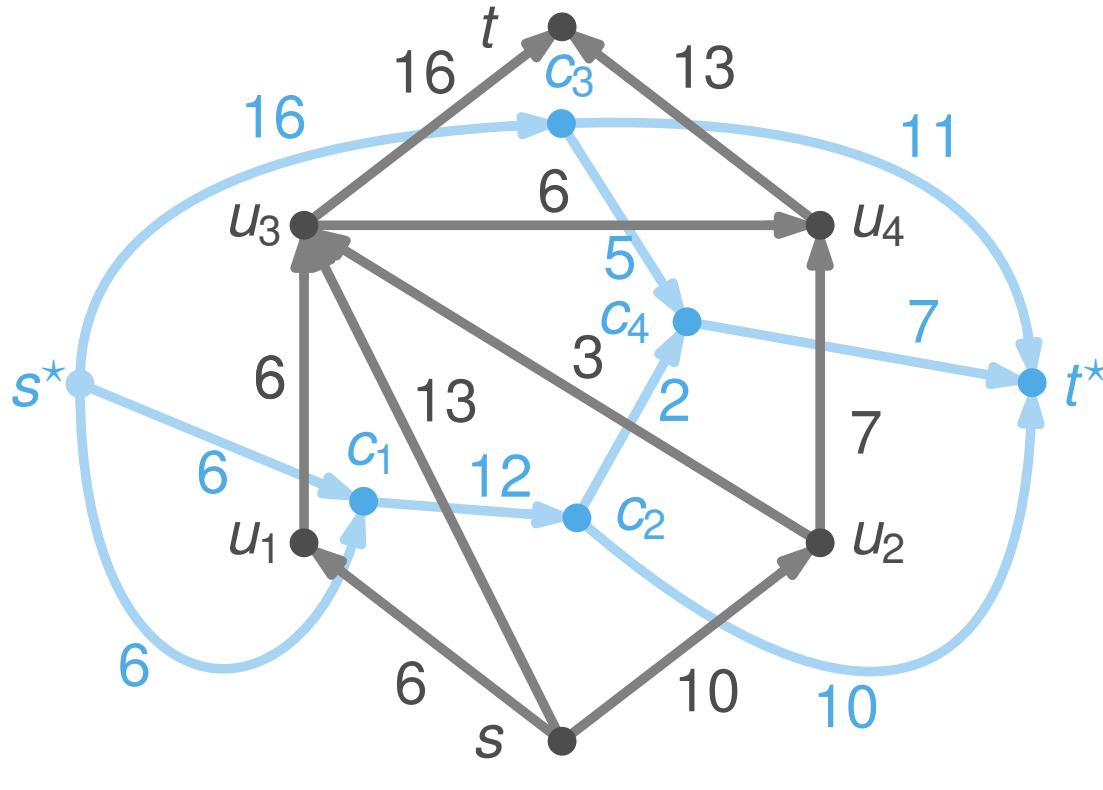
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## Observation 5 [Resolve KCL Conflicts]

Minimize in each `resolveConflict` step the total resizing of the outer rectangle, since a too large increase might skip a valid solution.

