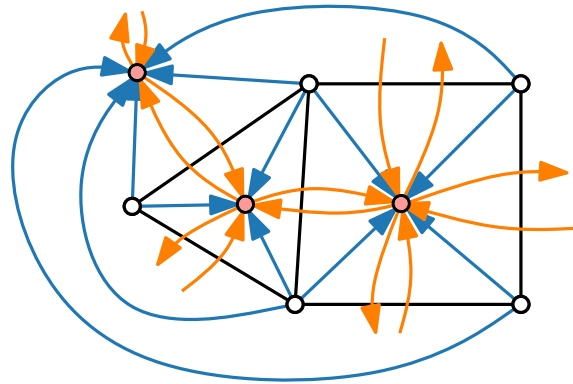
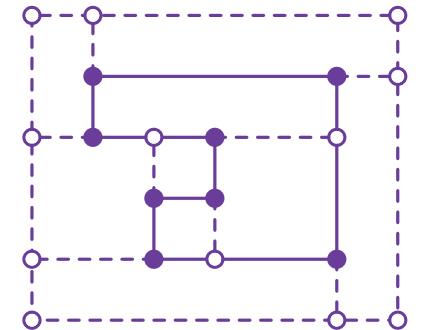


# Visualization of Graphs