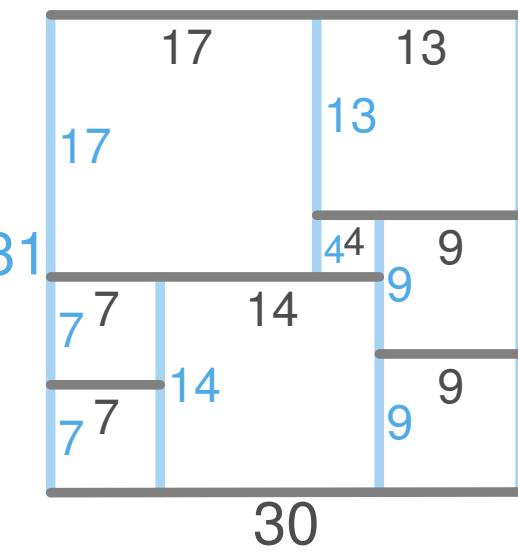
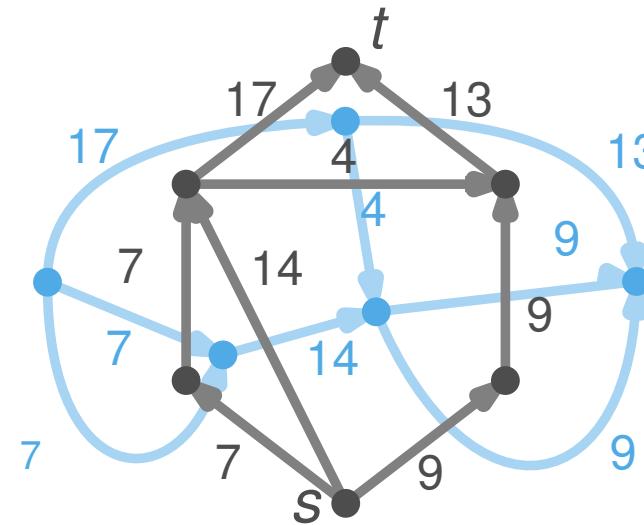
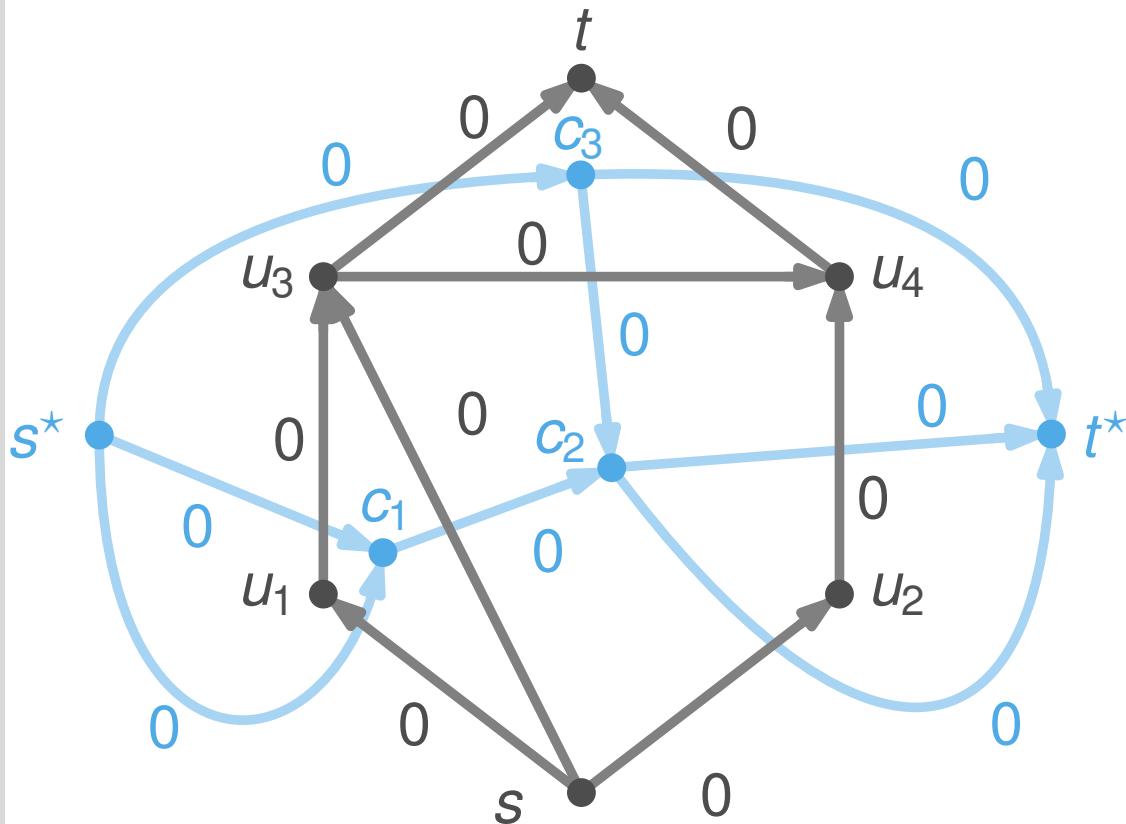
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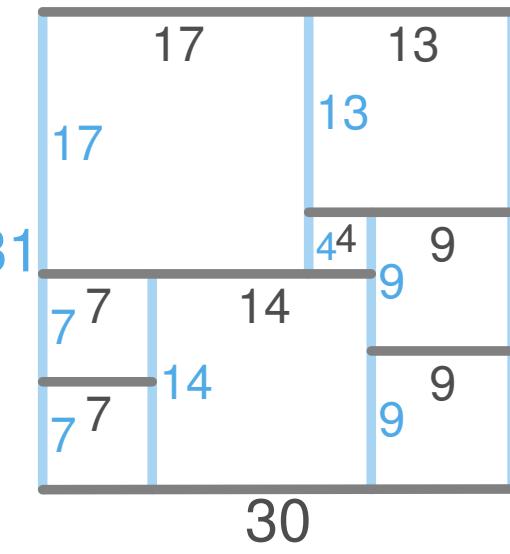
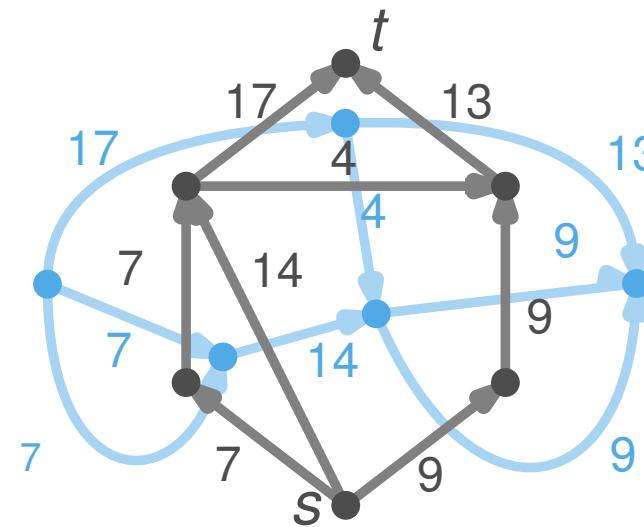
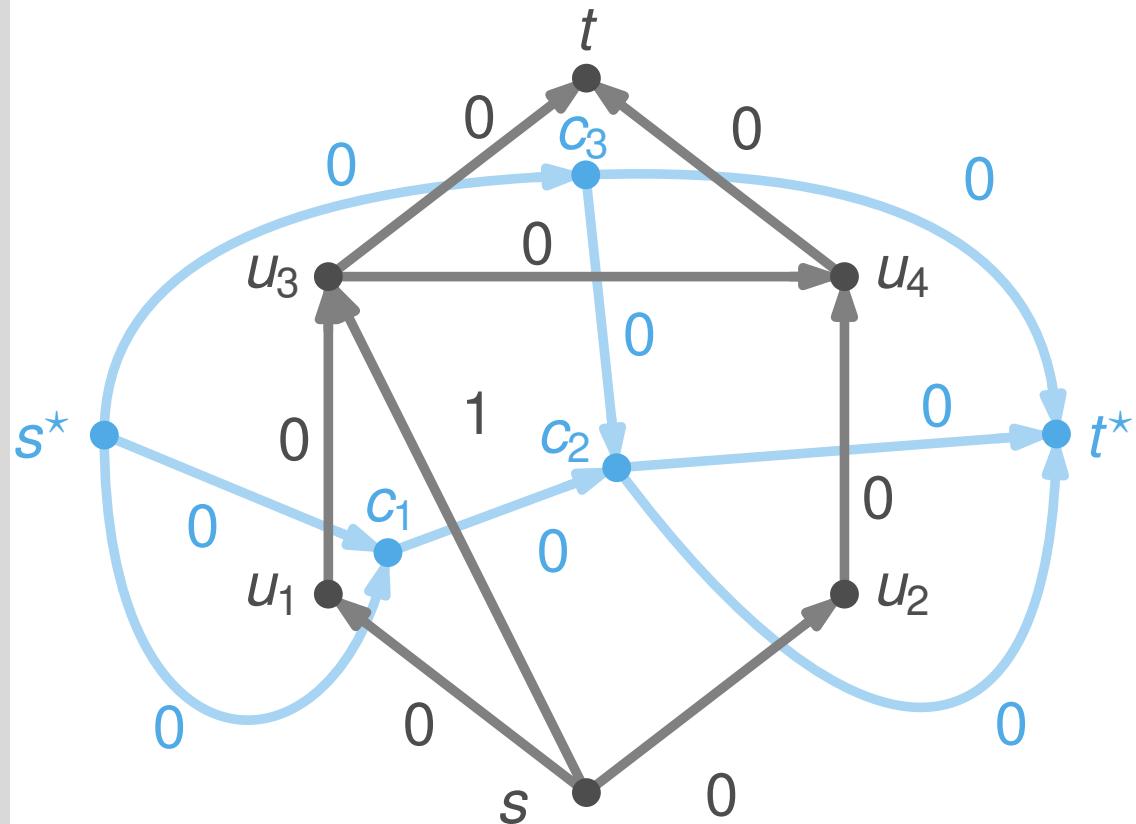
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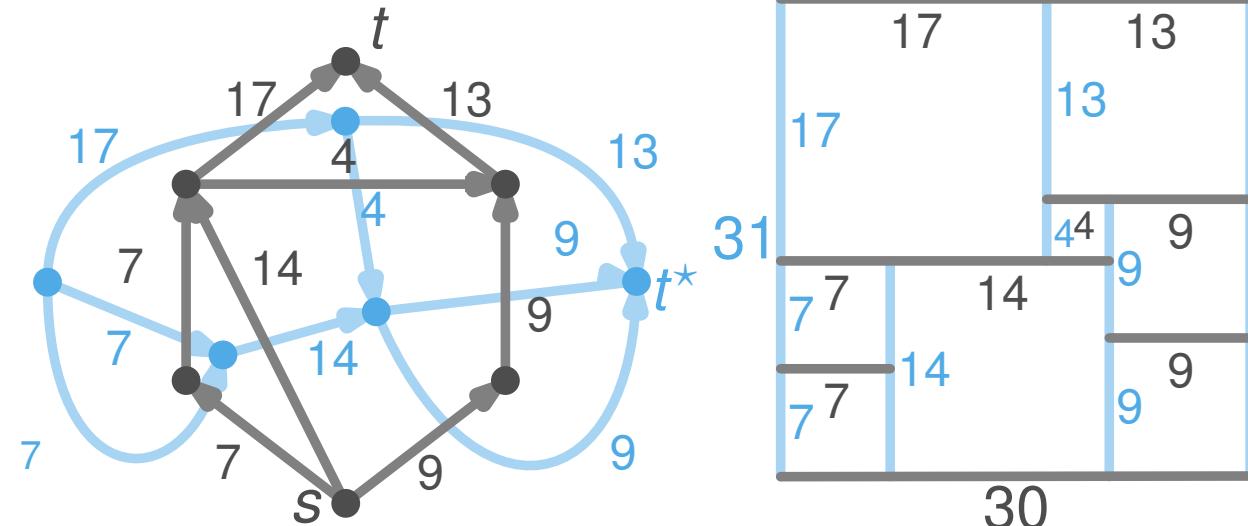
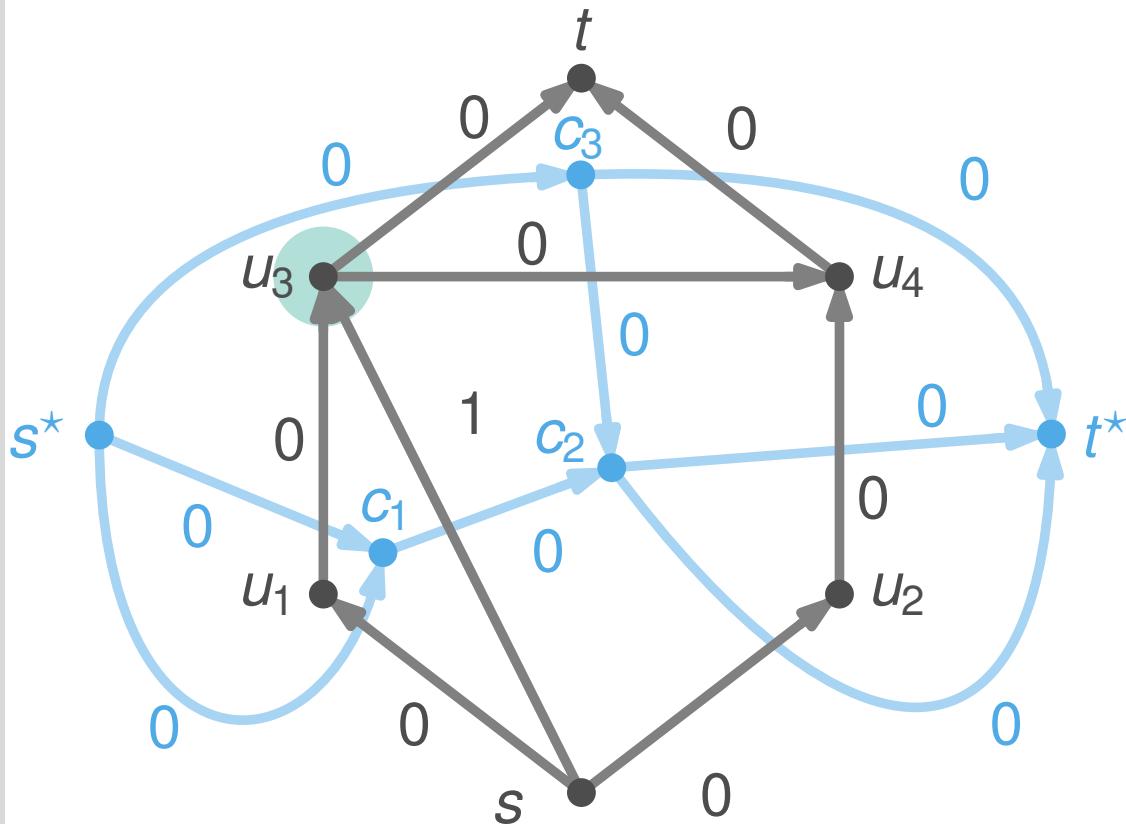
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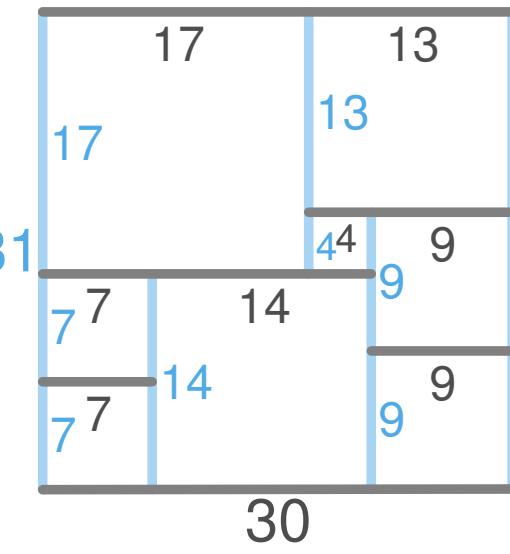
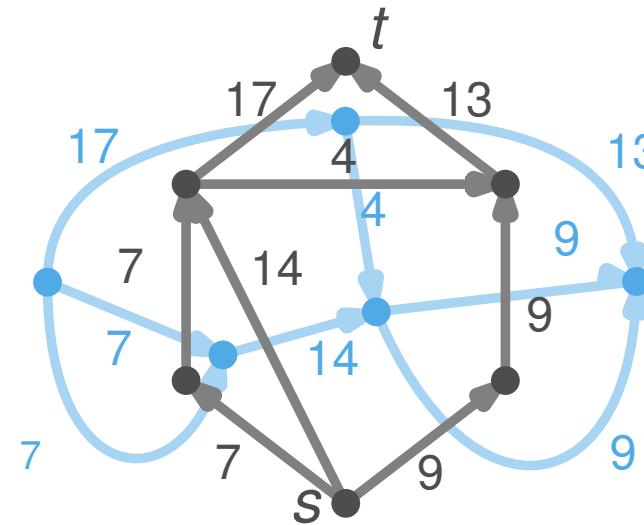
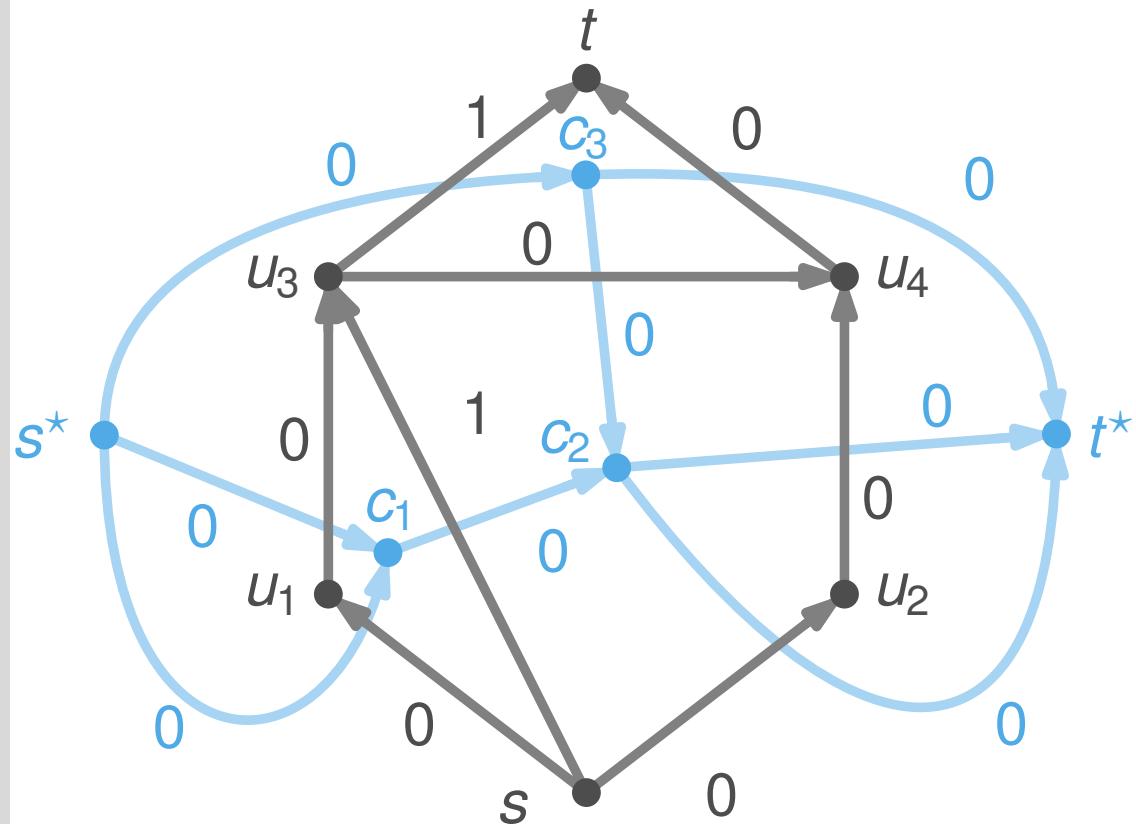
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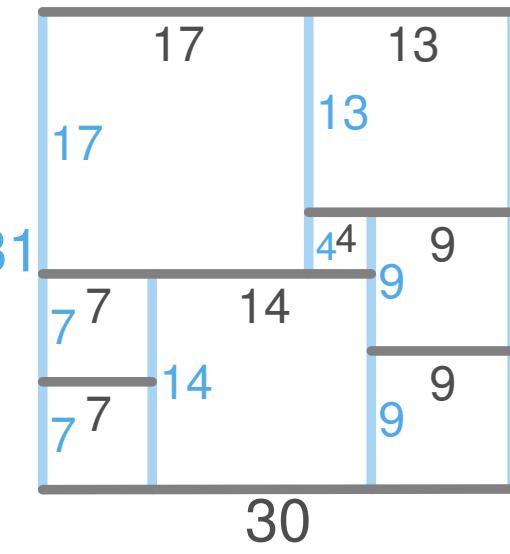
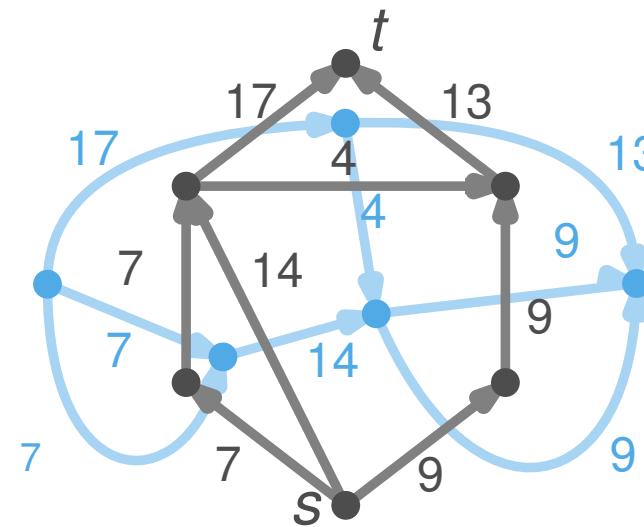
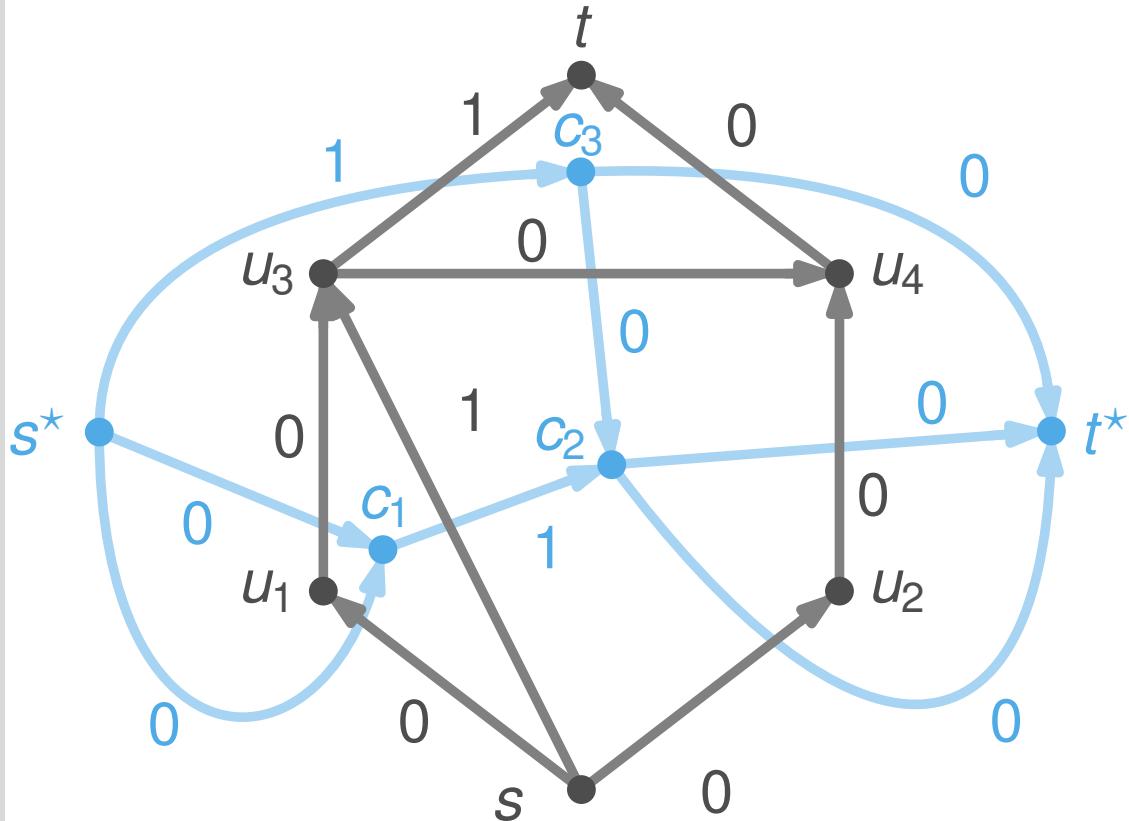
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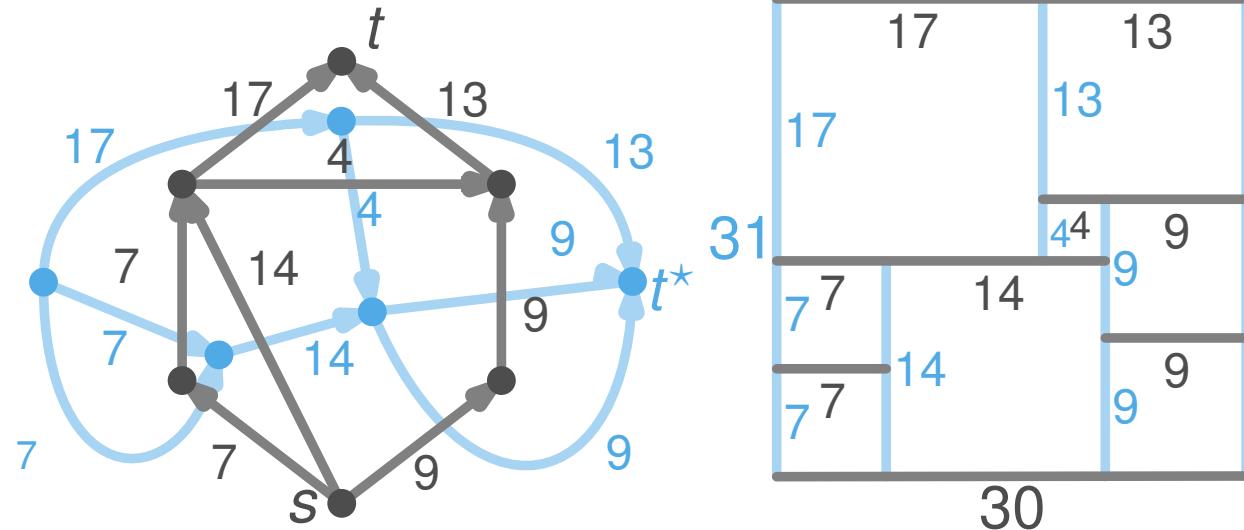
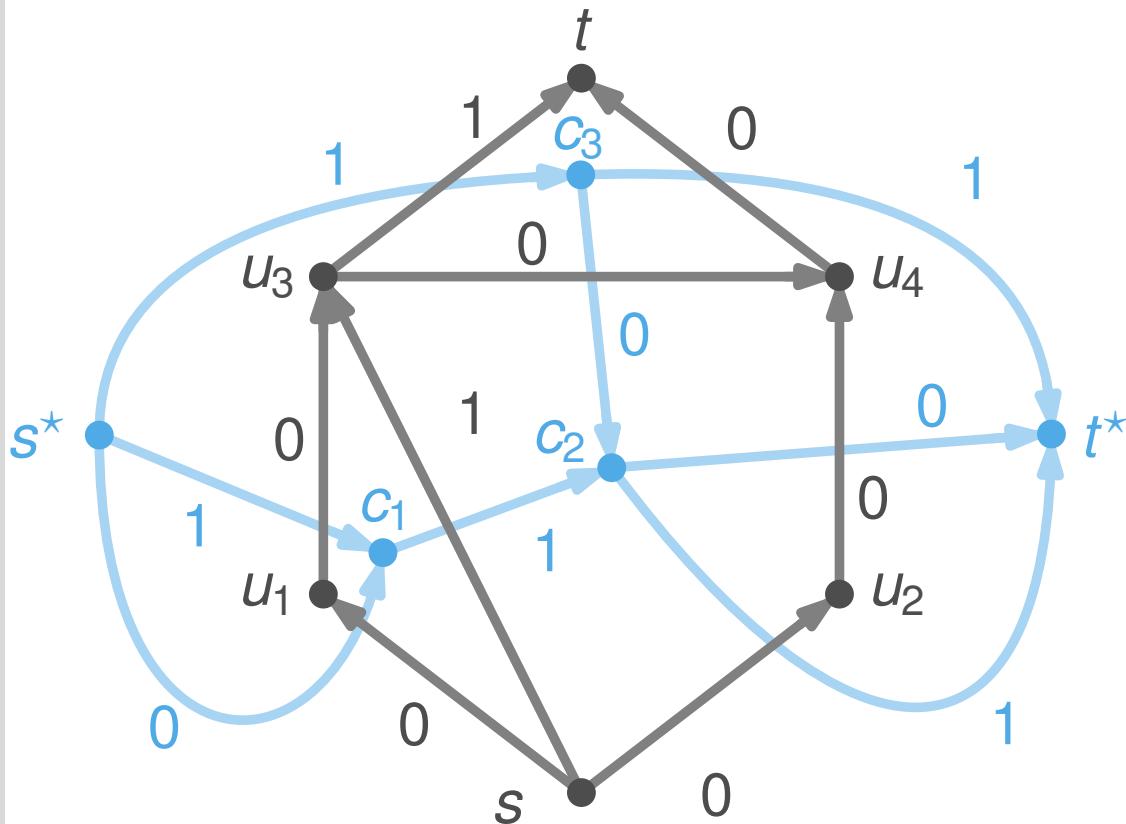
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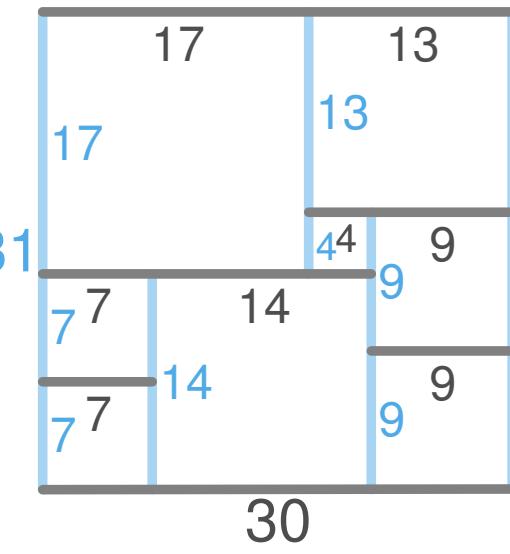
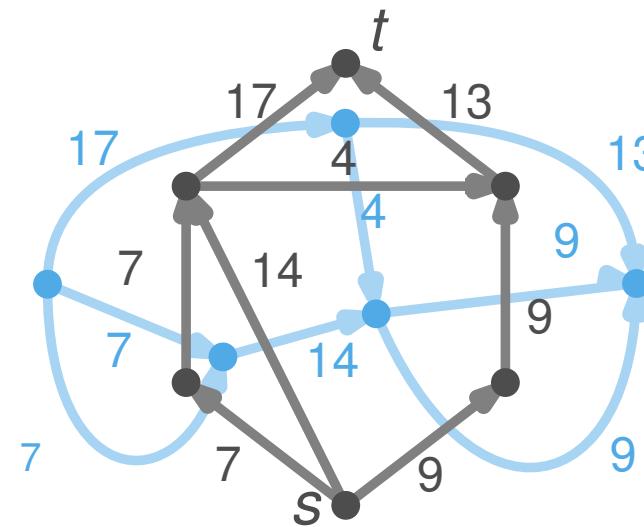
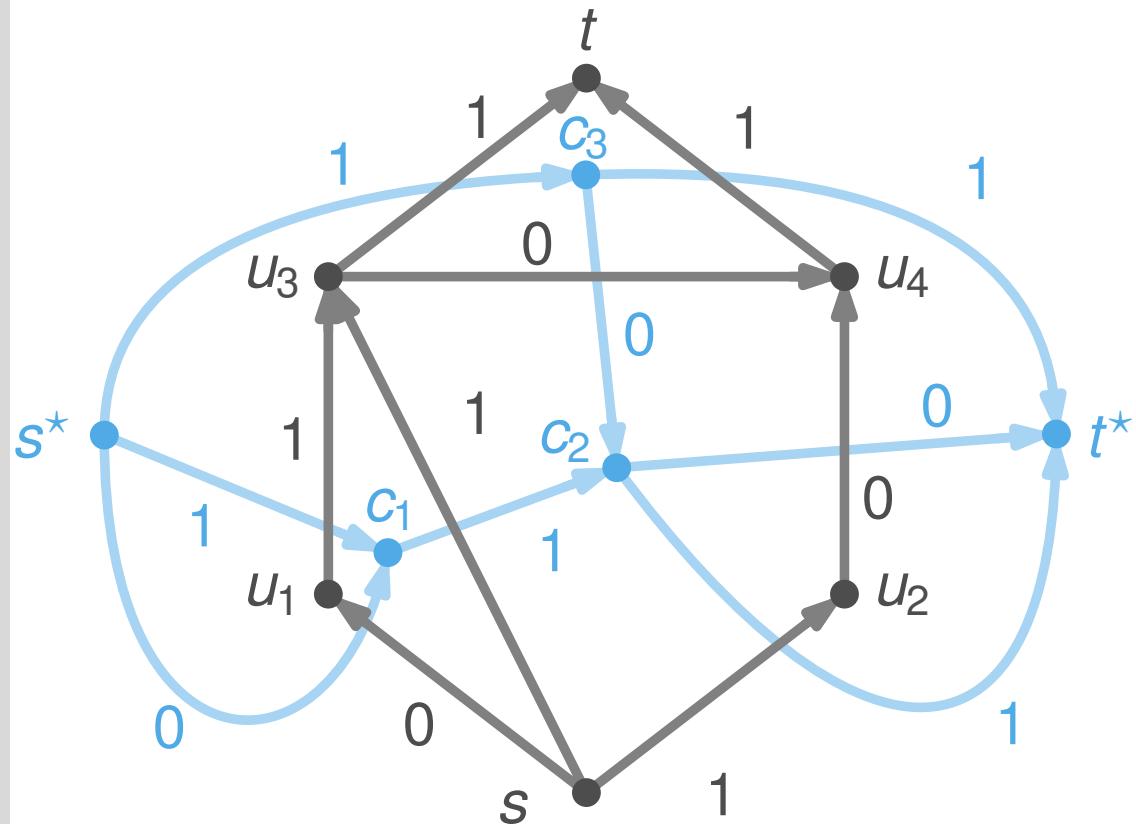
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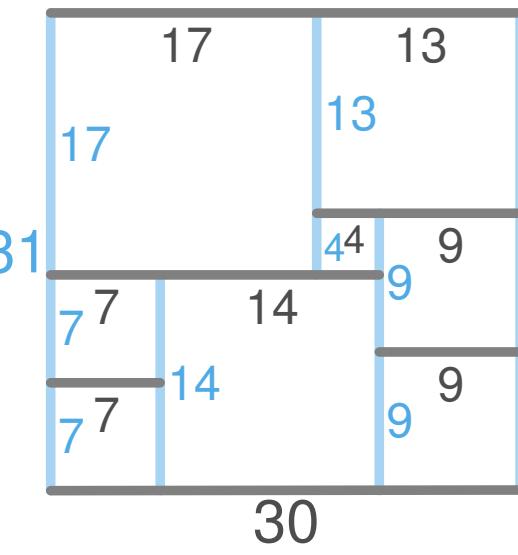
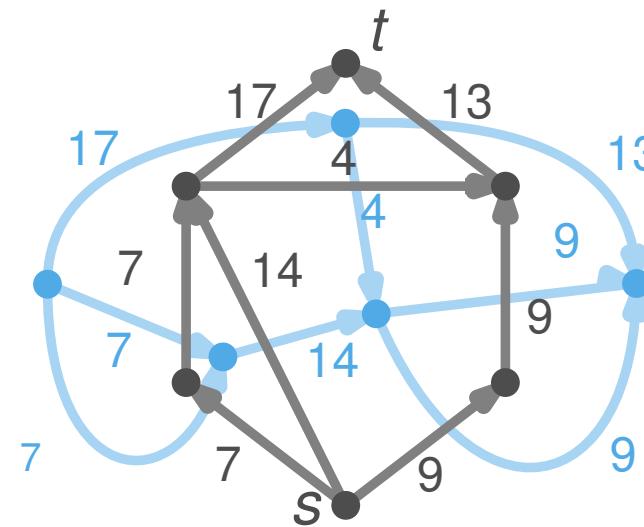
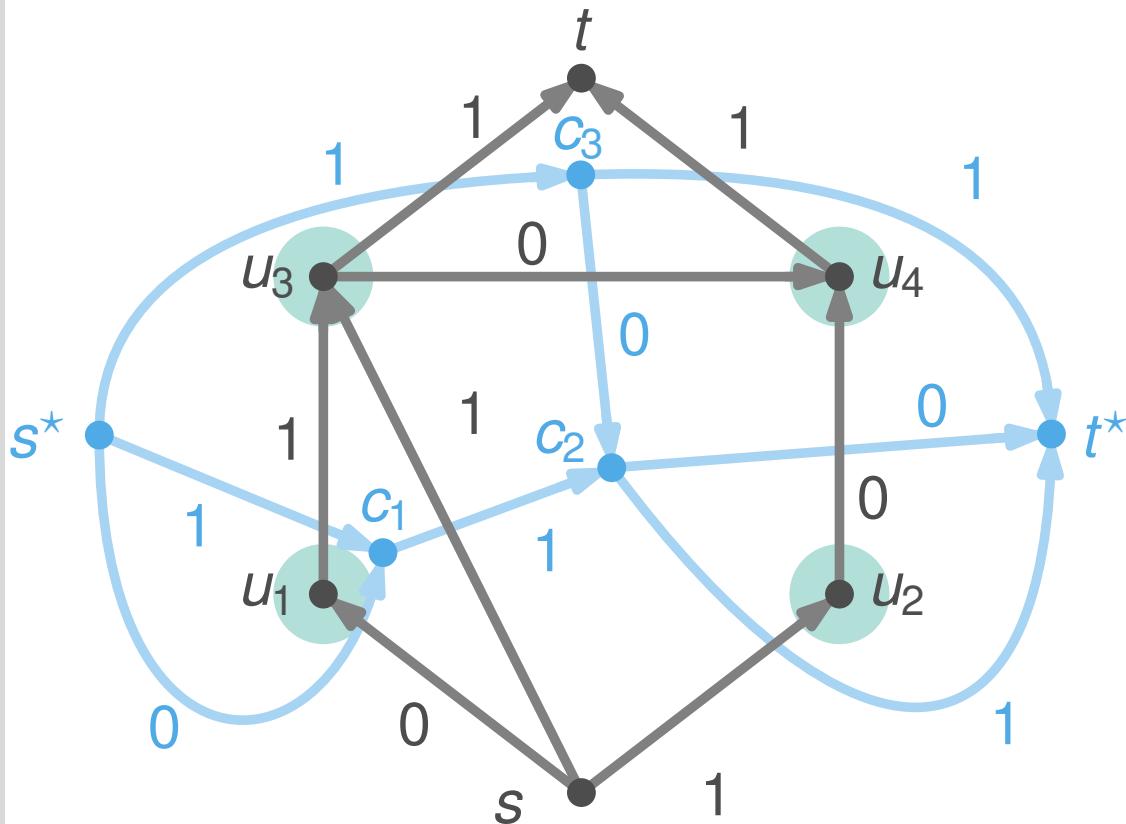
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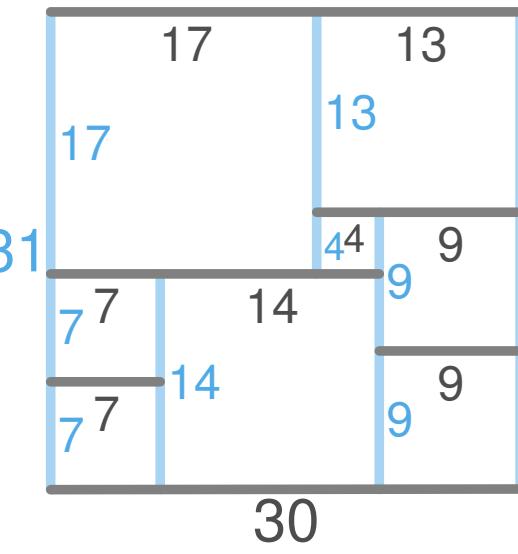
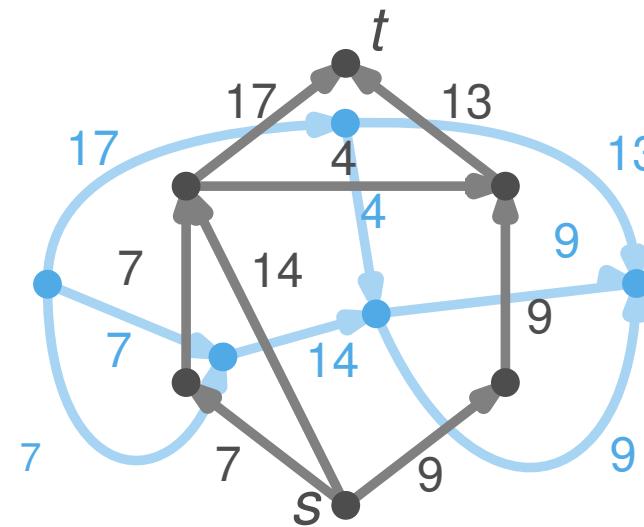
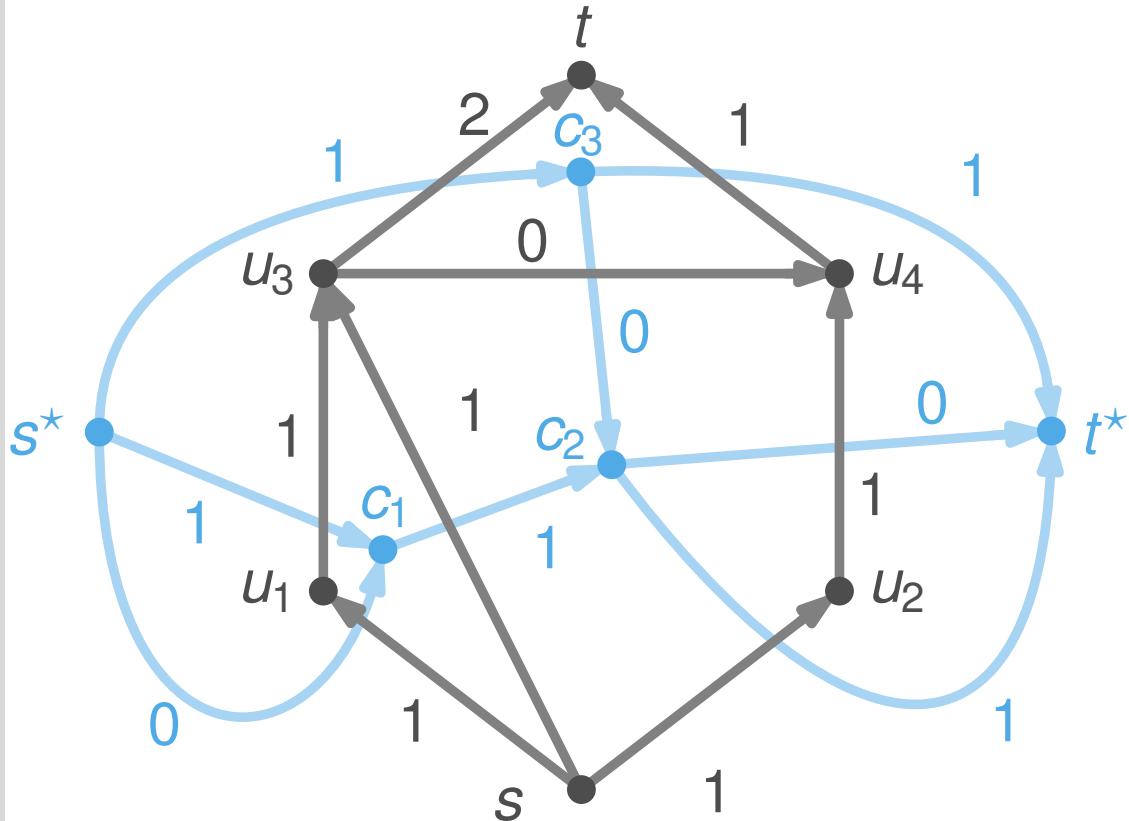
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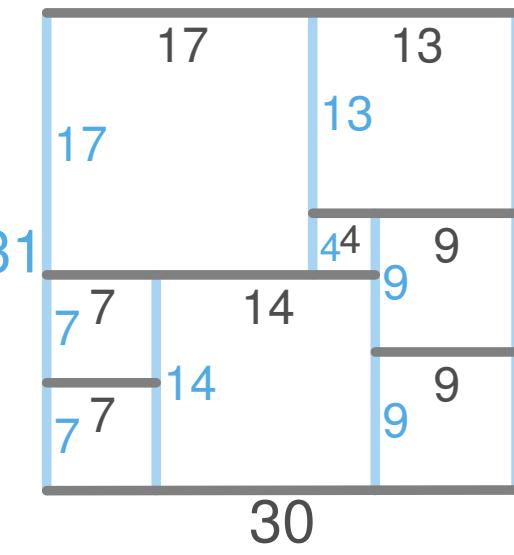
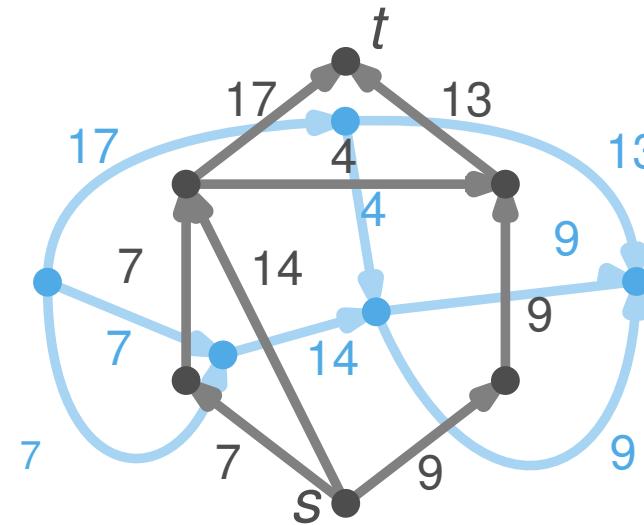
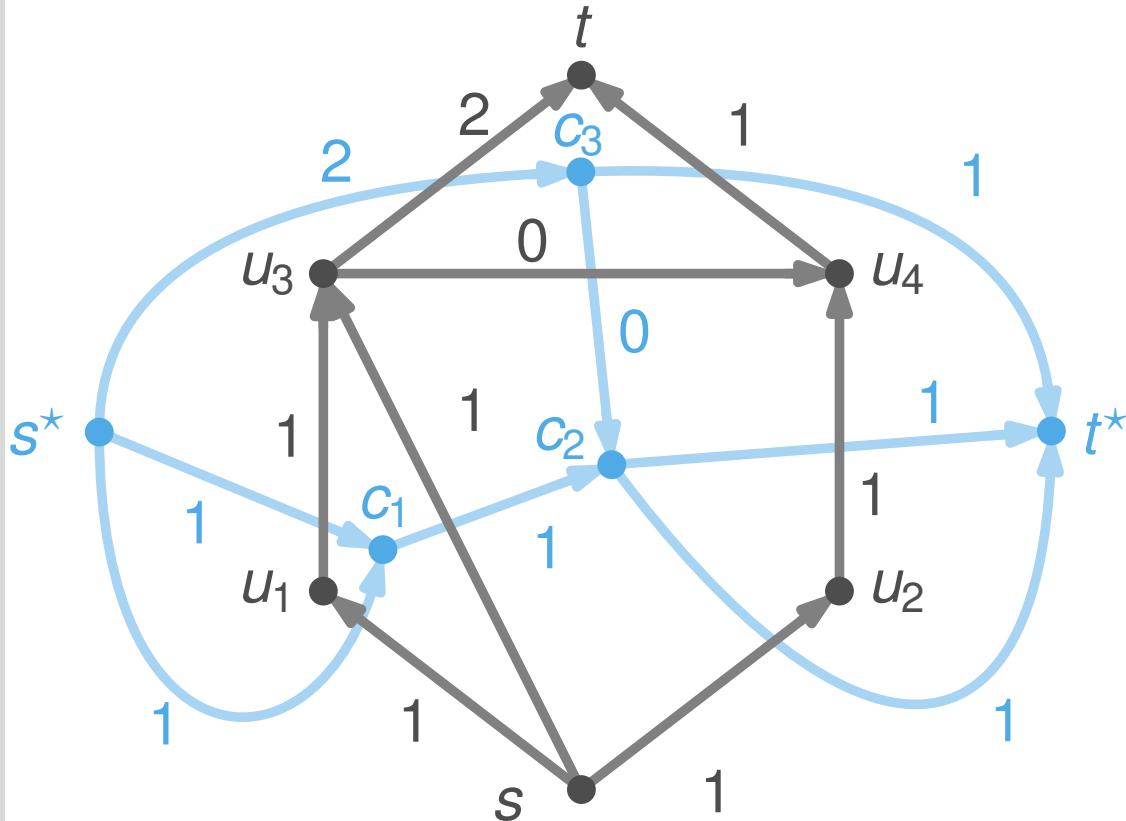
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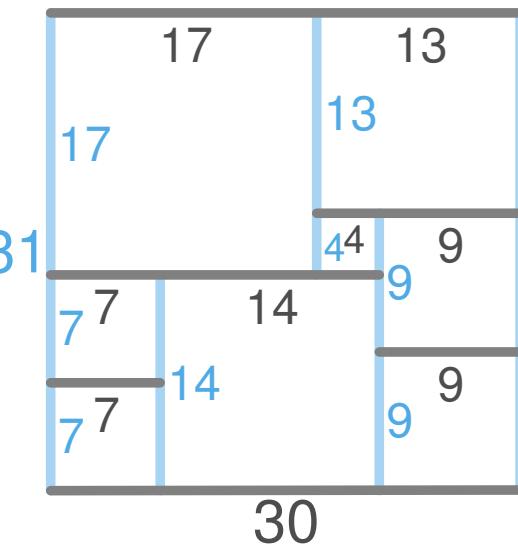
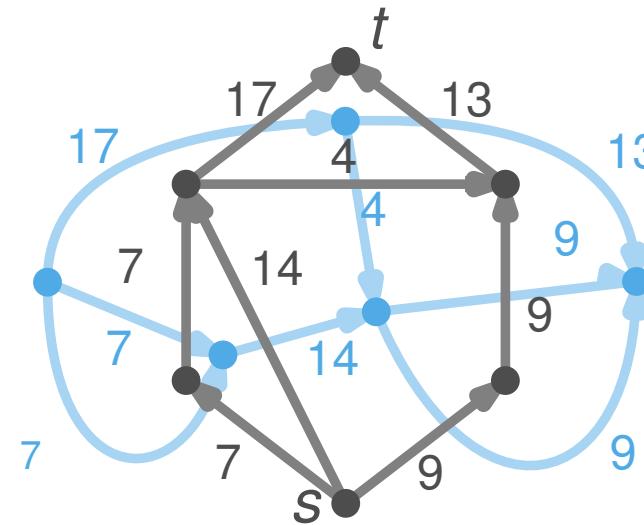
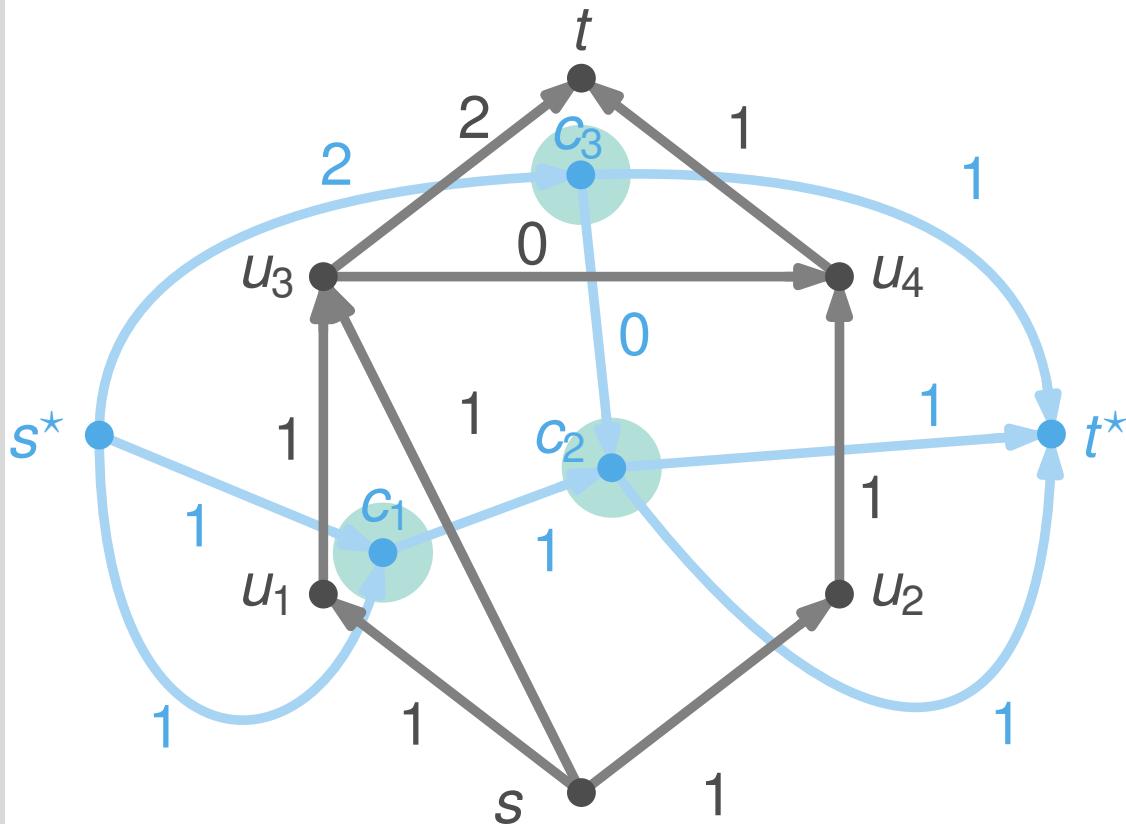
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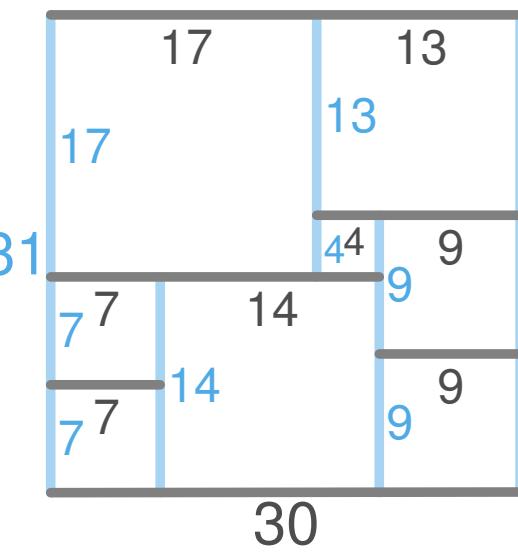
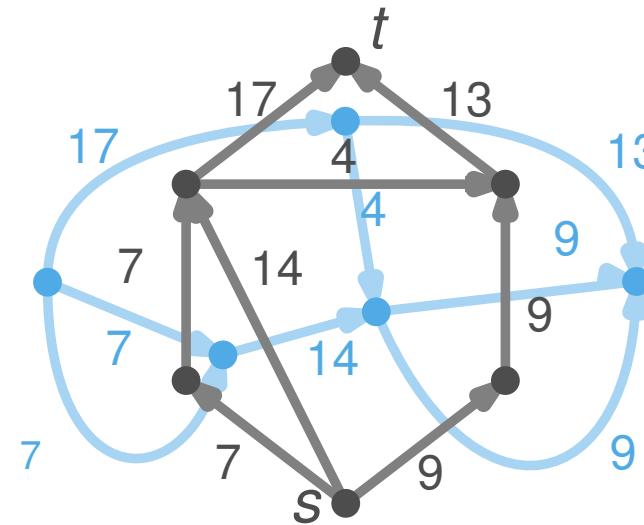
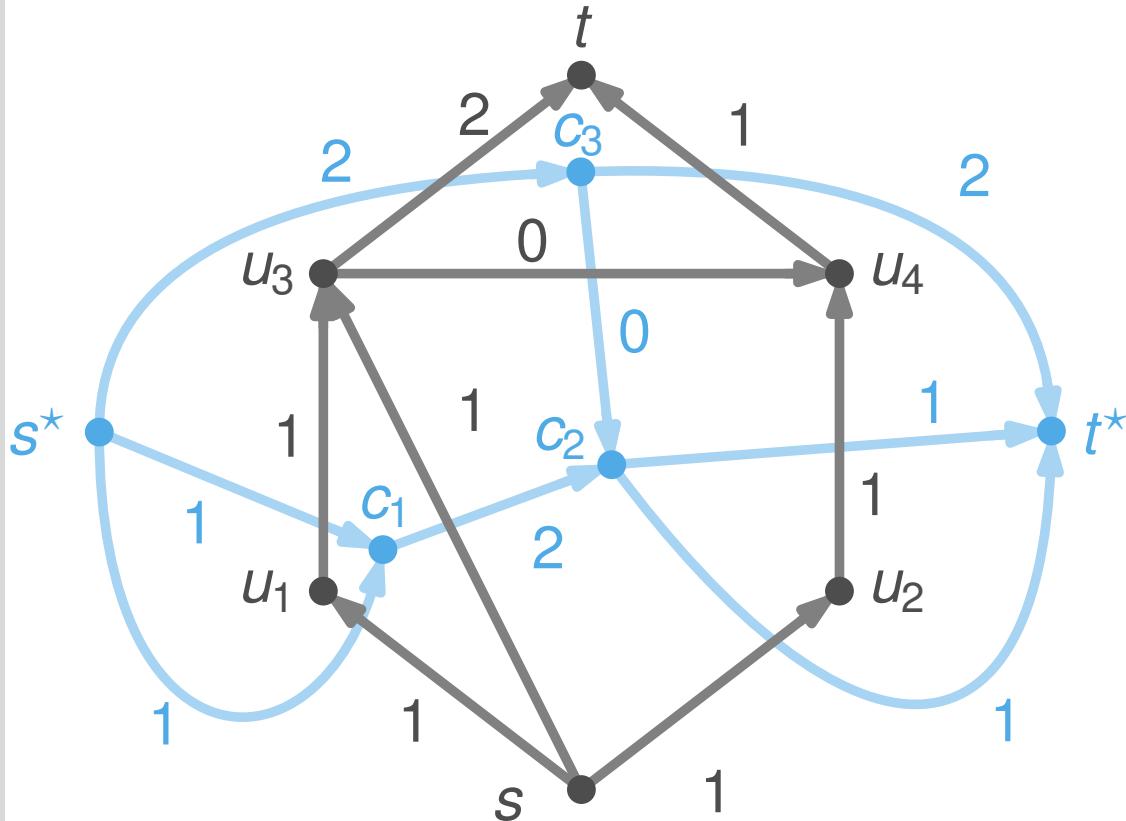
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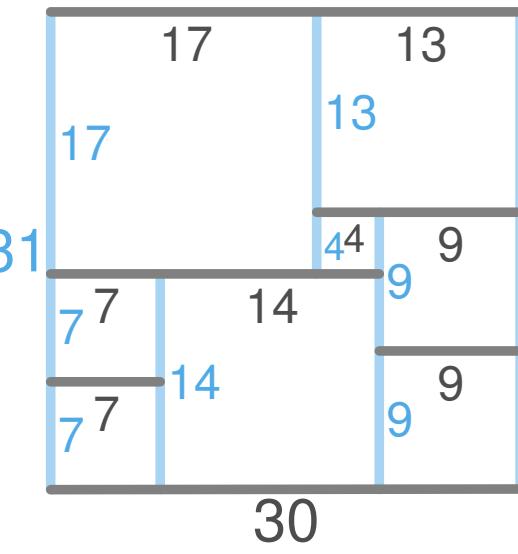
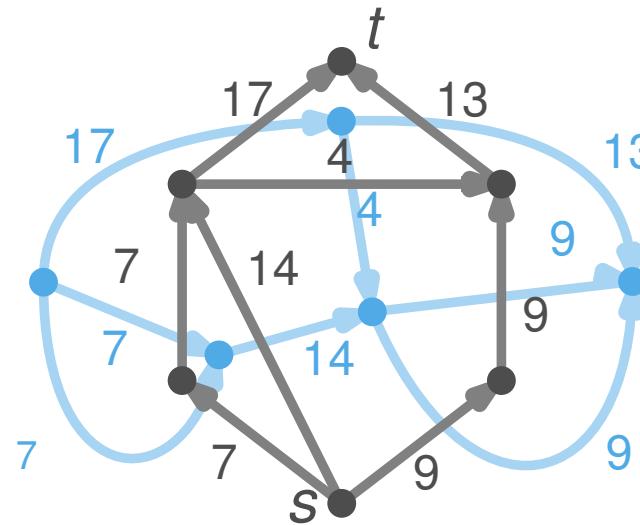
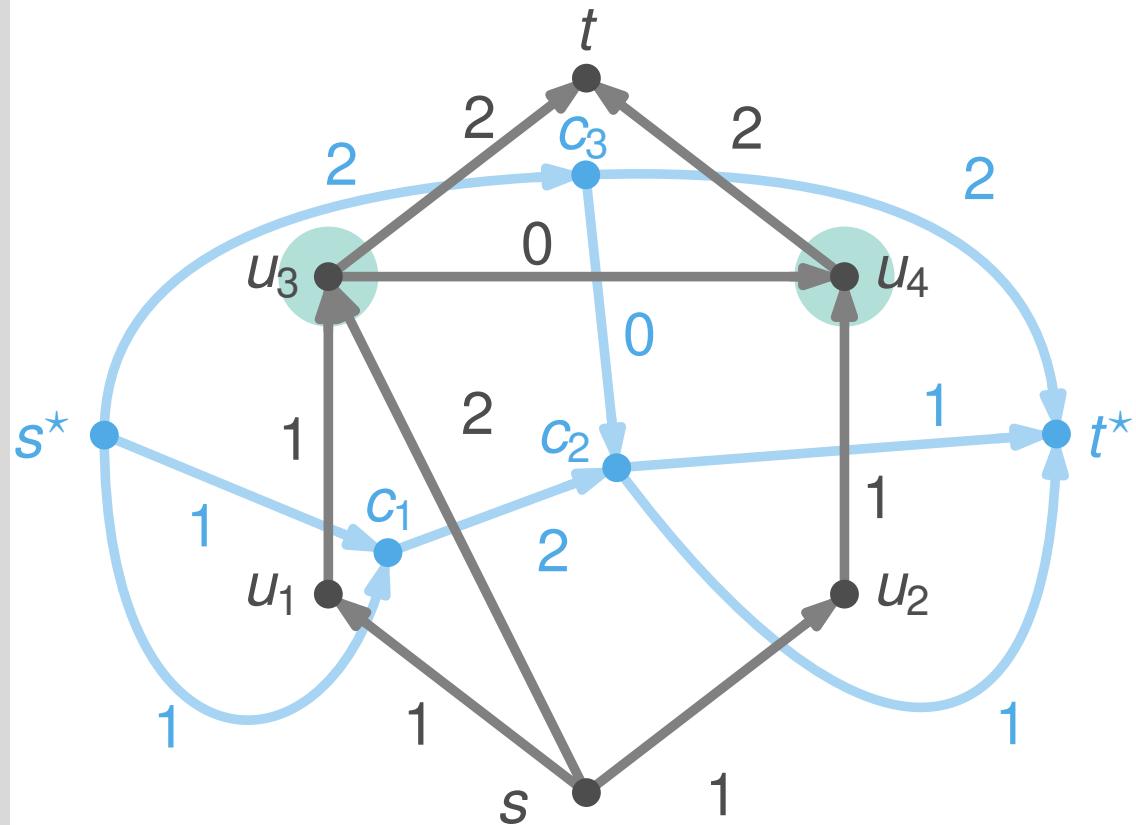
# Wrong Conflict Resolution



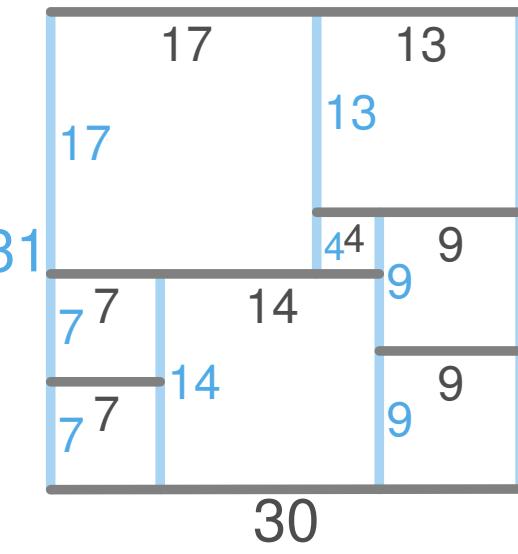
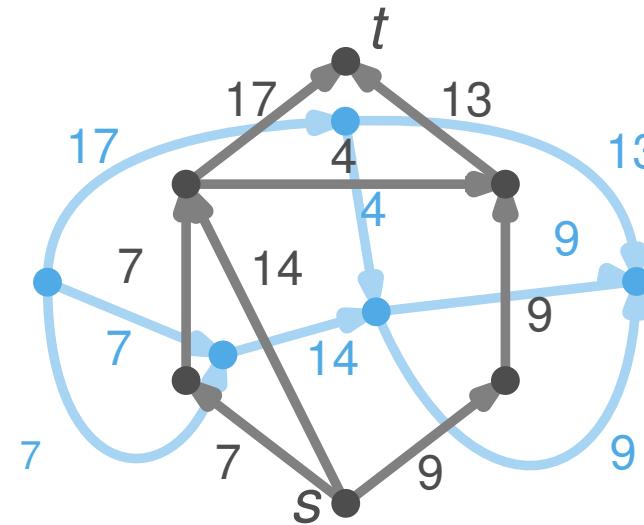
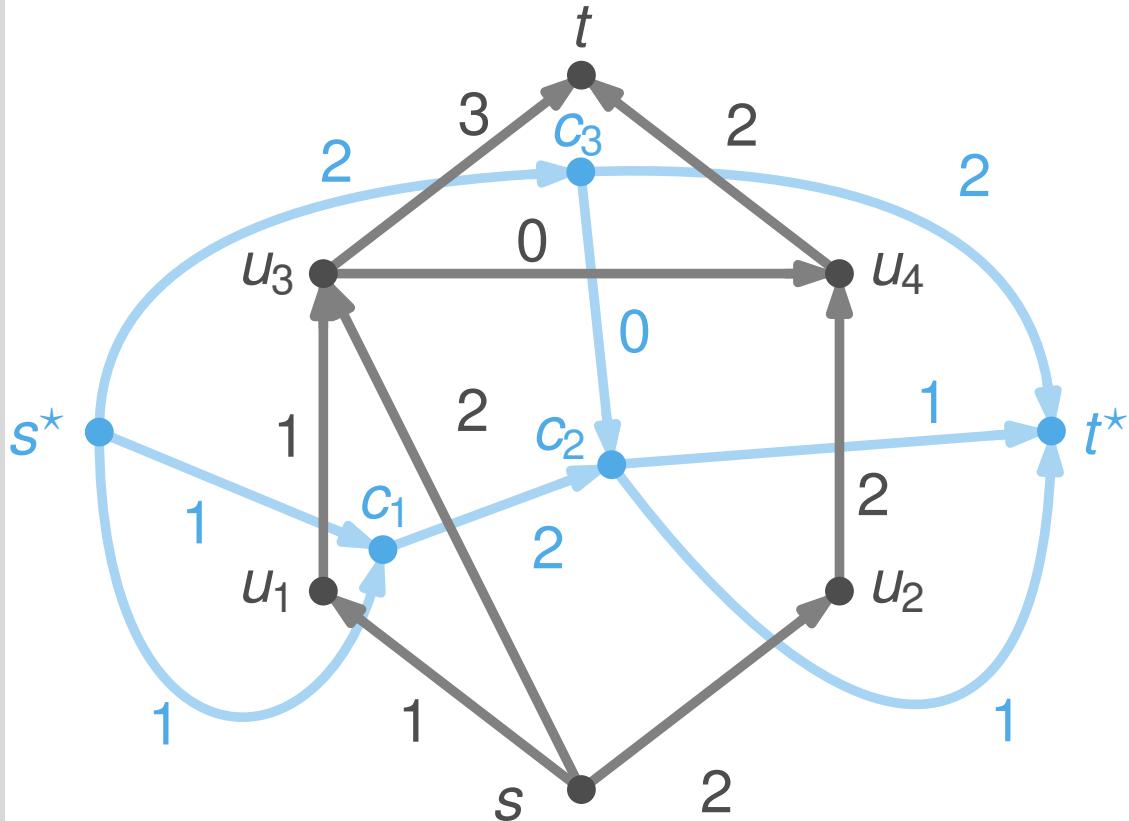
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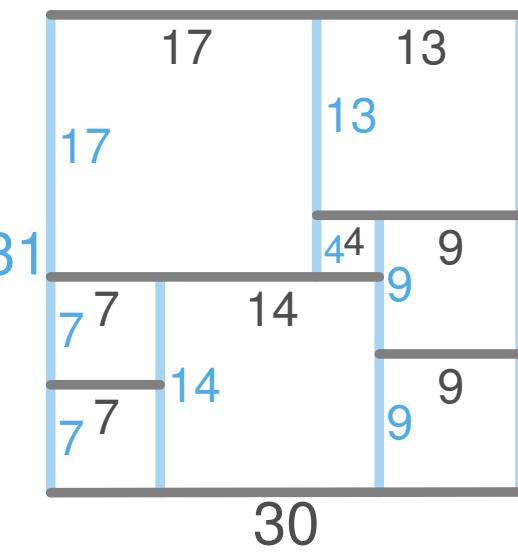
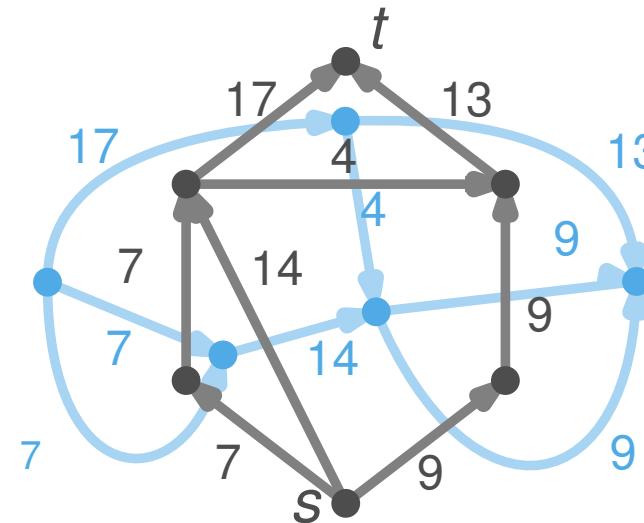
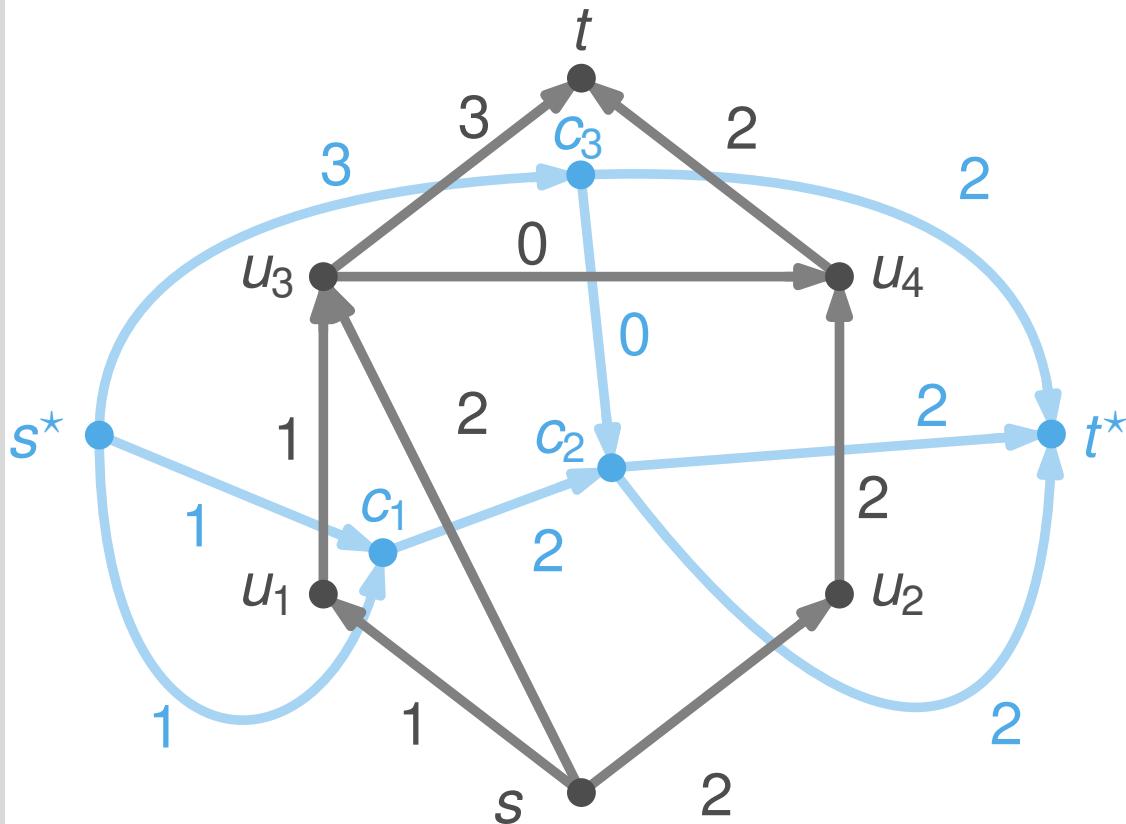
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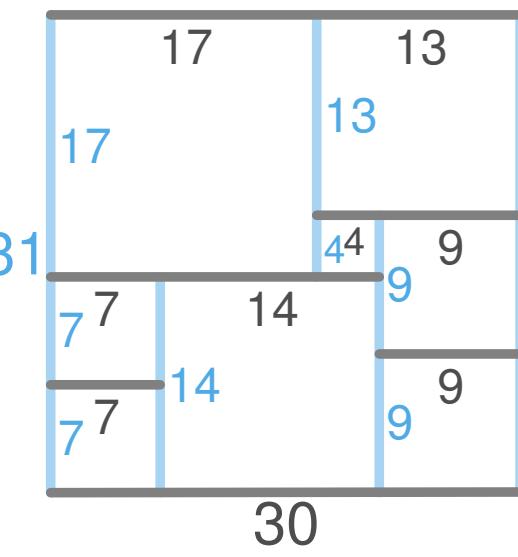
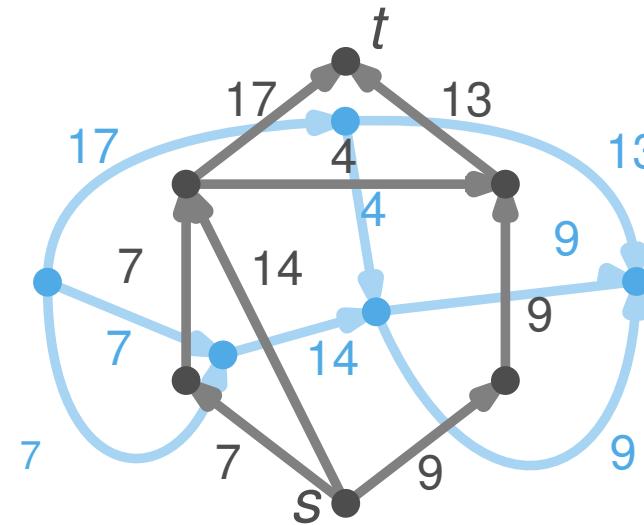
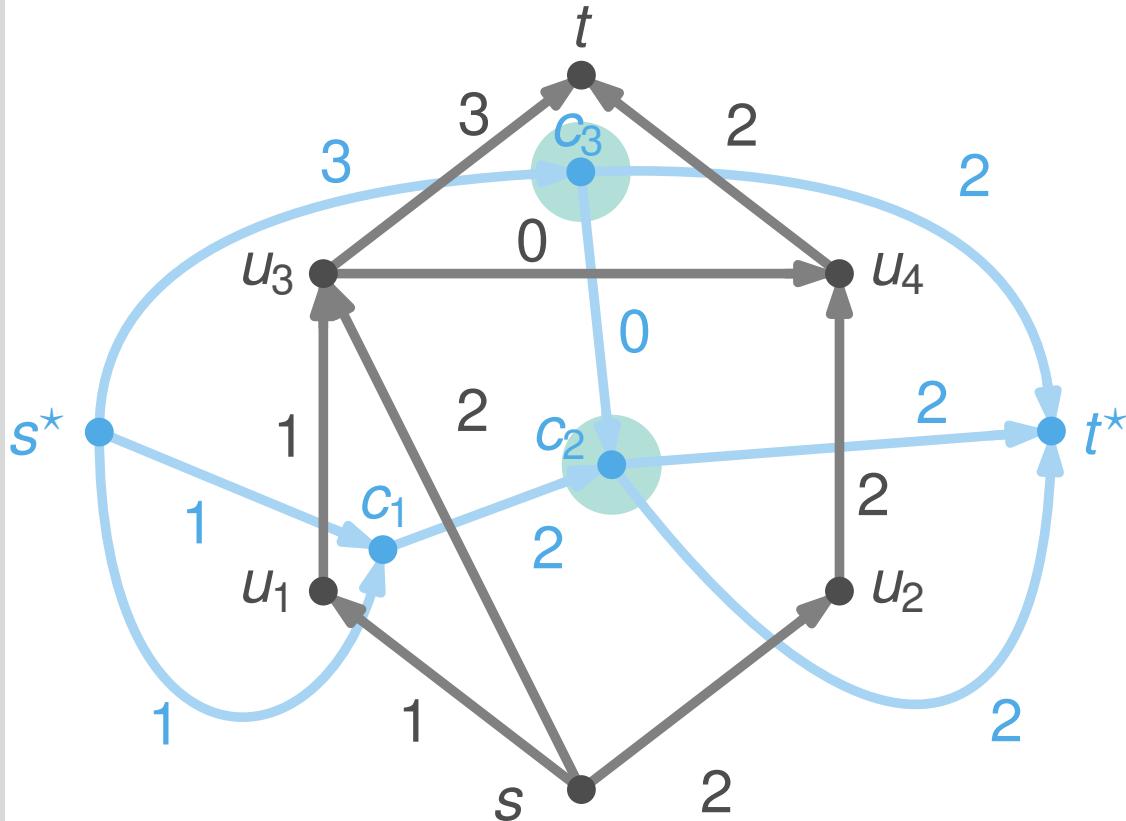
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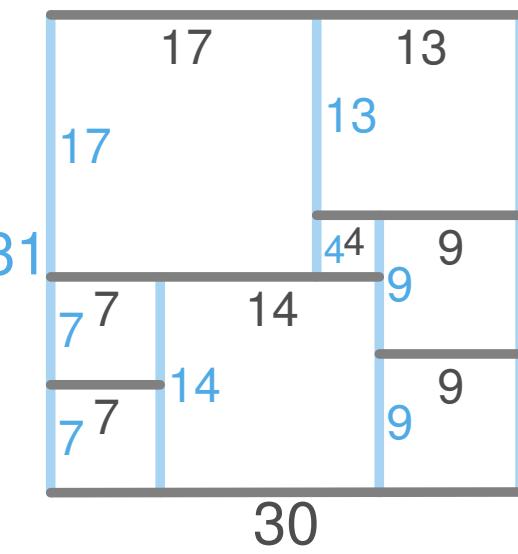
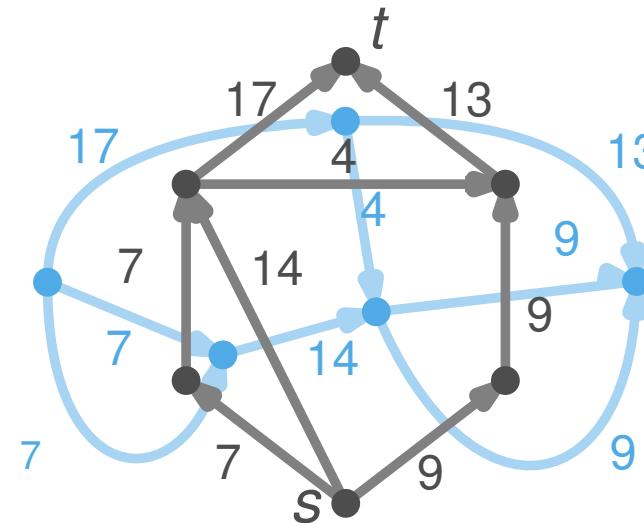
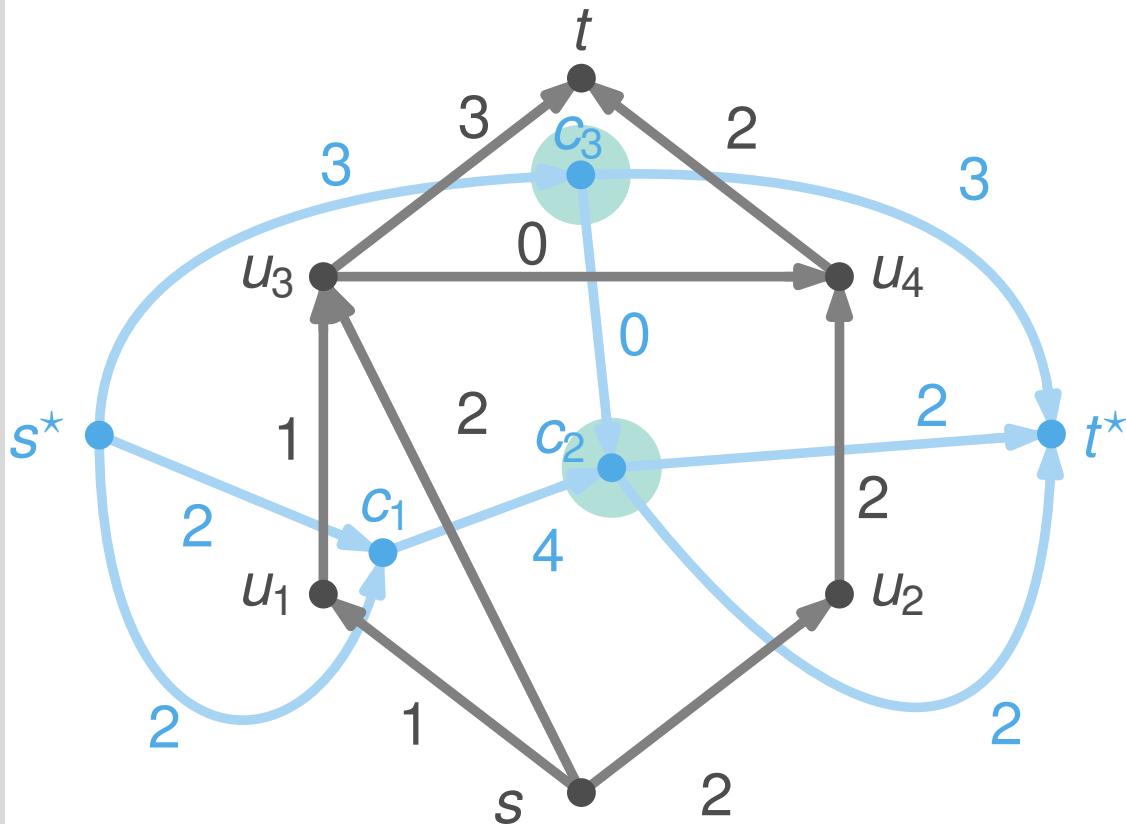
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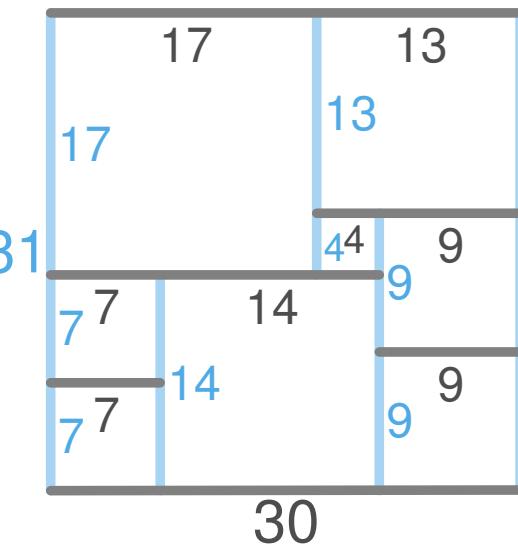
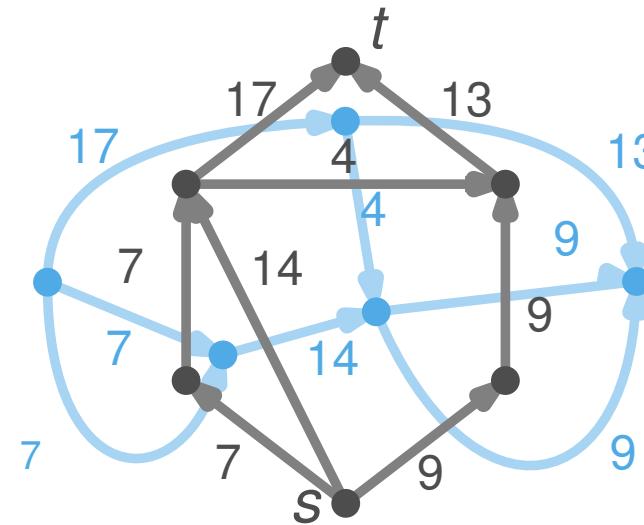
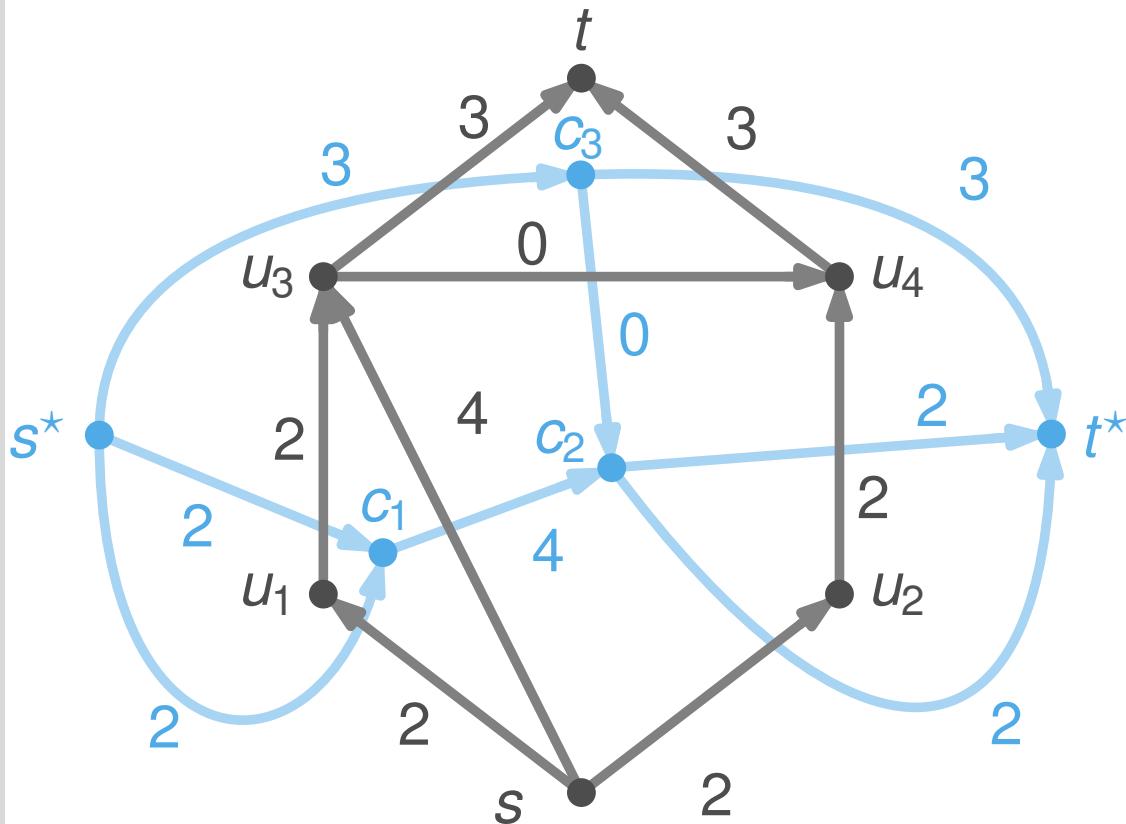
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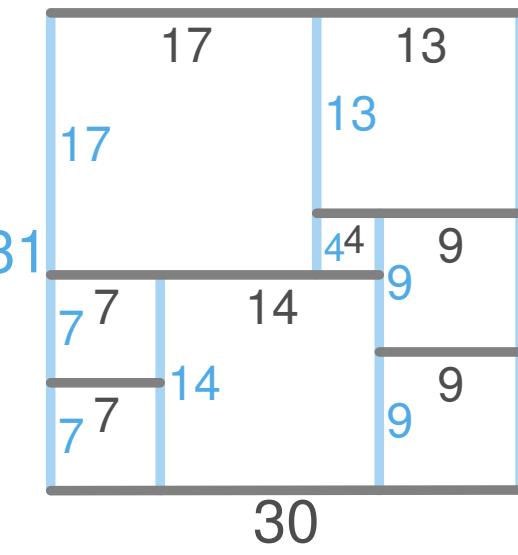
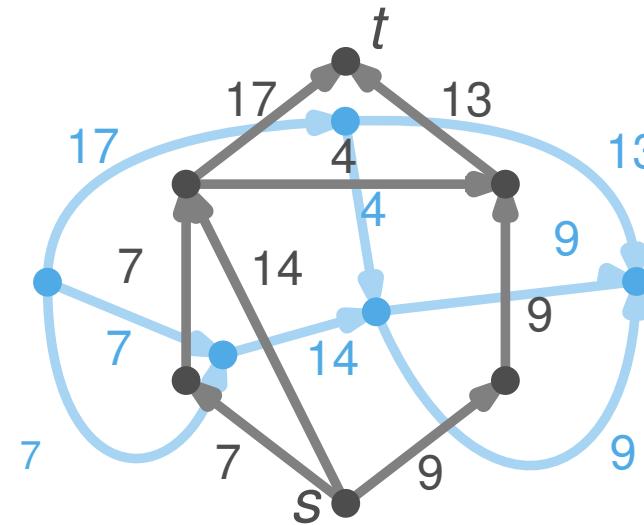
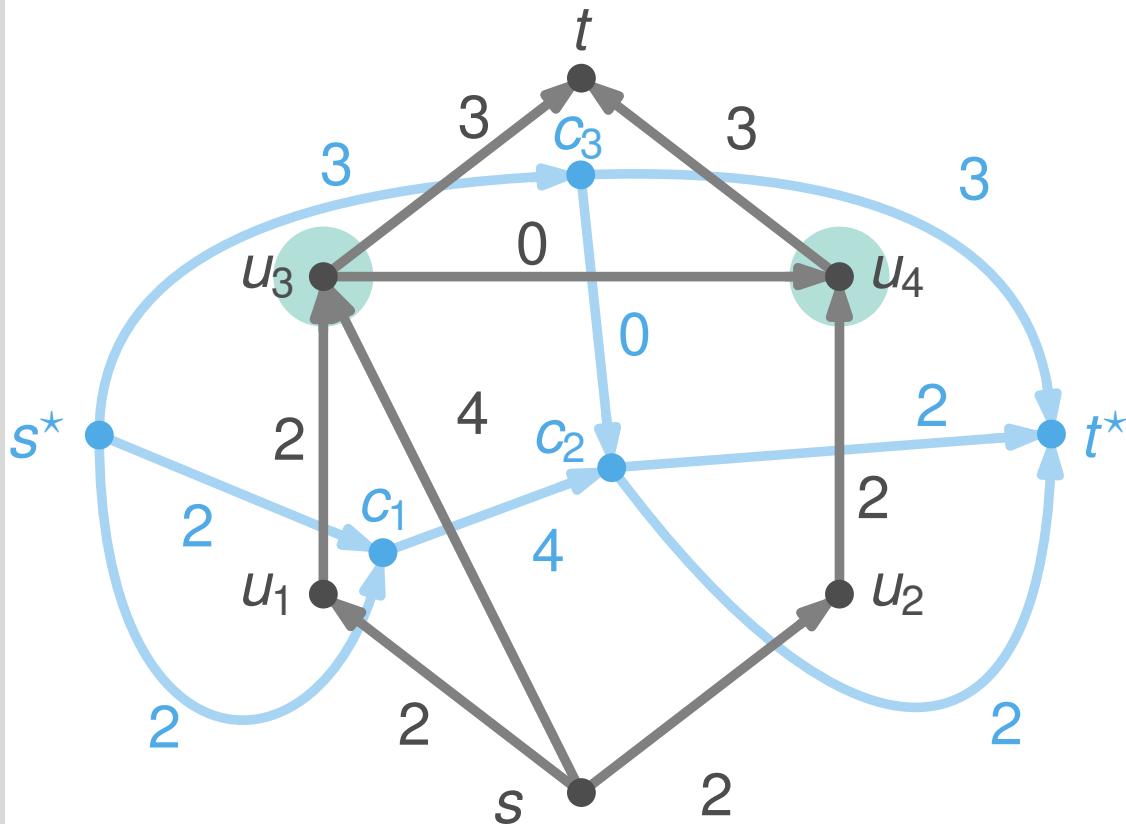
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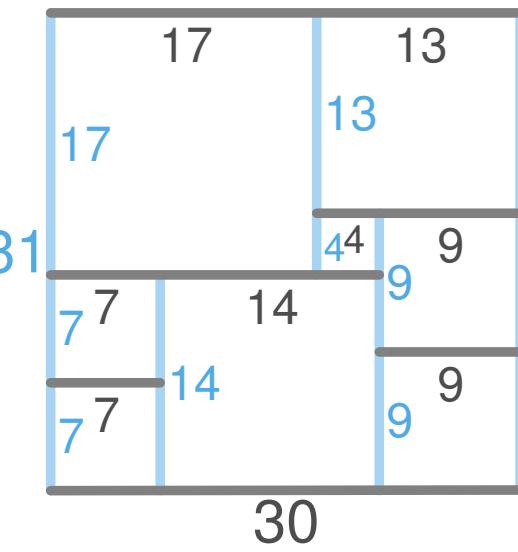
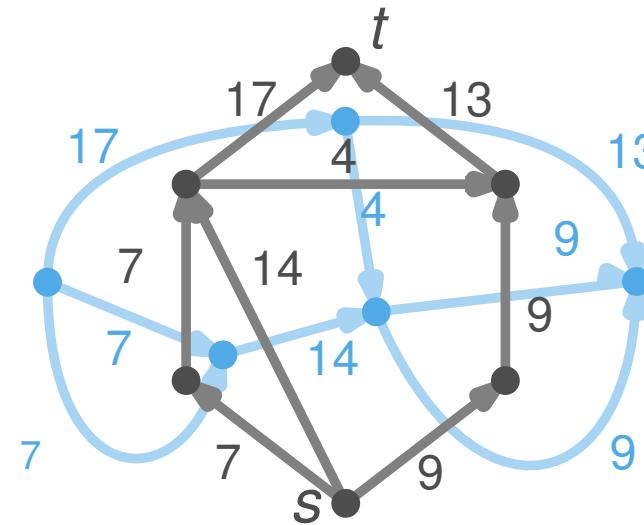
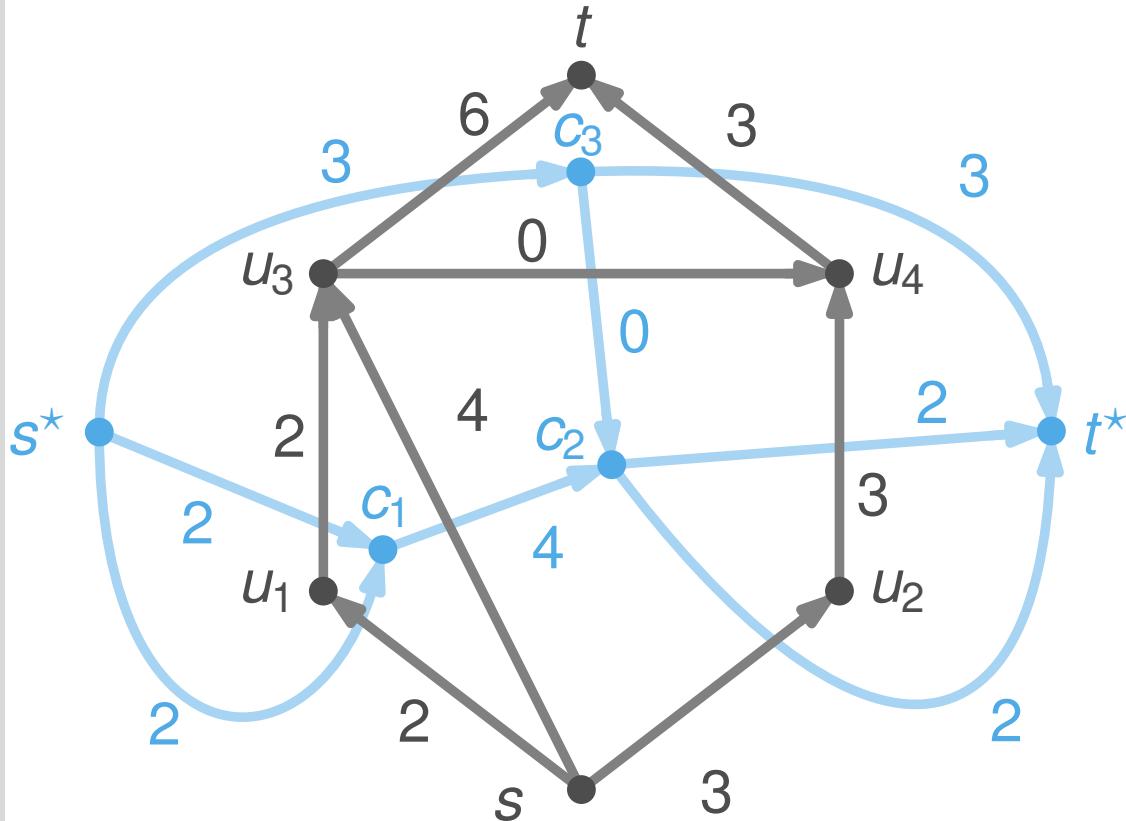
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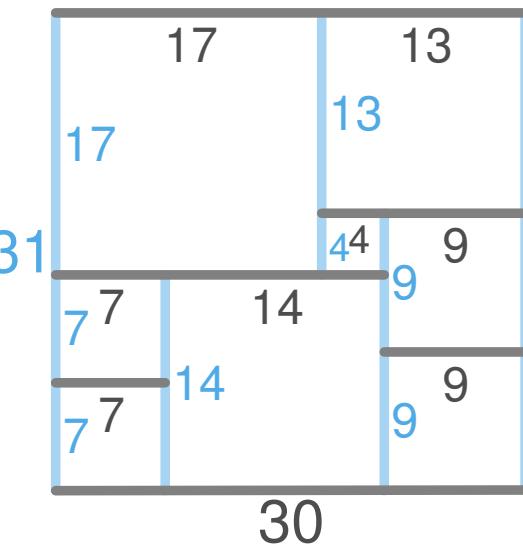
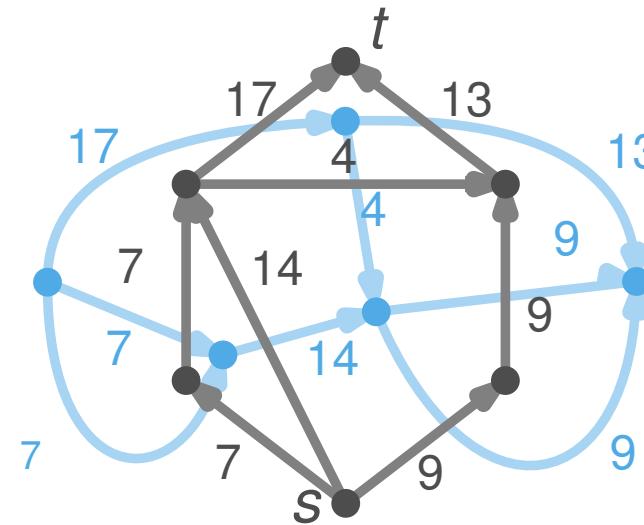
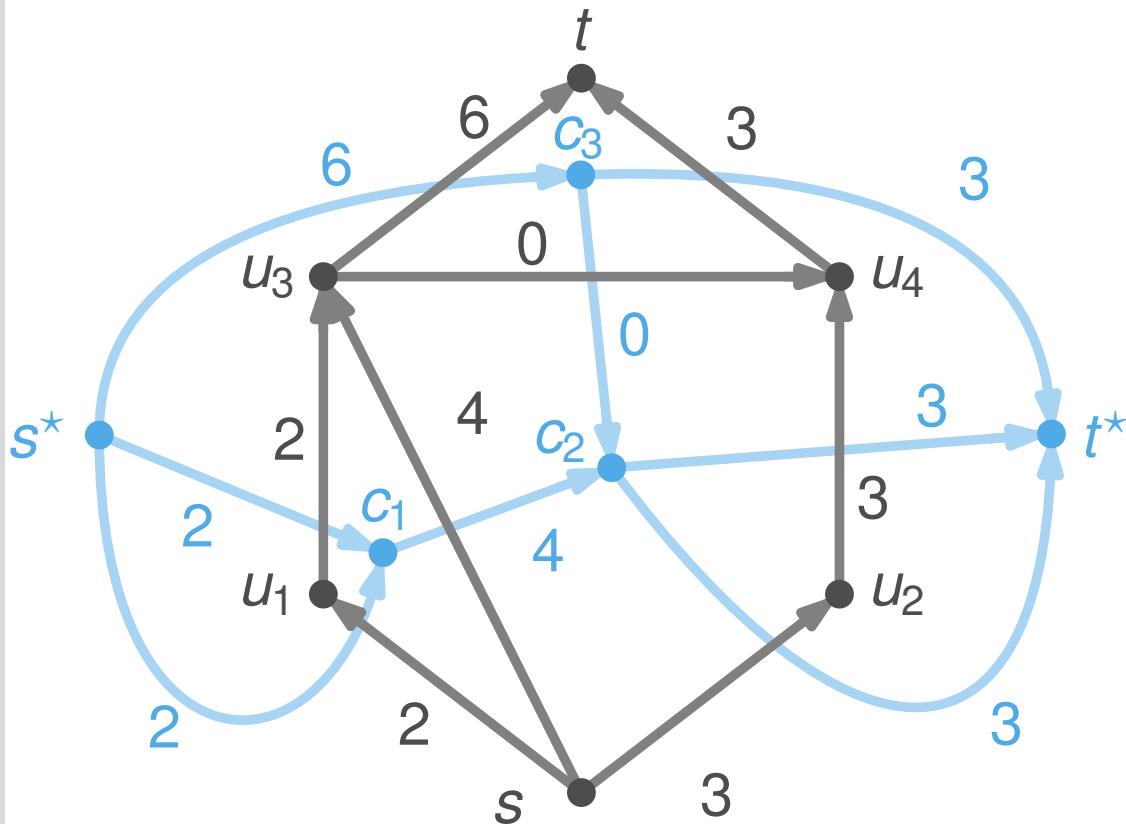
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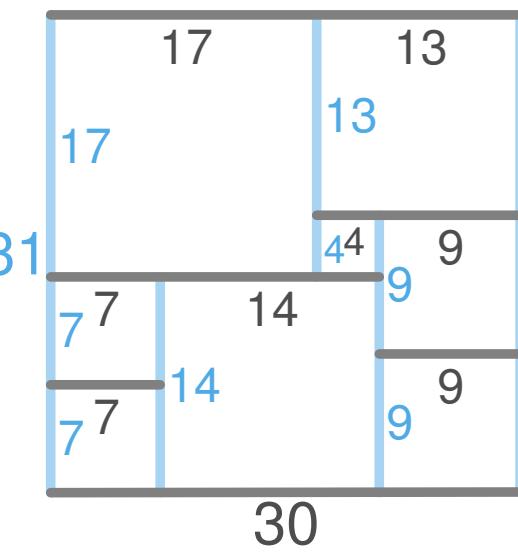
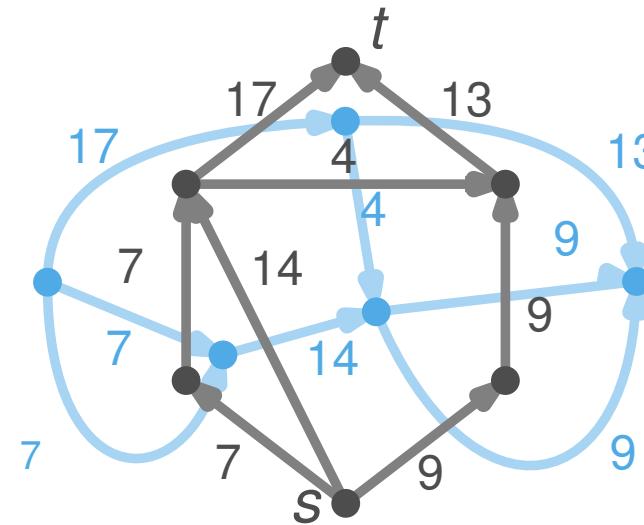
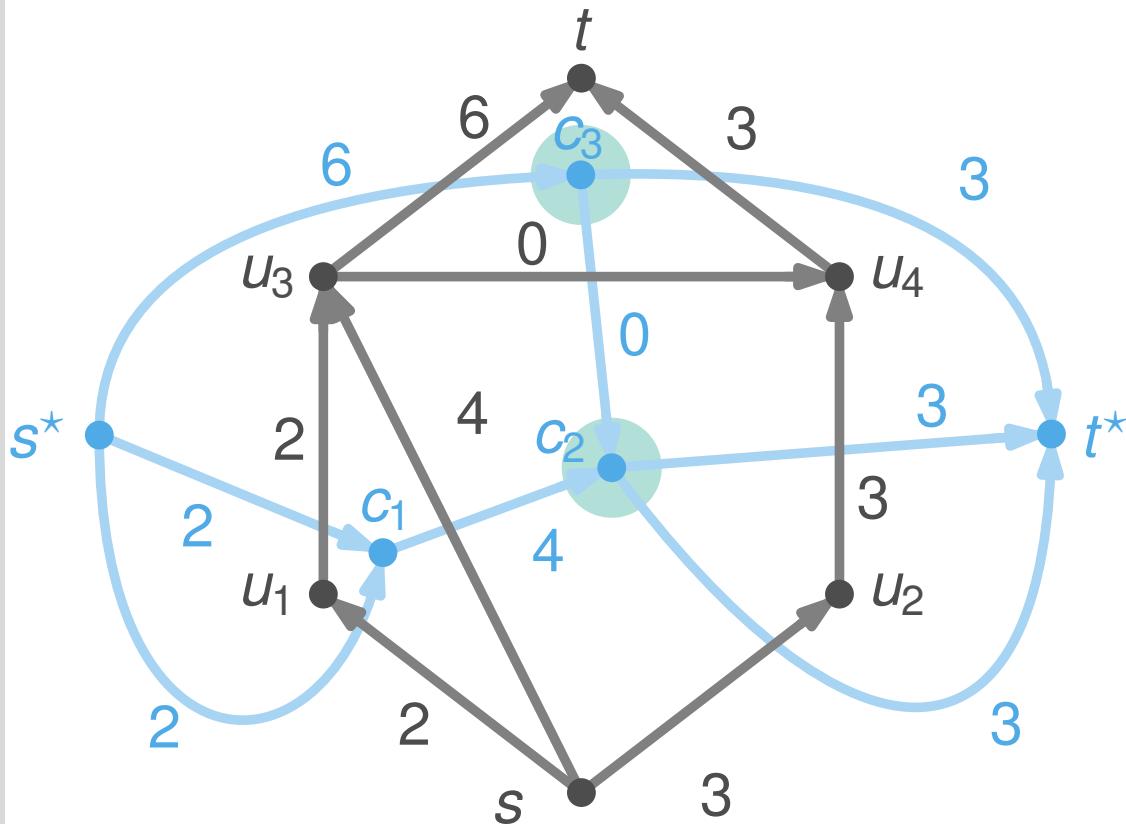
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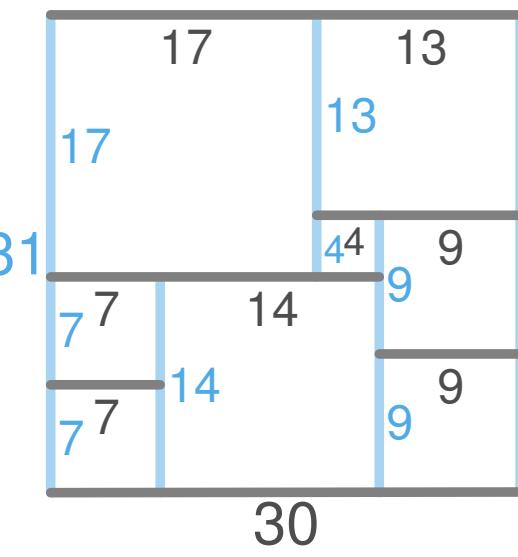
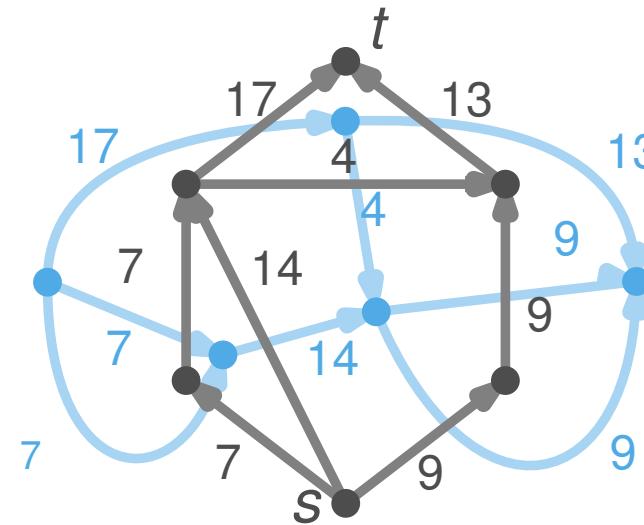
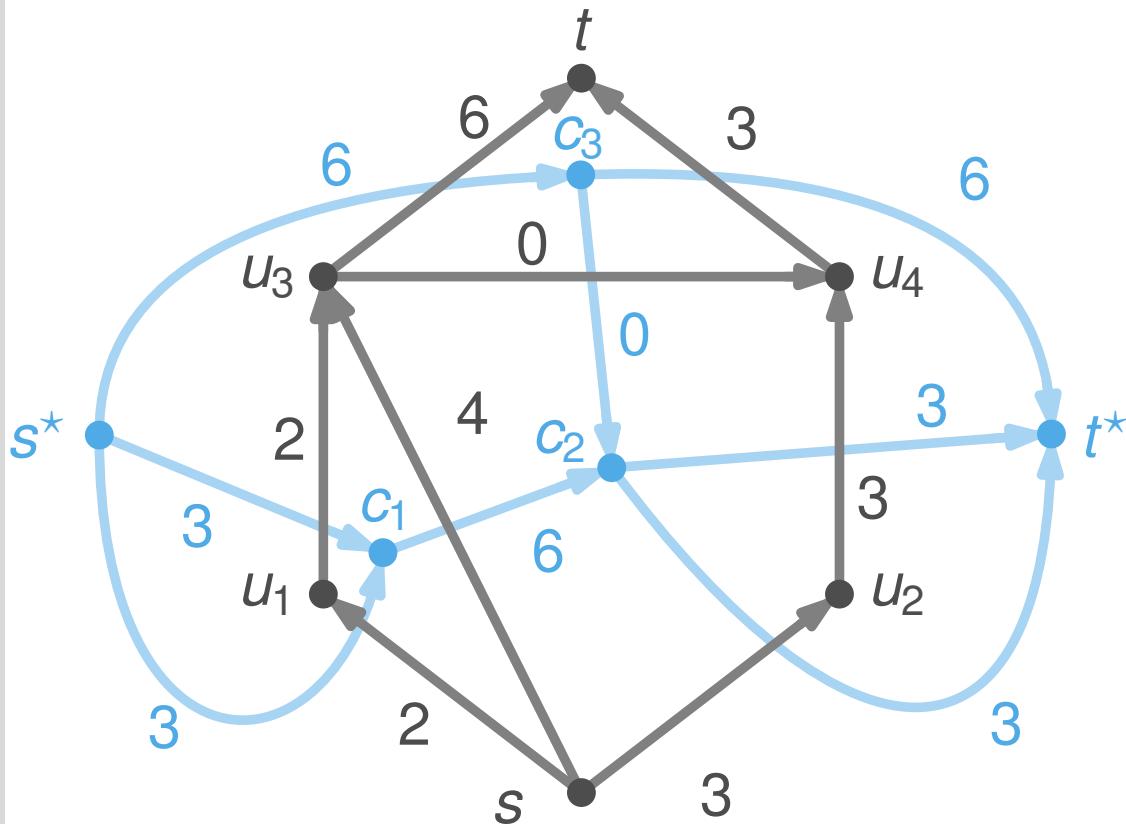
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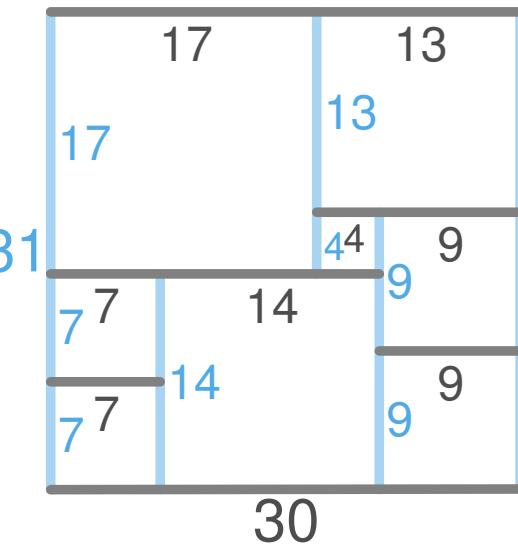
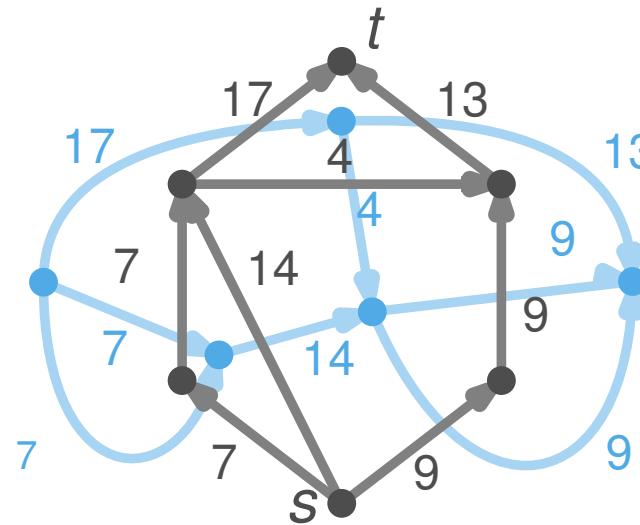
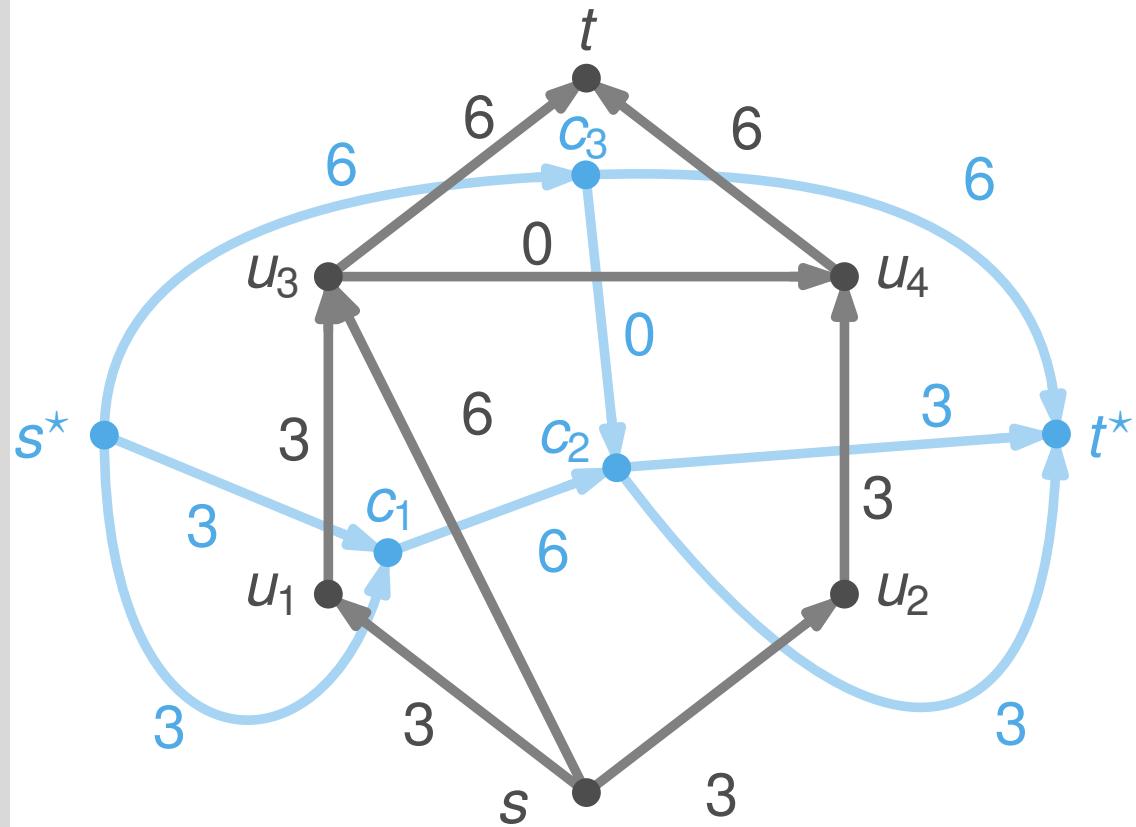
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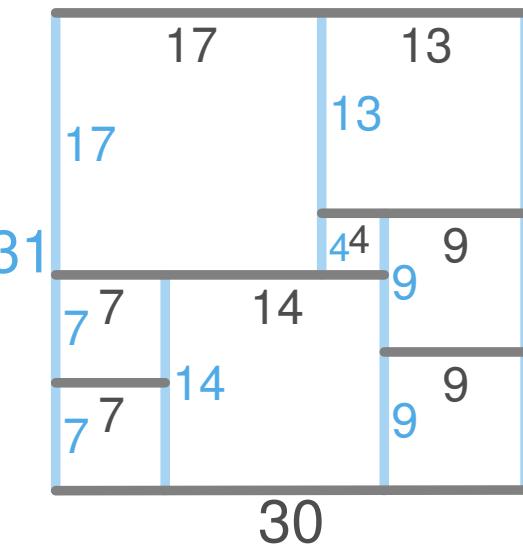
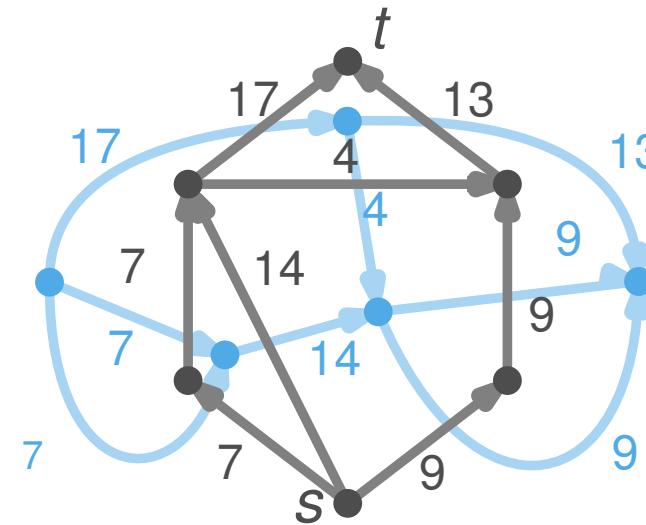
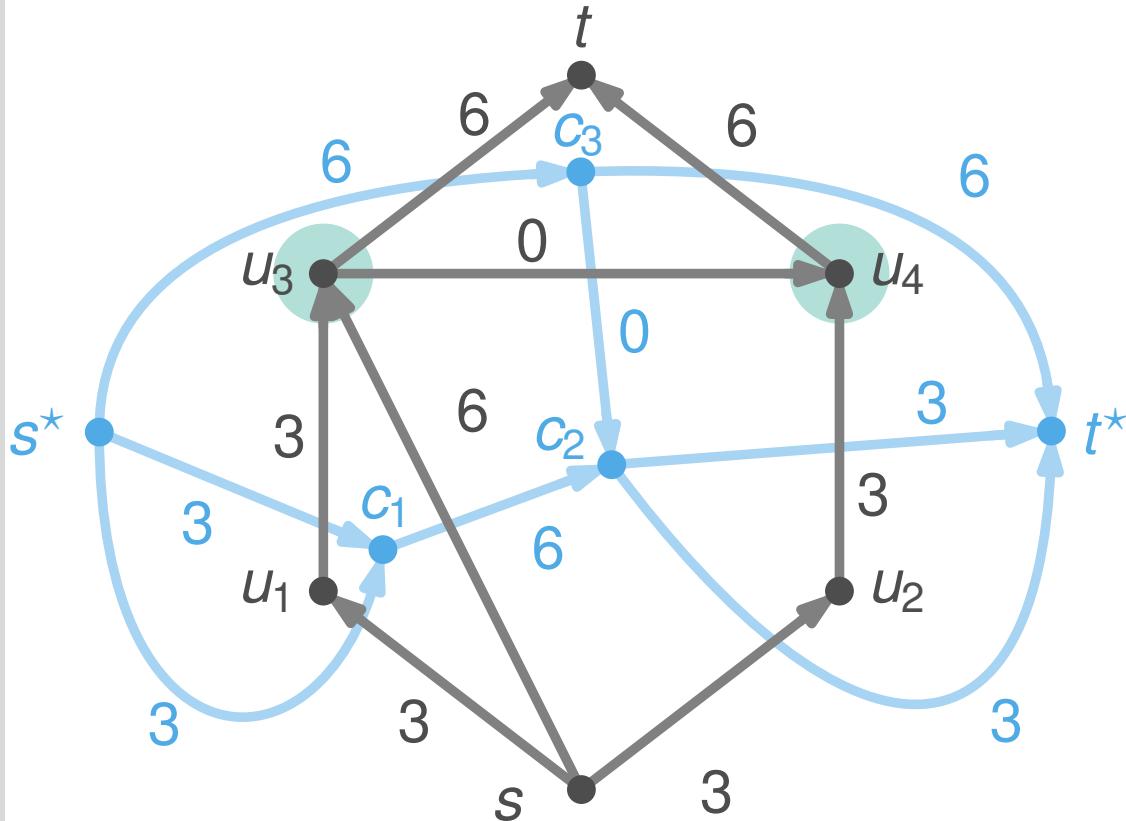
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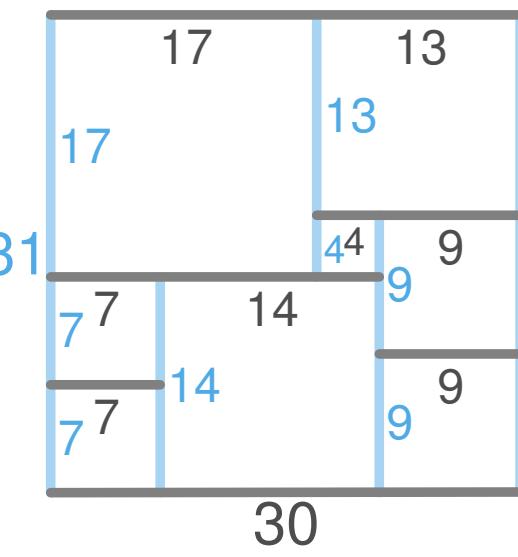
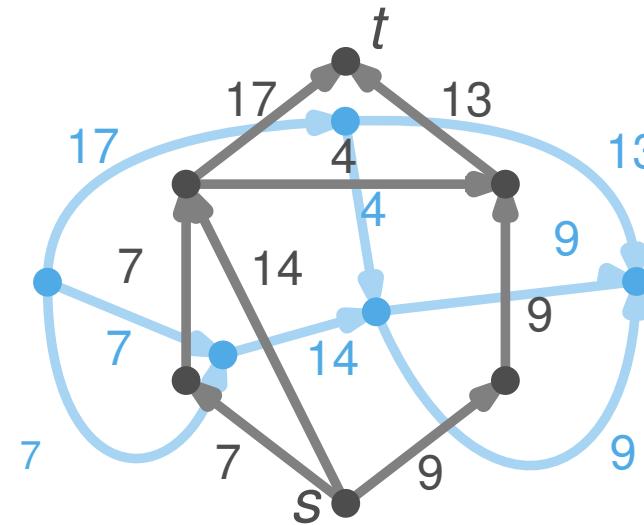
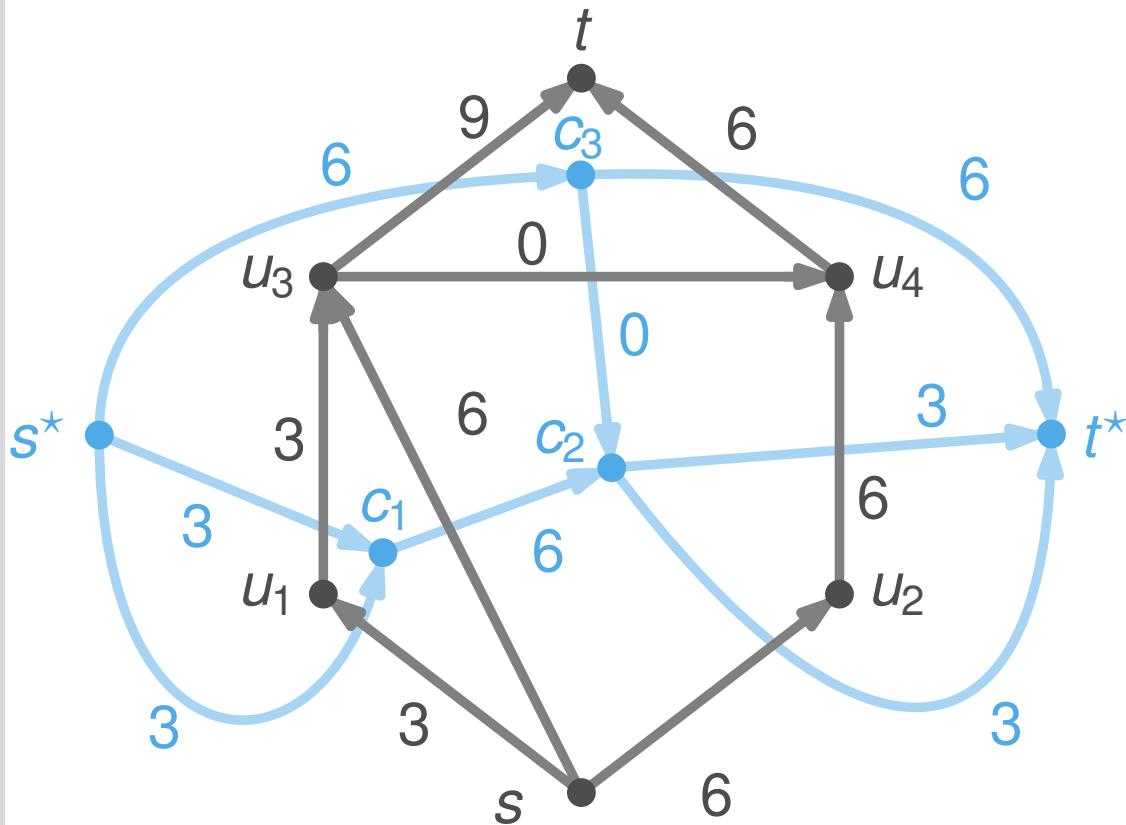
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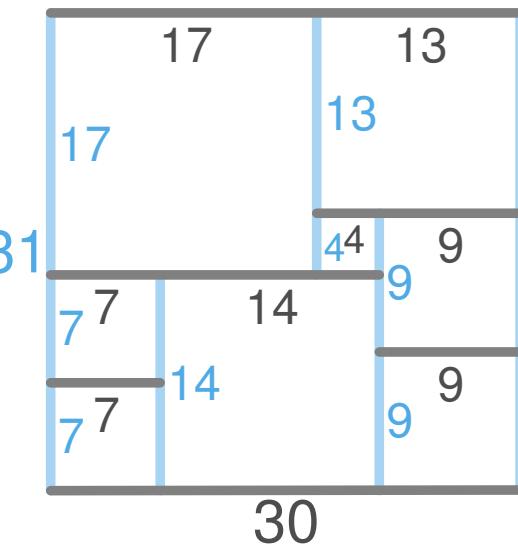
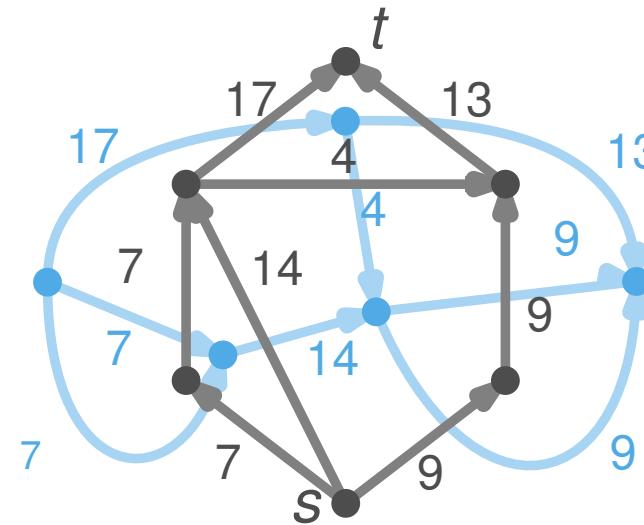
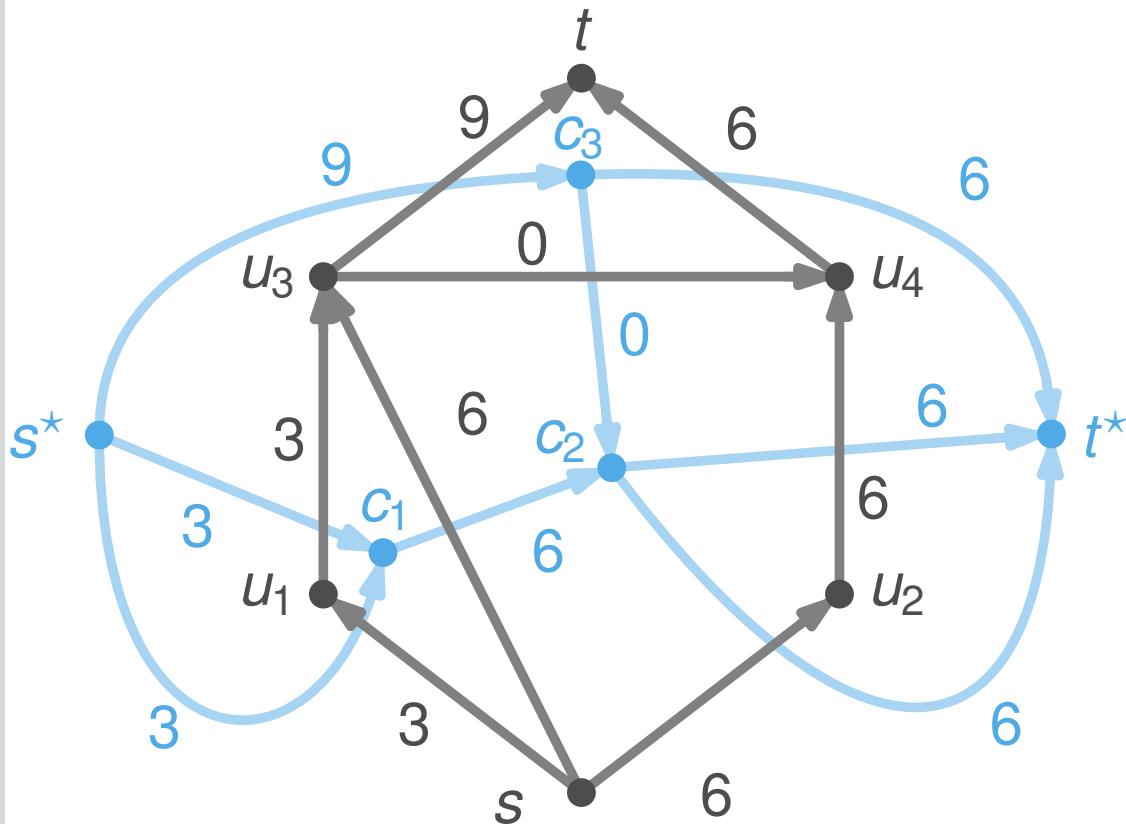
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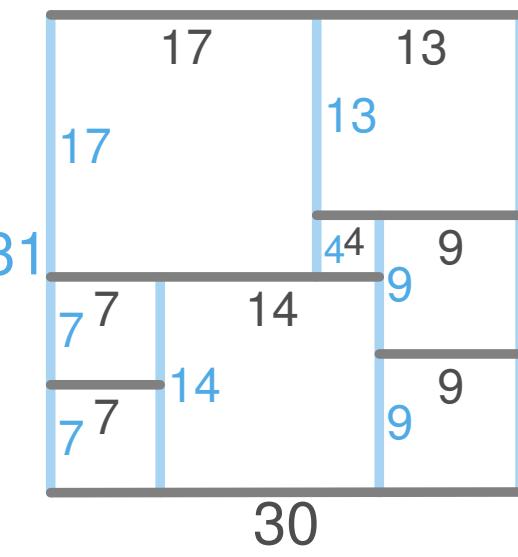
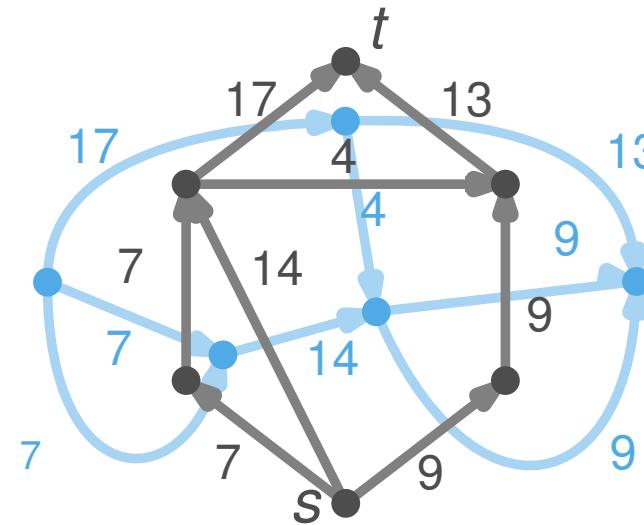
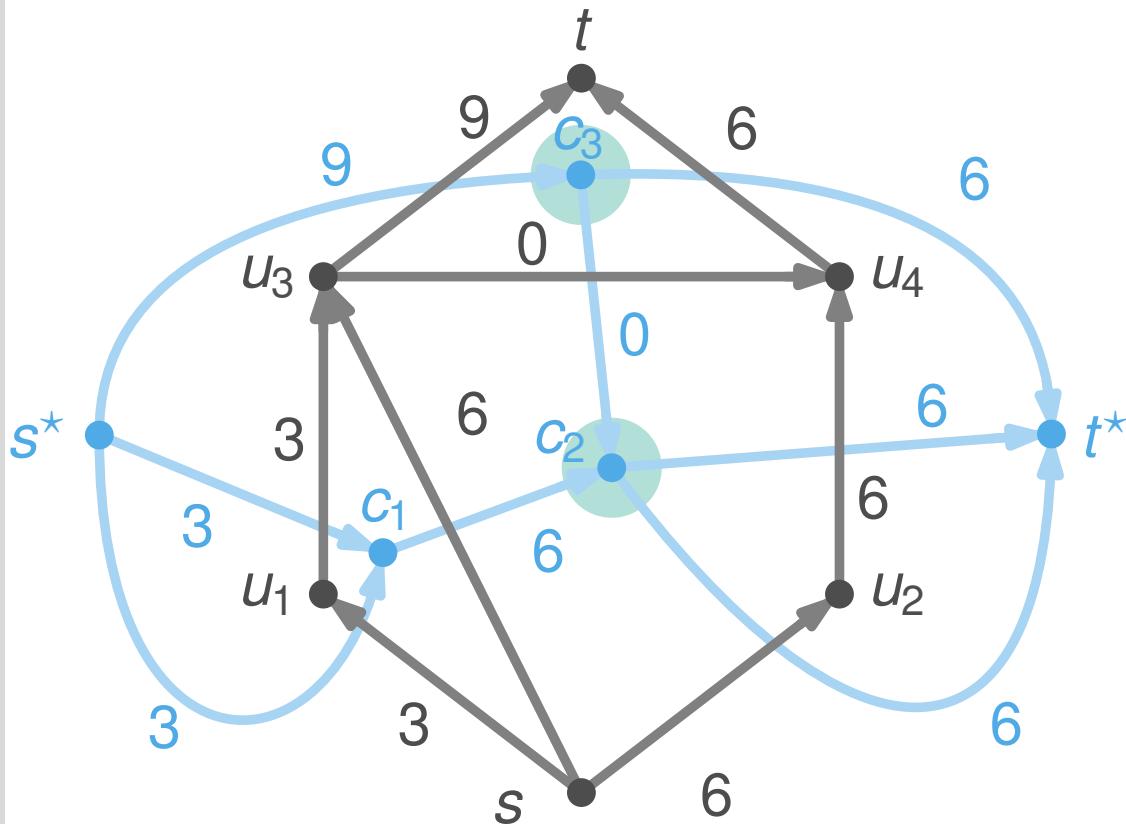
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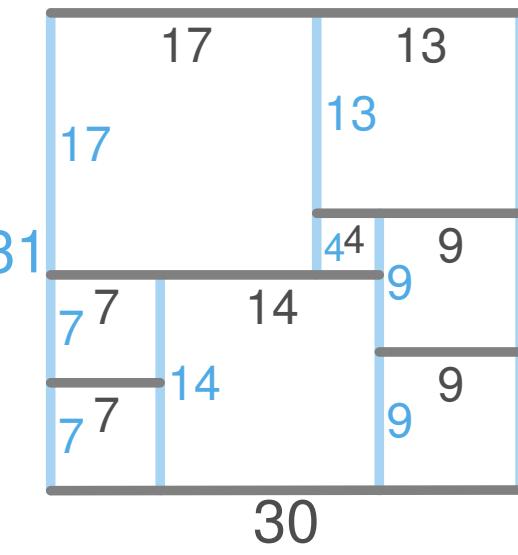
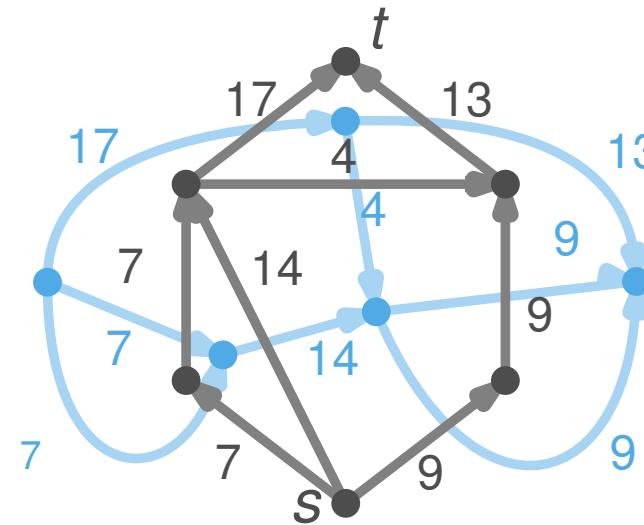
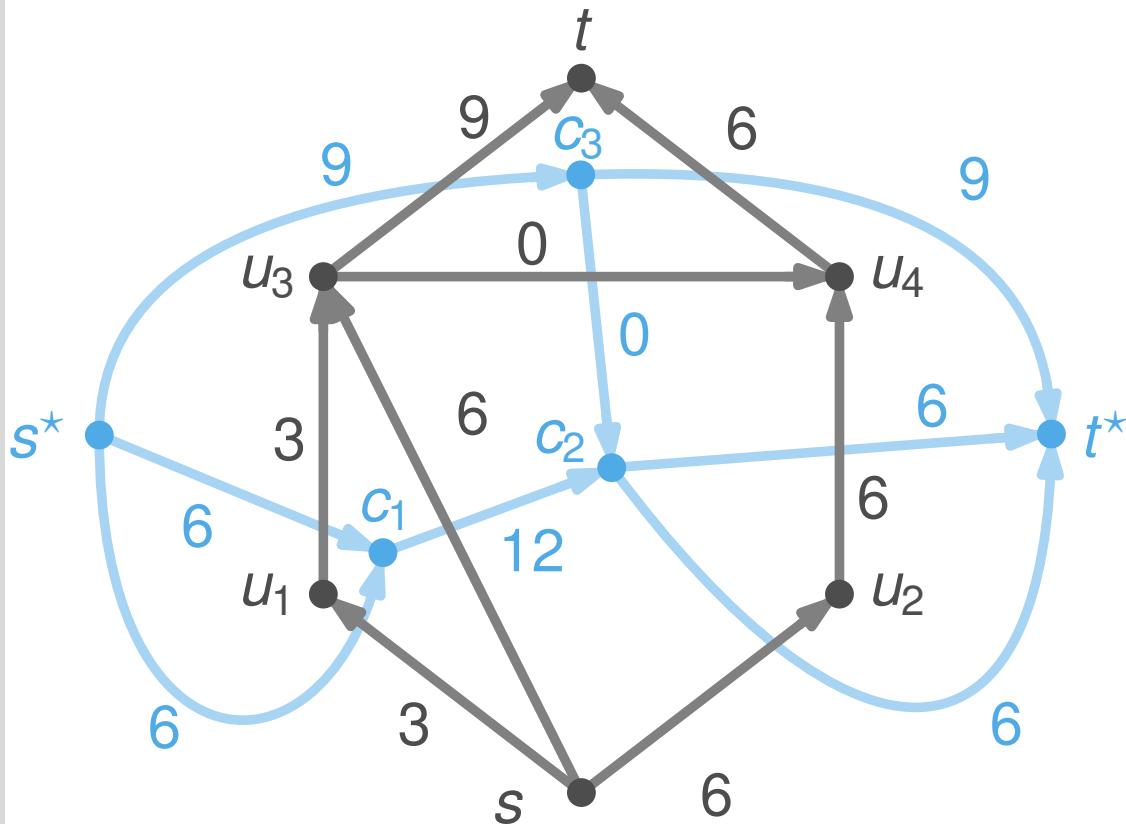
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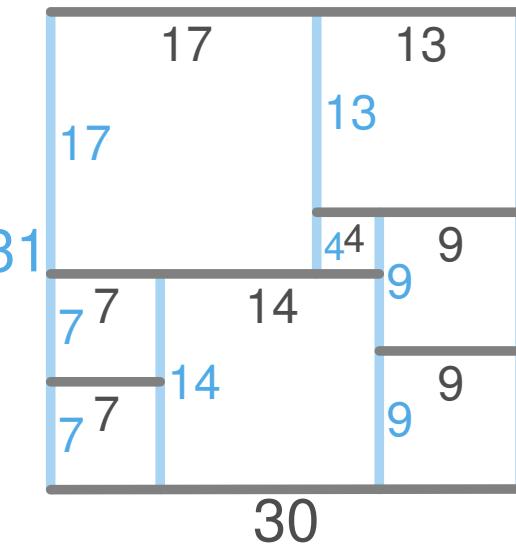
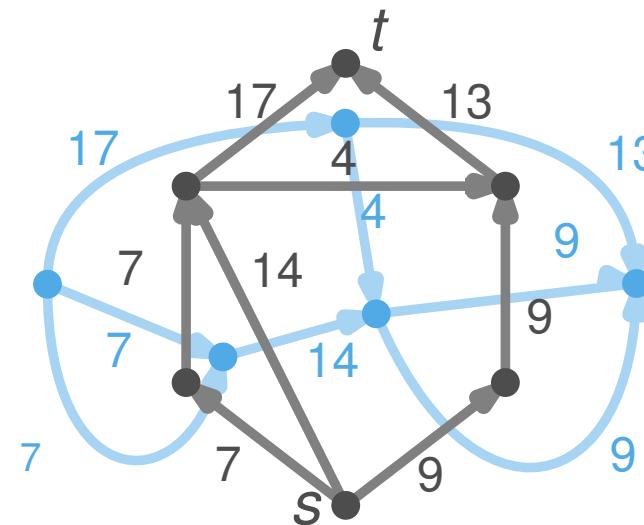
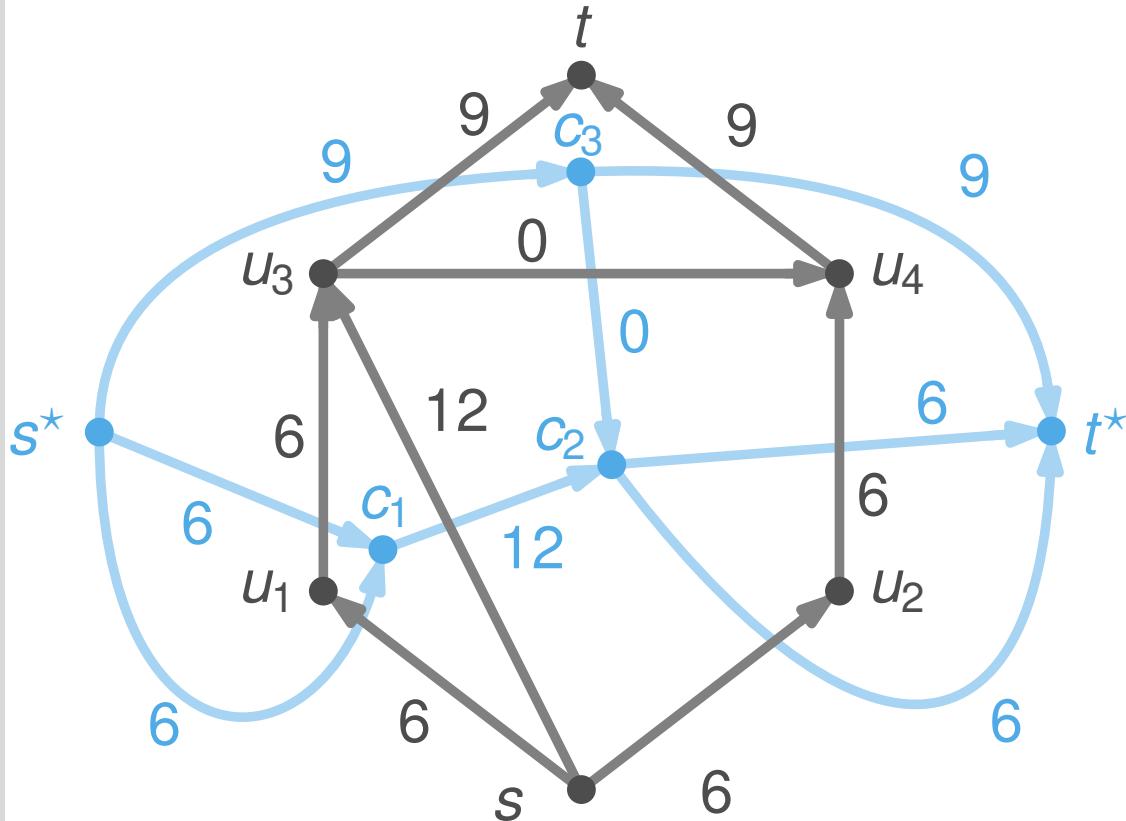
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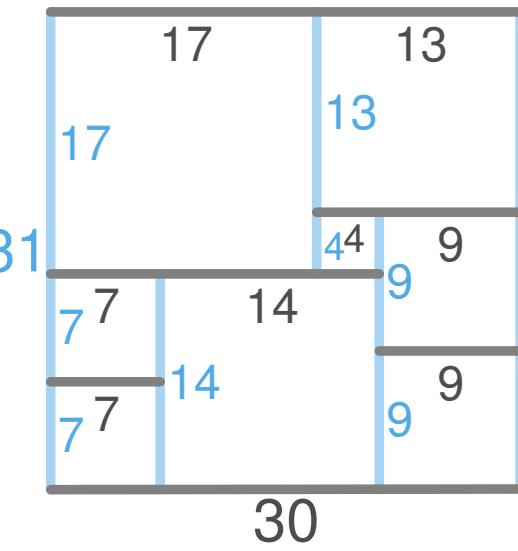
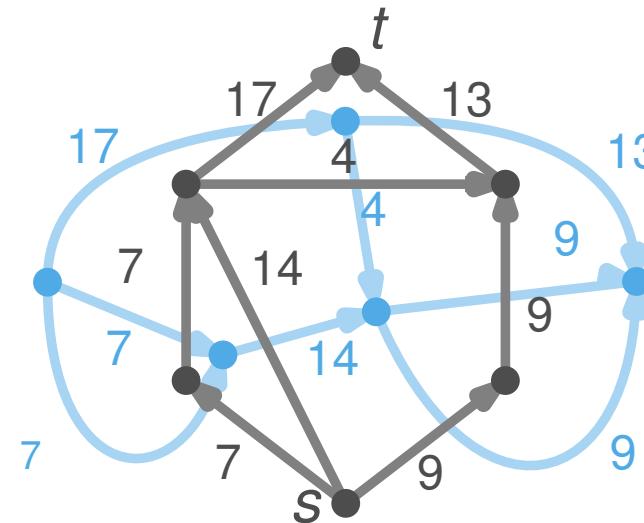
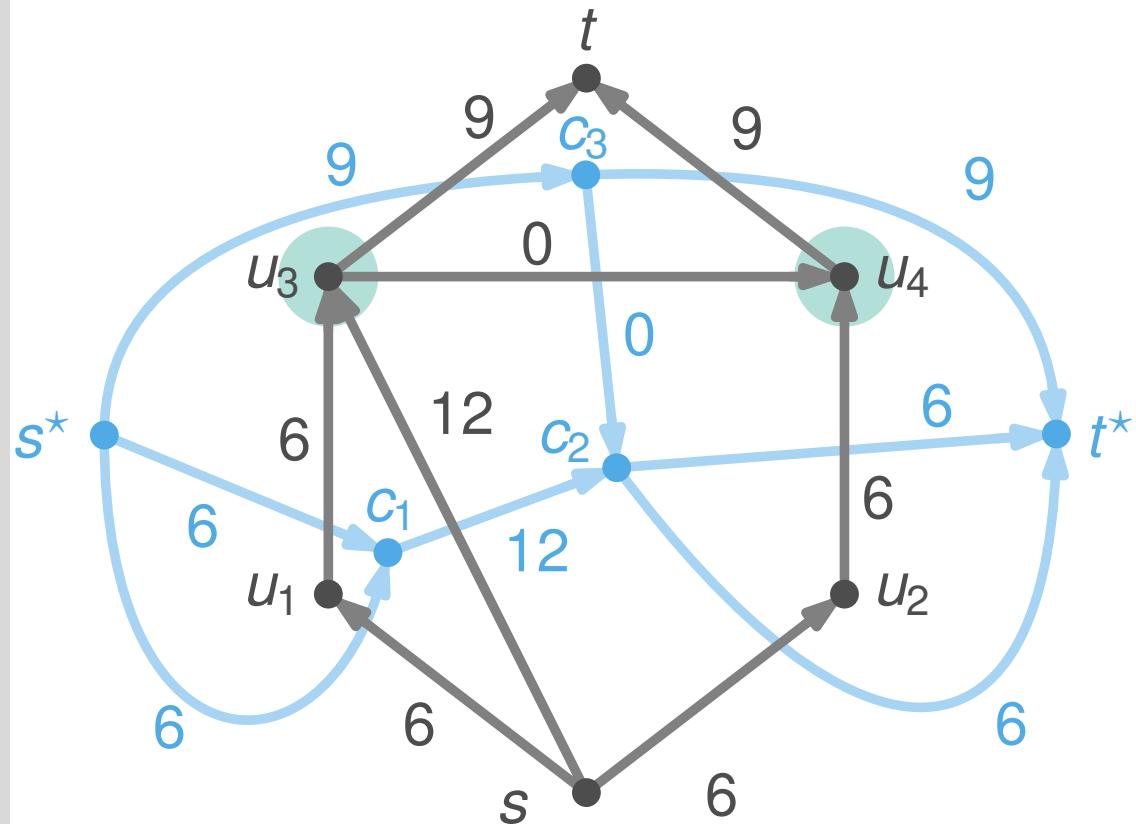
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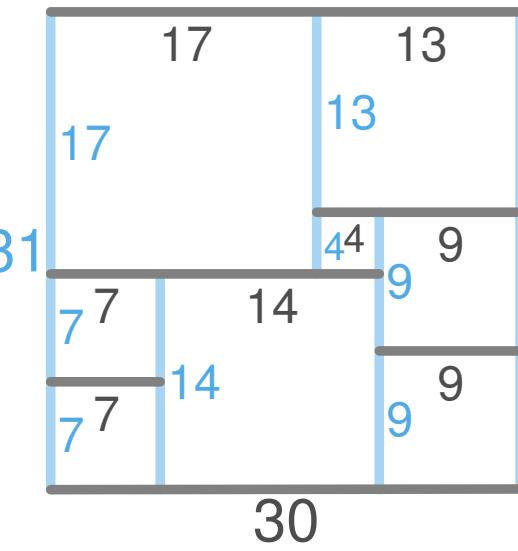
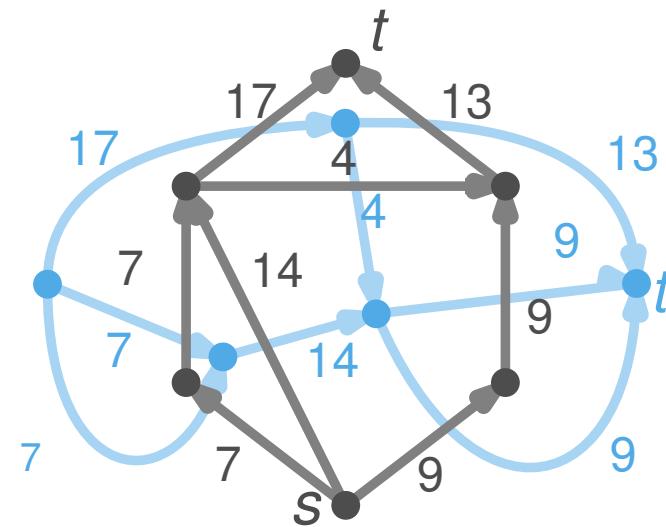
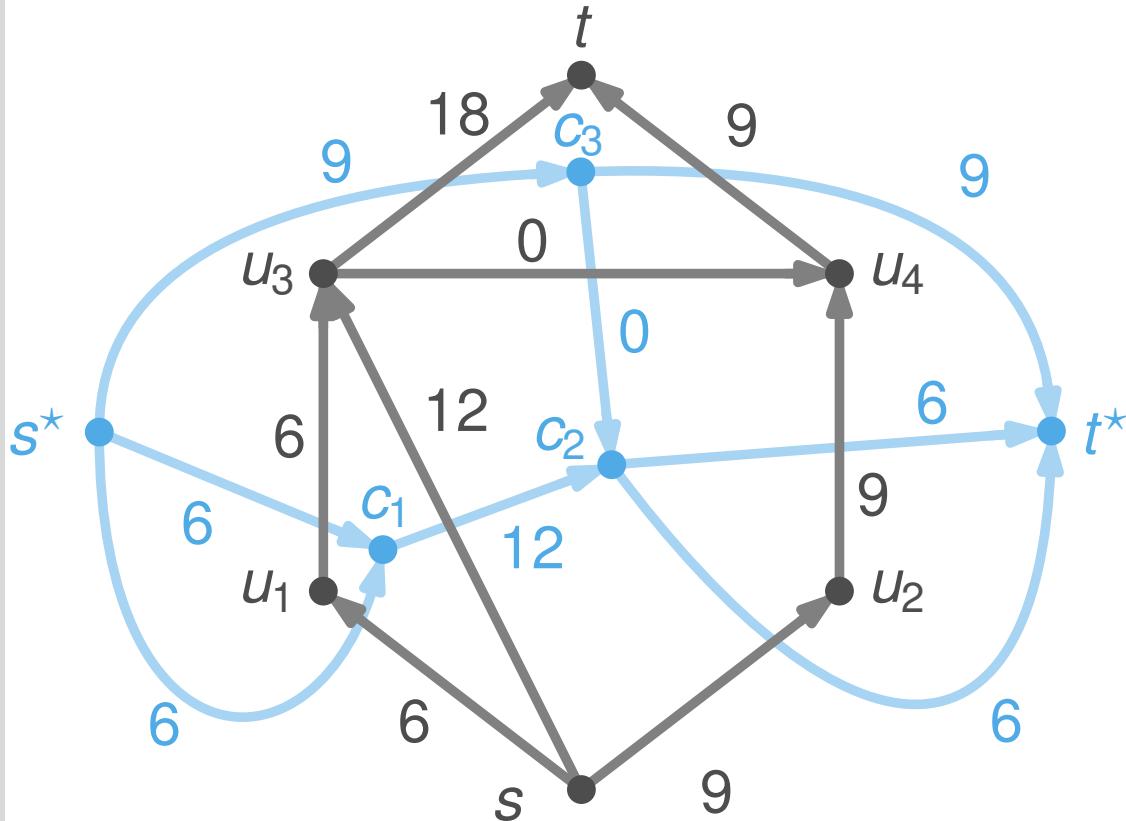
# Wrong Conflict Resolution



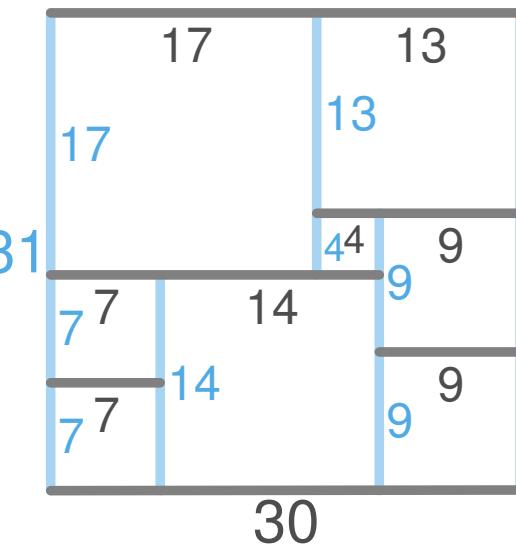
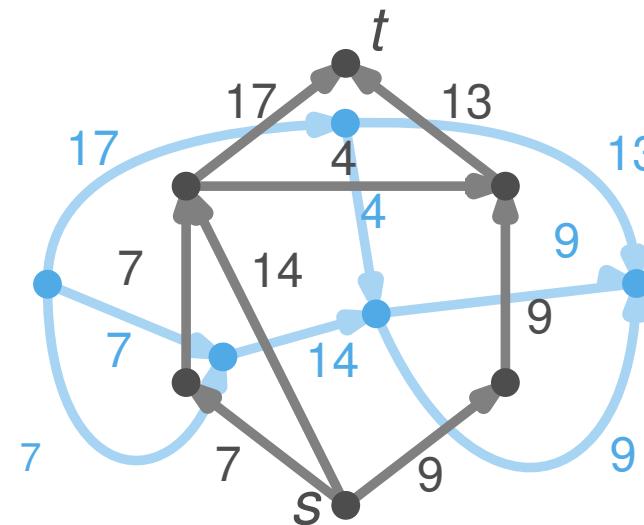
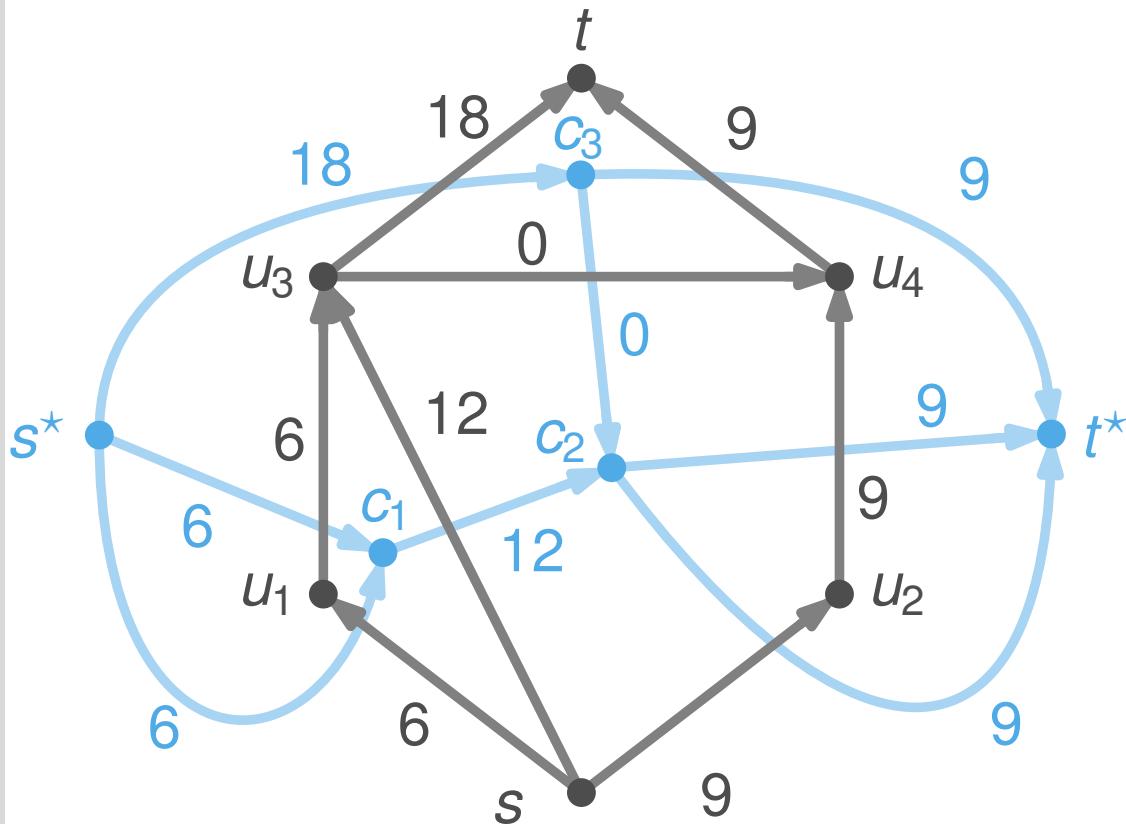
# Wrong Conflict Resolution



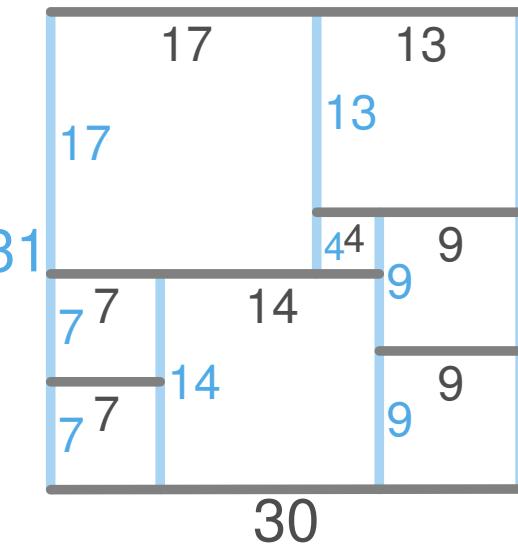
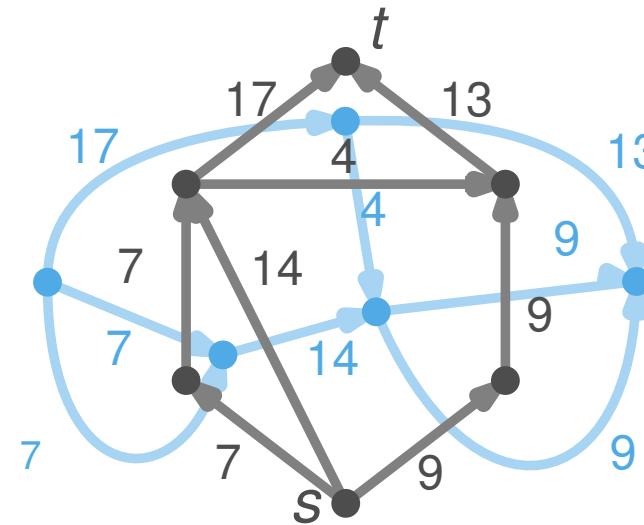
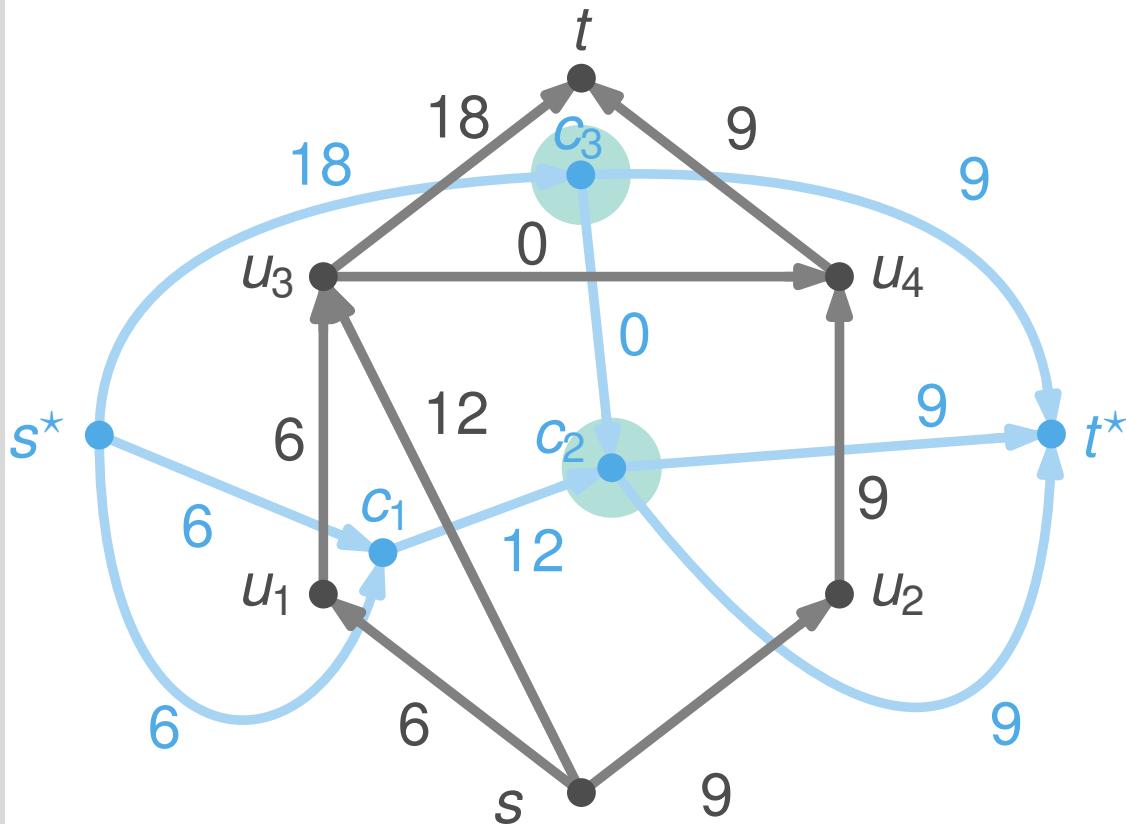
# Wrong Conflict Resolution



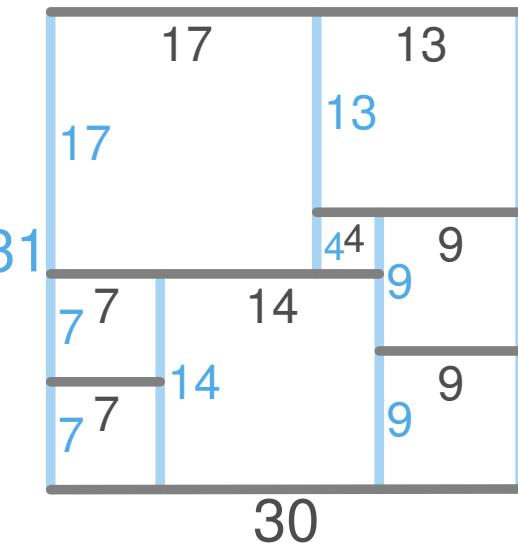
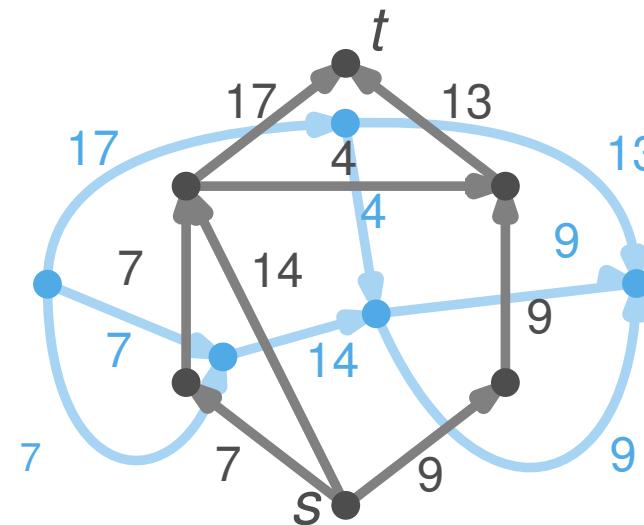
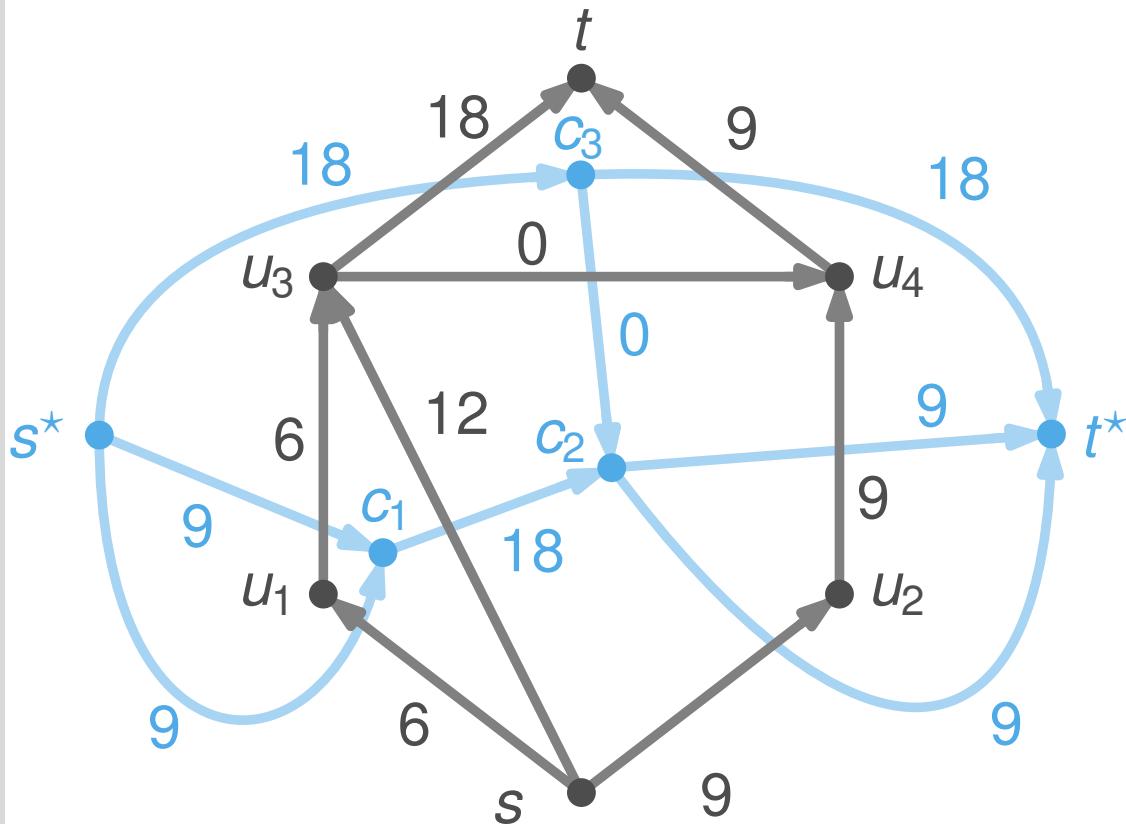
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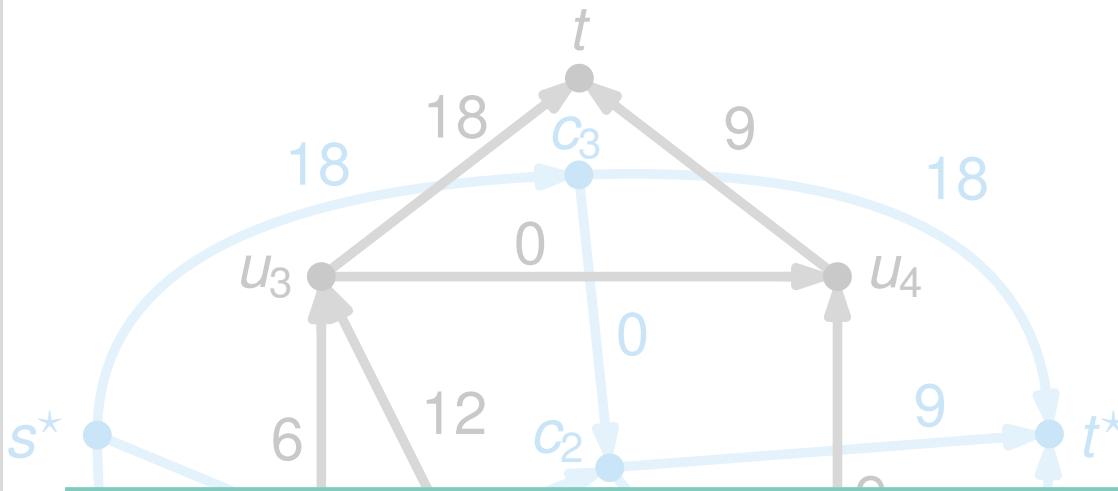
# Wrong Conflict Resolution



# Wrong Conflict Resolution

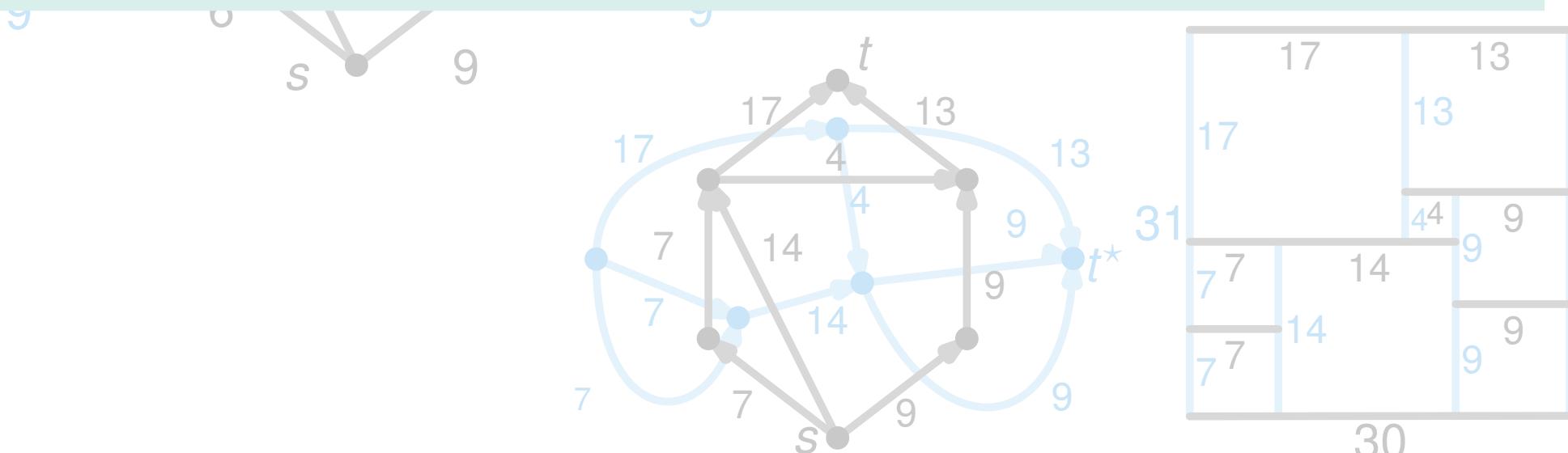


# Wrong Conflict Resolution

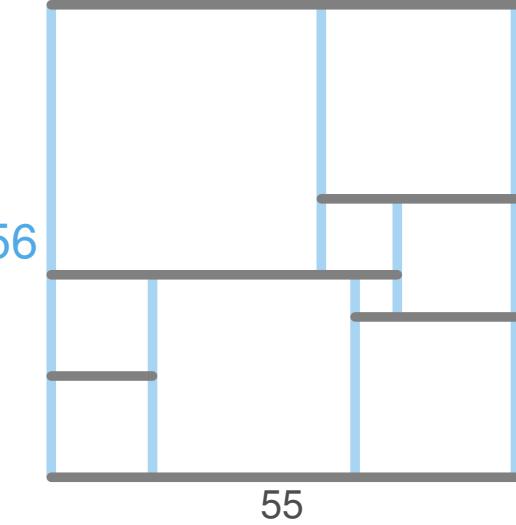
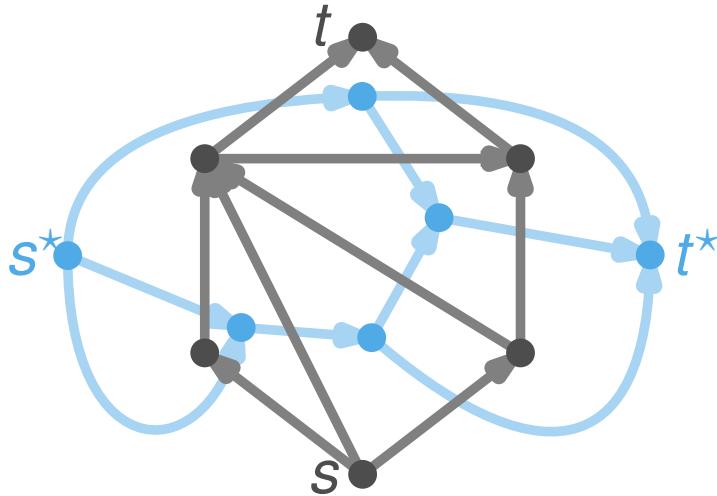
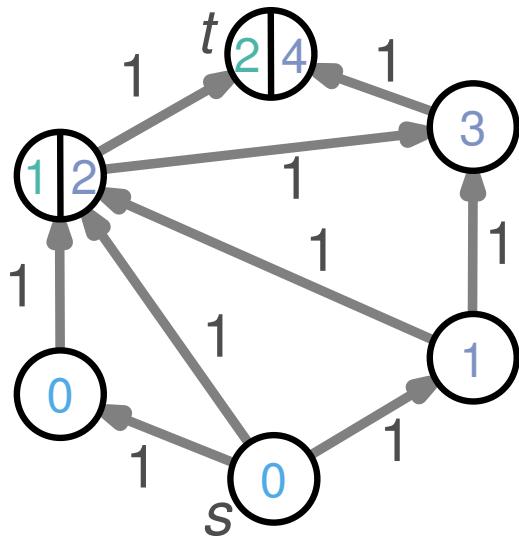


## Observation 6

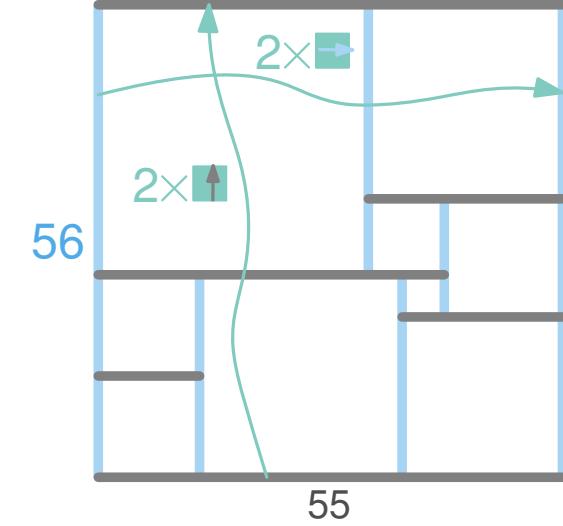
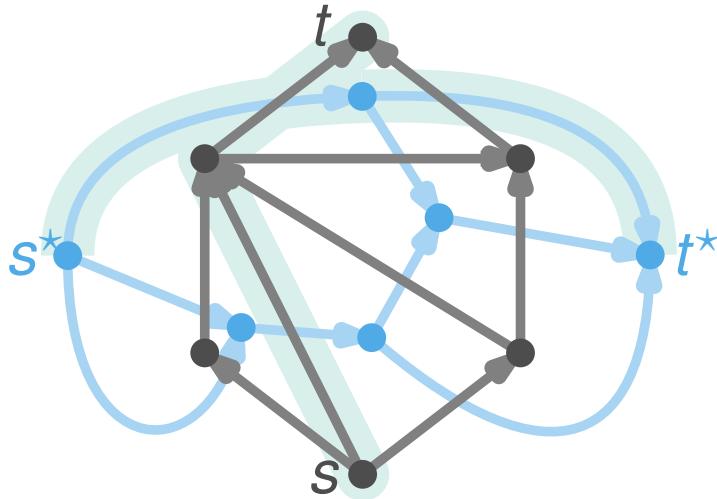
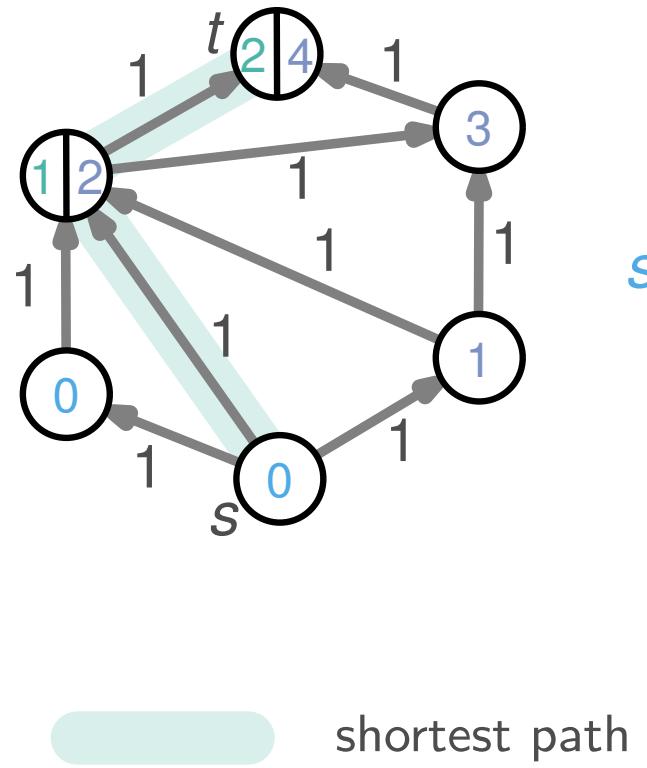
A wrong conflict resolution might never lead to a feasible power flow.



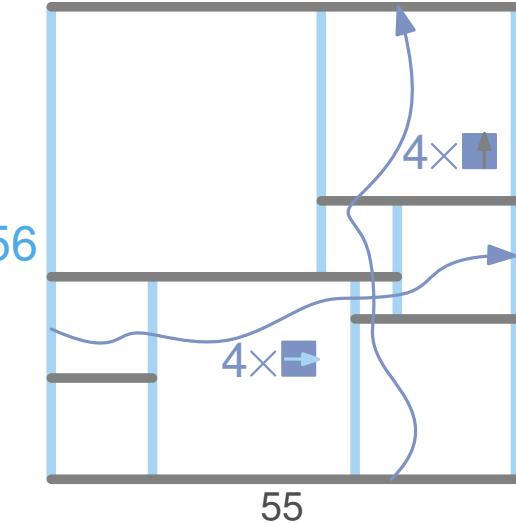
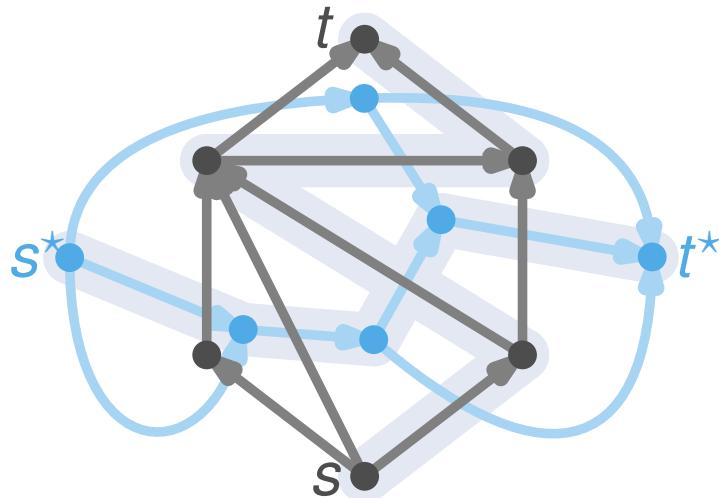
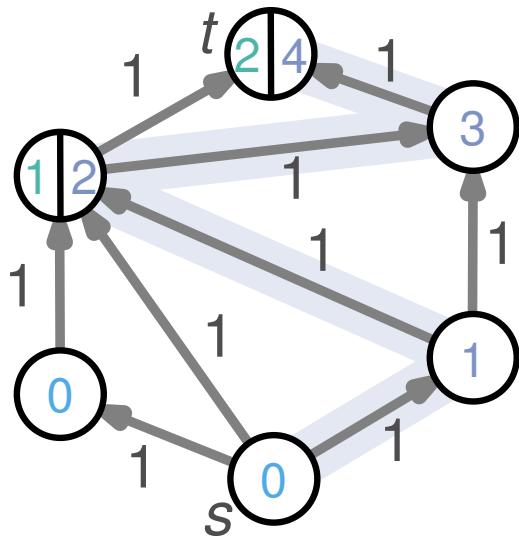
# The Property of Balancing



# The Property of Balancing

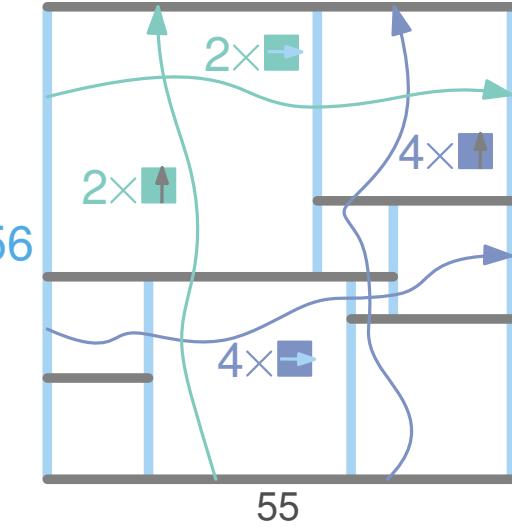
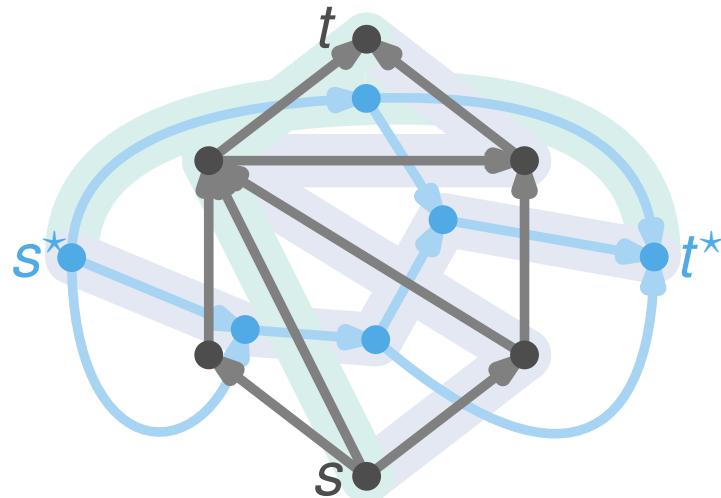
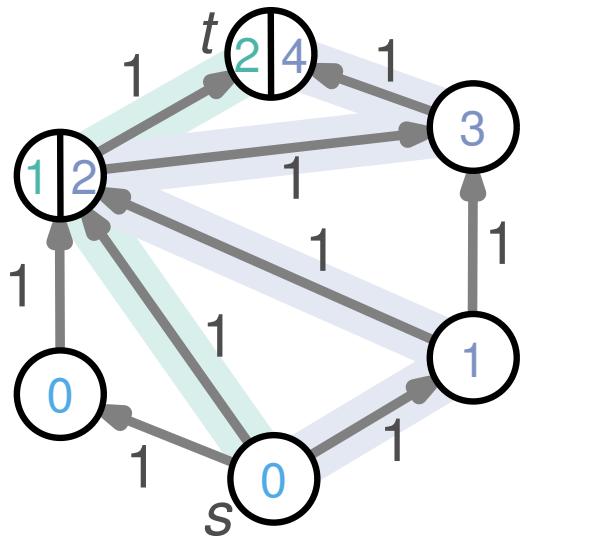


# The Property of Balancing



longest path

# The Property of Balancing



# shortest path

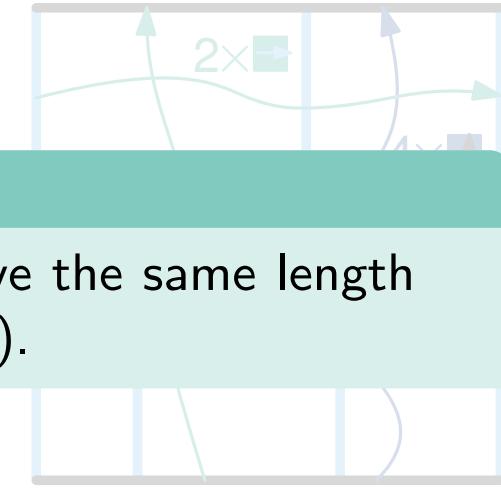
# longest path

# The Property of Balancing



## Observation 7 [Termination]

A flow  $f$  is a power flow if the longest and shortest path have the same length in the primal and dual graph separately (w.r.t. the metric  $f/b$ ).



shortest path

longest path

# Scalability of Power Flows

- Graph-theoretical flow algorithms use scaling techniques
  1. Capacity scaling [Edmonds and Karp, 1972]
  2. Excess scaling [Ahuja and Orlin, 1989]
- Power flows excluding a trivial power flow ( $f \equiv 0$ ) can be scaled up and down by a factor  $\chi$

## Lemma 8 [Scaling]

Every non-zero electrical flow  $f': E \rightarrow \mathbb{R}_{>0}$  can be rescaled to a new feasible electrical flow  $f$  by applying a scaling factor

$$0 \leq \chi \leq \min_{e' \in E} \frac{\text{cap}(e')}{f'(e')} =: \bar{\chi} \quad (1)$$

to  $f(e) = f'(e) \cdot \chi$  for all  $e \in E$ .

# Continuous Changes to the Power Grid

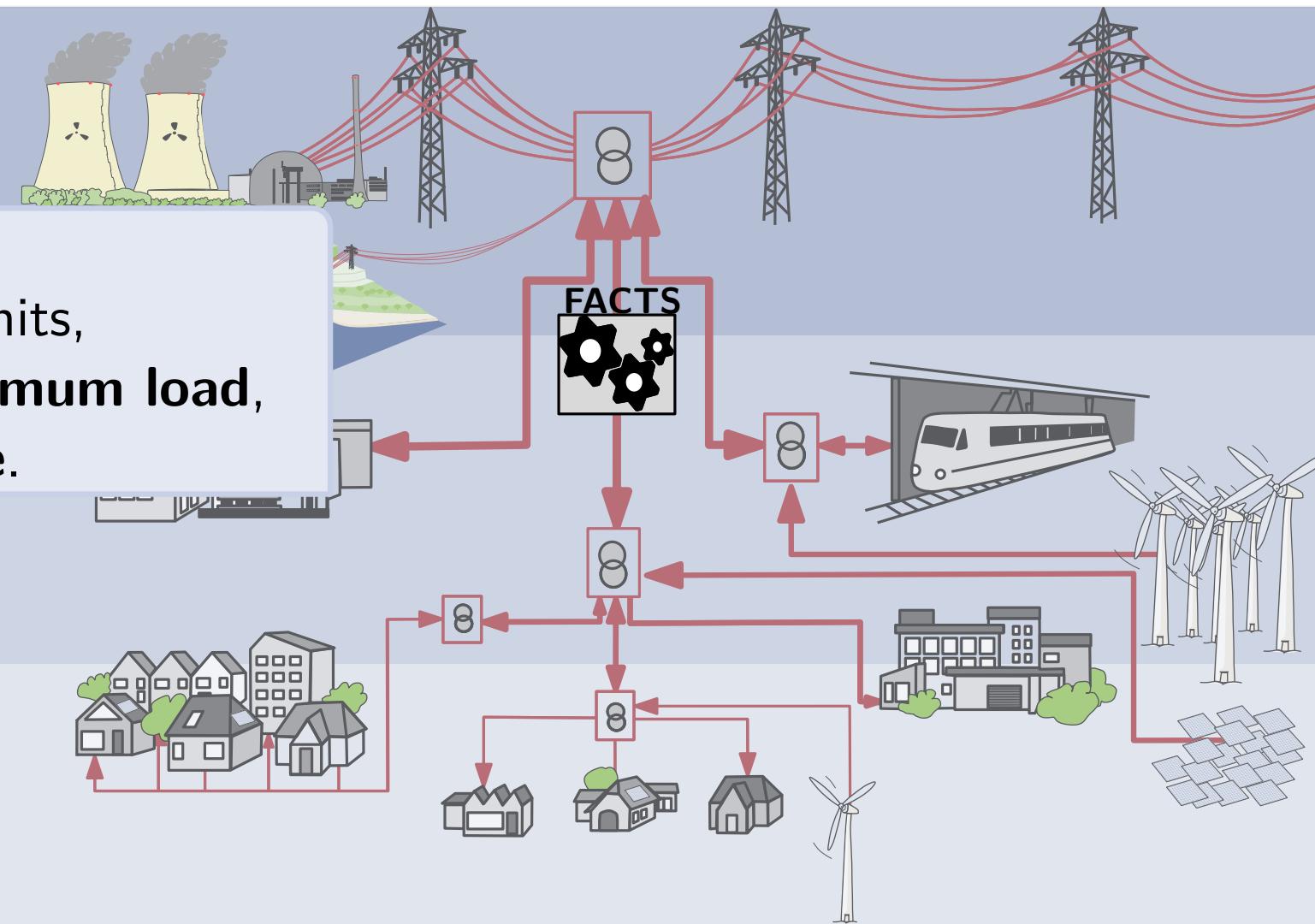
Producer

FACTS...

- are **control units**,
- increase **maximum load**,
- are **expensive**.

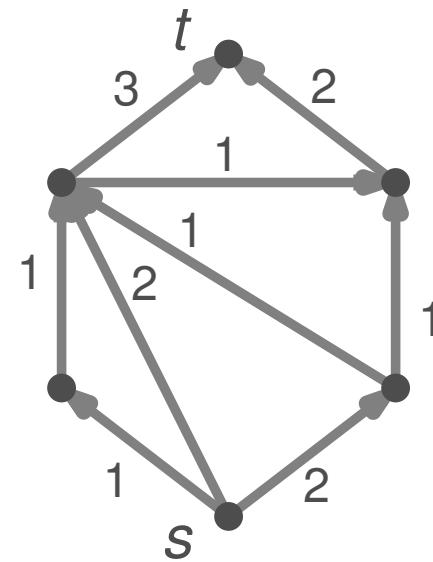
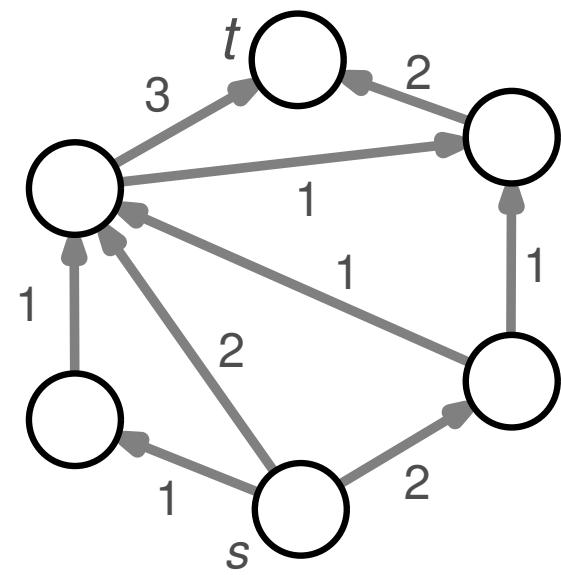
Power Grid

Prosumer



# Susceptance Scaling

- Apply a **feasible flow**

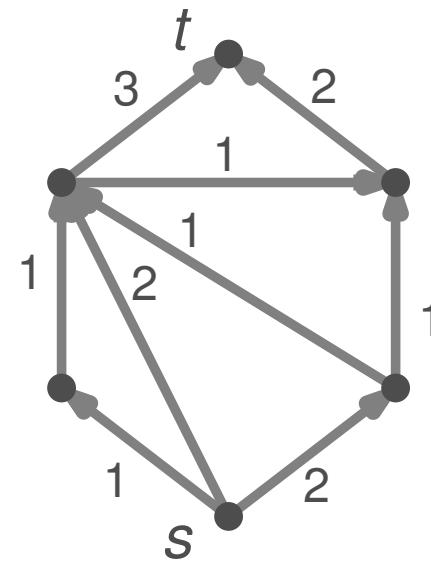
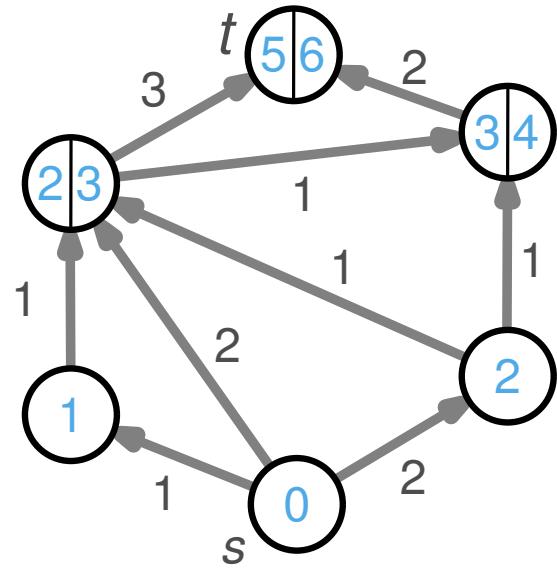


## Observation 9 [KCL and KVL Duality]

KVL conflict in the primal graph  $G$  is a KCL conflict in the dual graph  $G^*$

# Susceptance Scaling

- Apply a **feasible flow**

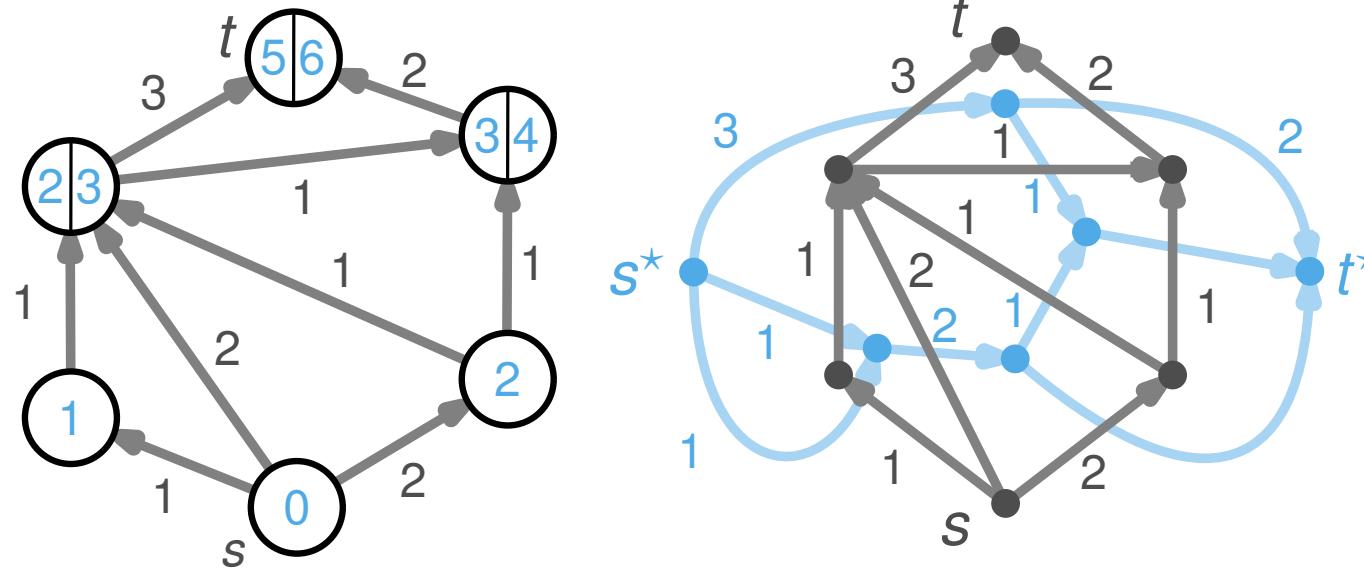


## Observation 9 [KCL and KVL Duality]

KVL conflict in the primal graph  $G$  is a KCL conflict in the dual graph  $G^*$

# Susceptance Scaling

- Apply a **feasible flow**
- Can we make it **electrical feasible?**

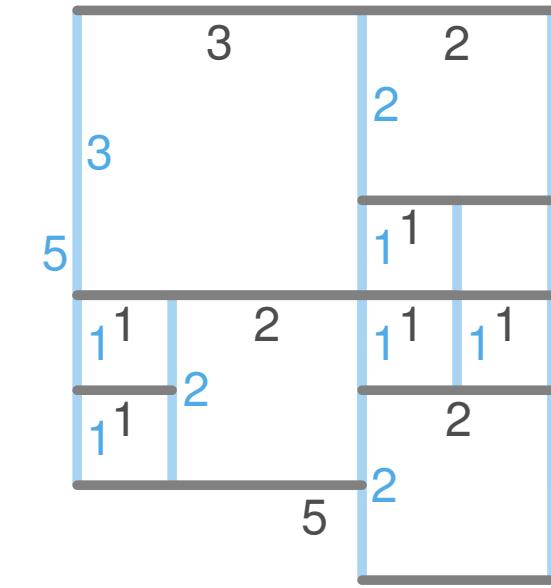
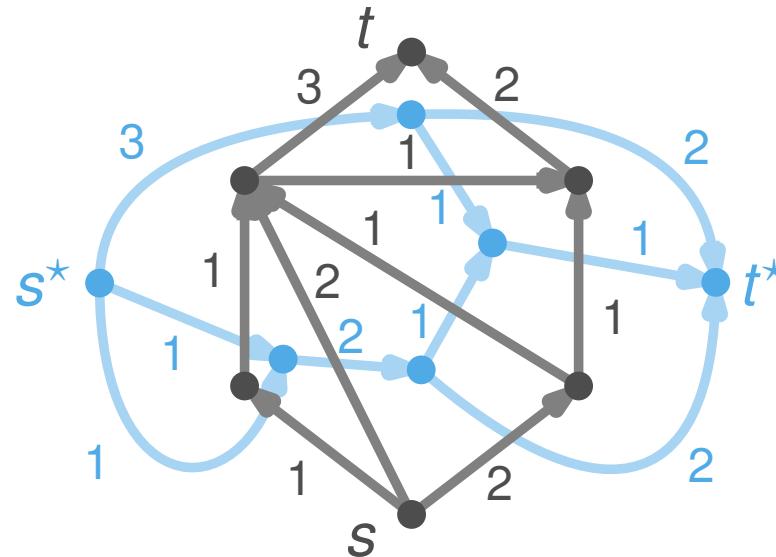
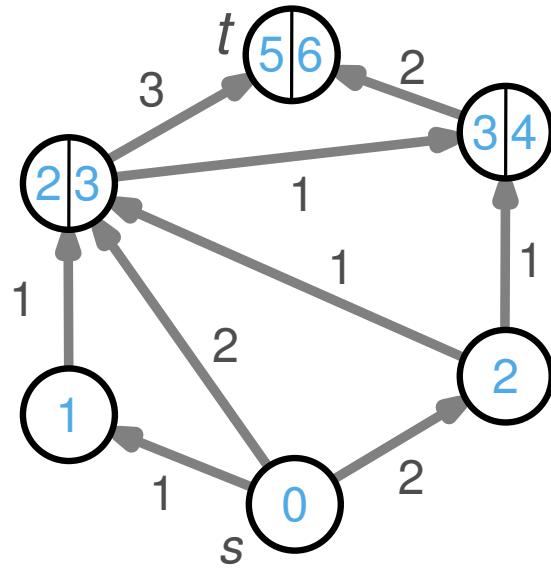


## Observation 9 [KCL and KVL Duality]

KVL conflict in the primal graph  $G$  is a KCL conflict in the dual graph  $G^*$

# Susceptance Scaling

- Apply a **feasible flow**
- Can we make it **electrical feasible?**

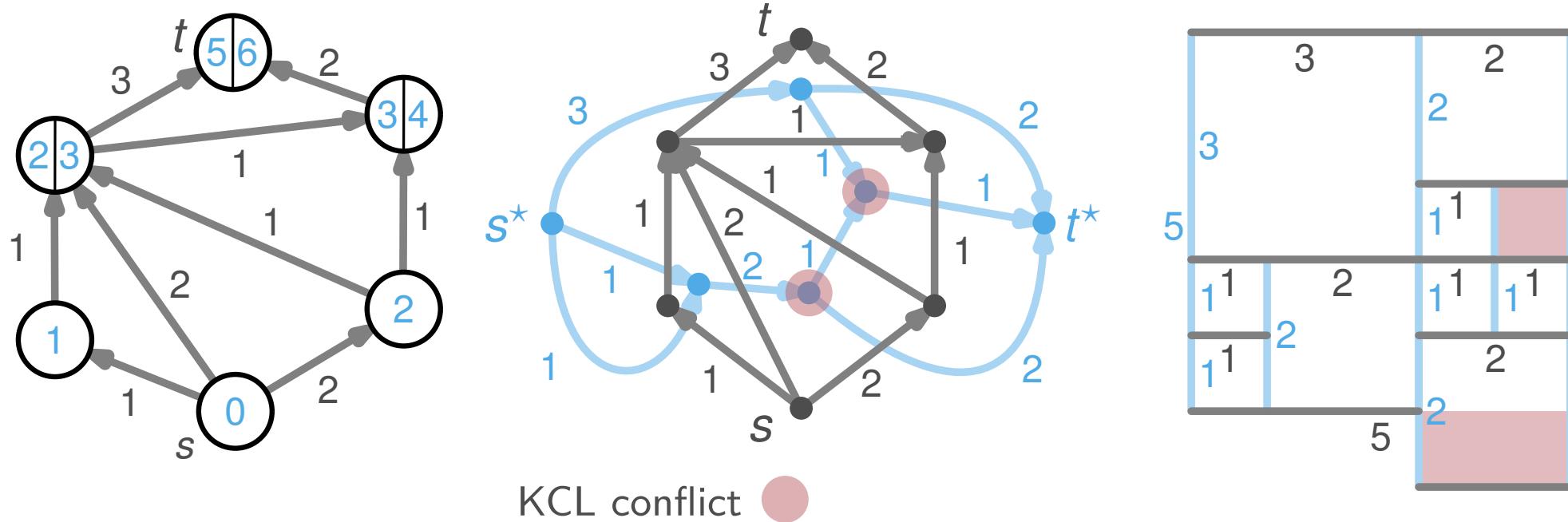


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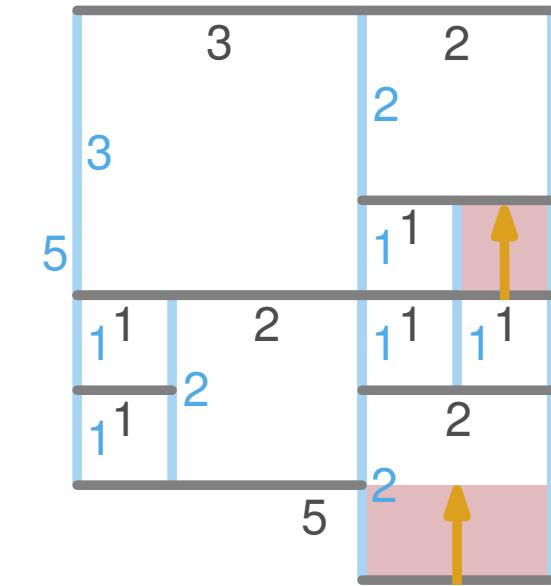
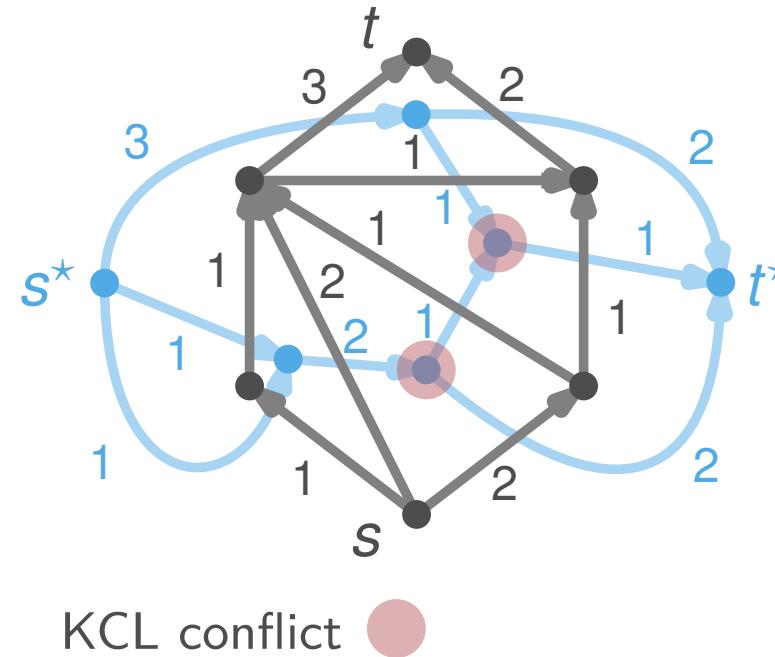
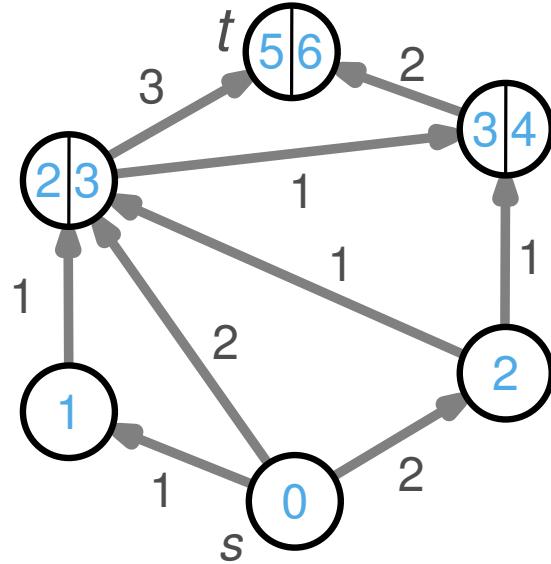


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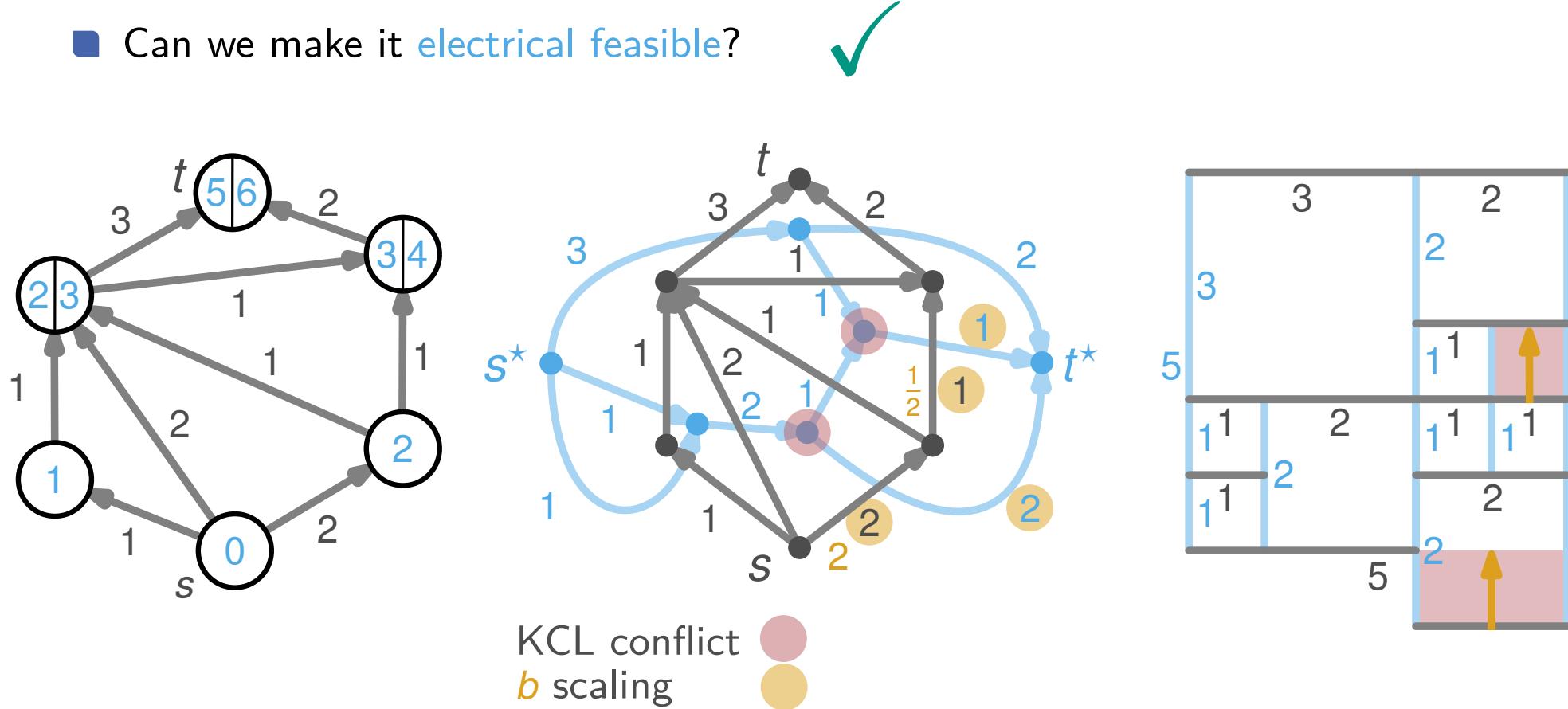


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# Susceptance Scaling

- Apply a **feasible flow**
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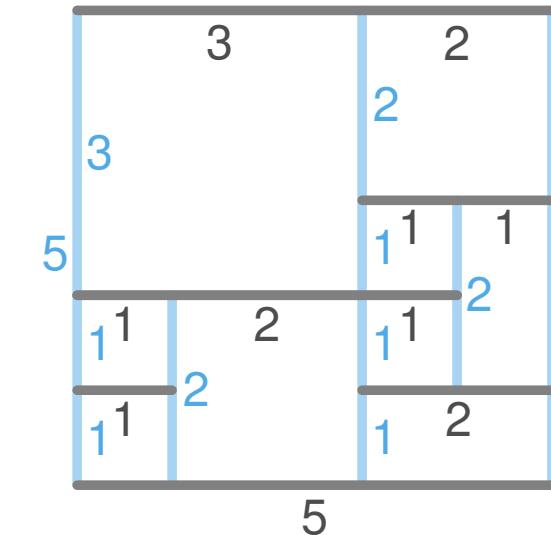
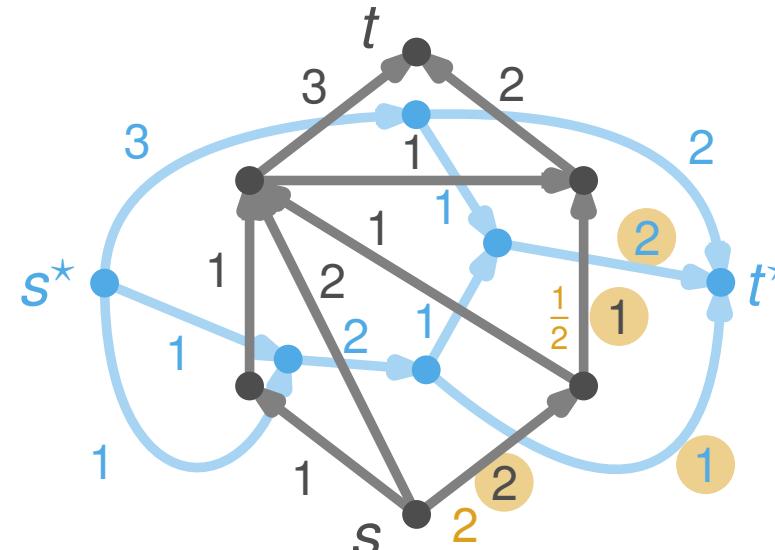
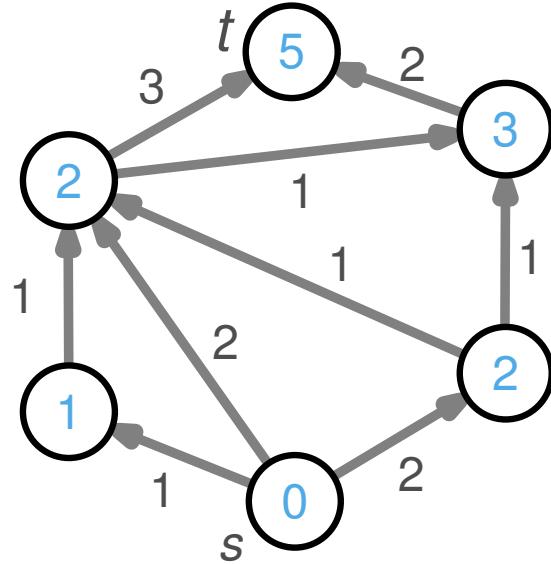


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# Susceptance Scaling

- Apply a **feasible flow**
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$b$  scaling

## Observation 9 [KCL and KVL Duality]

KVL conflict in the primal graph  $G$  is a KCL conflict in the dual graph  $G^*$

# Discrete Changes to the Power Grid

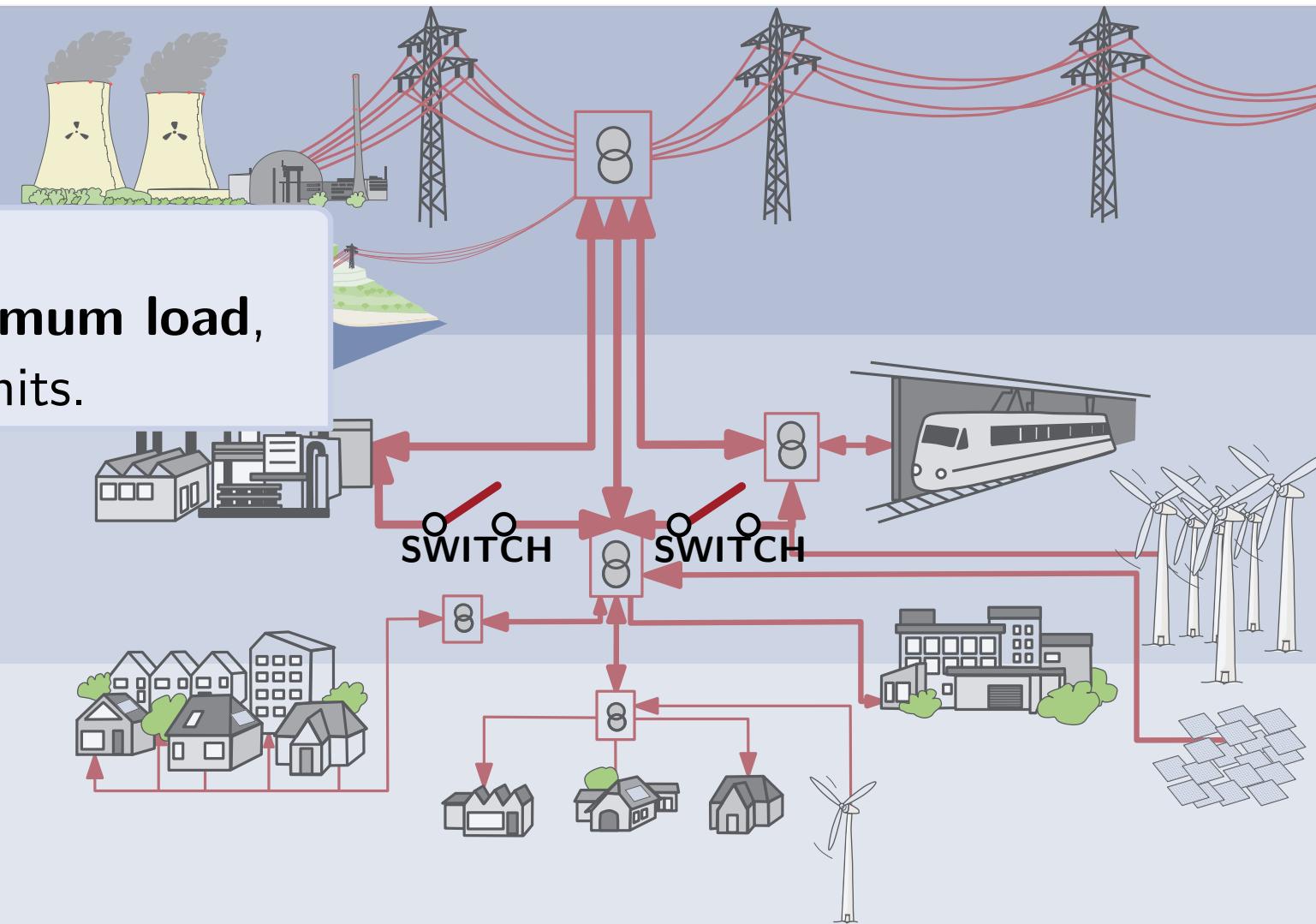
Producer

Switches...

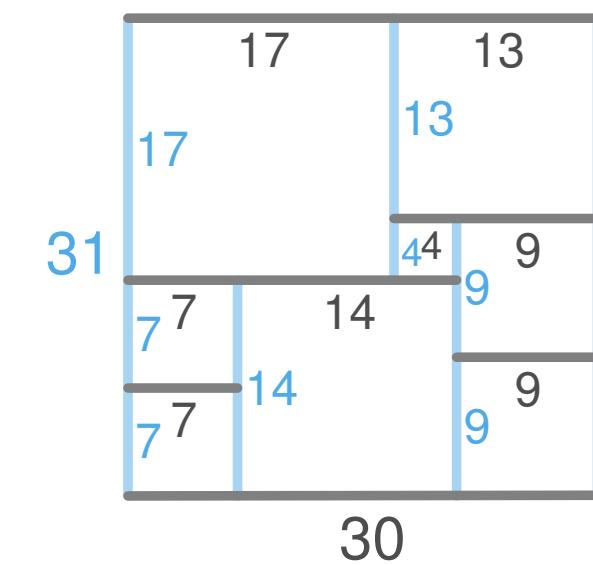
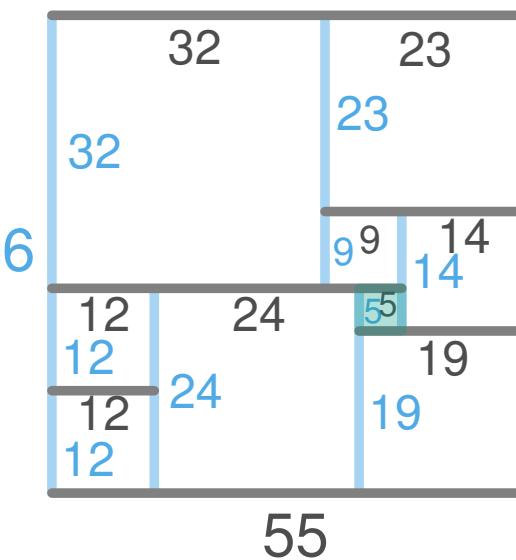
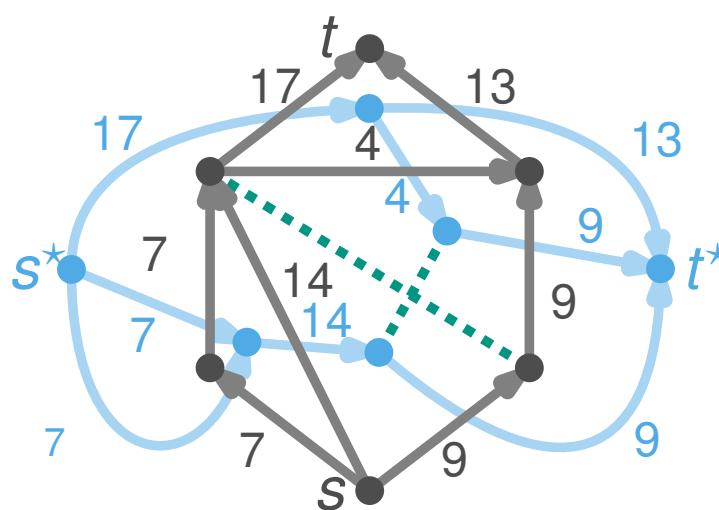
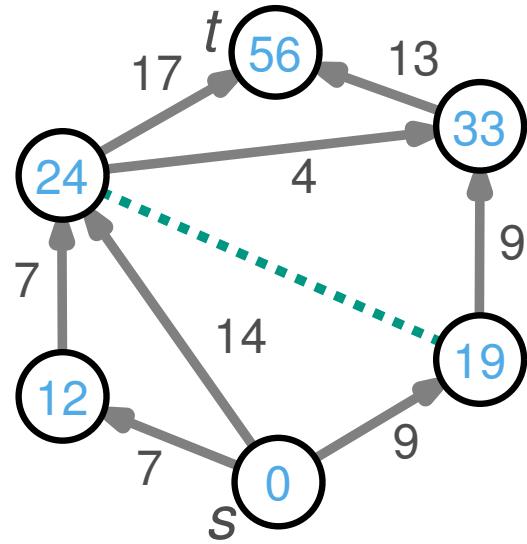
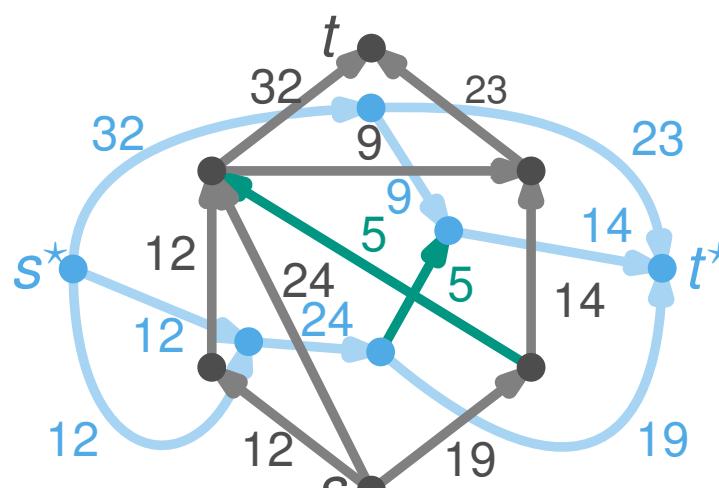
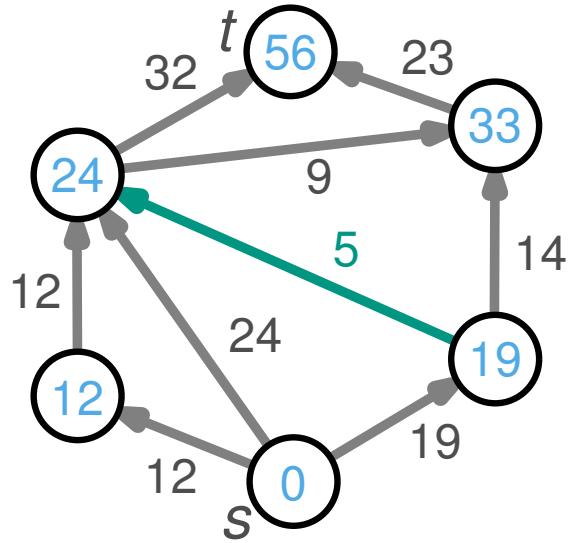
- increase **maximum load**,
- are **control units**.

Power Grid

Prosumer



# Switching



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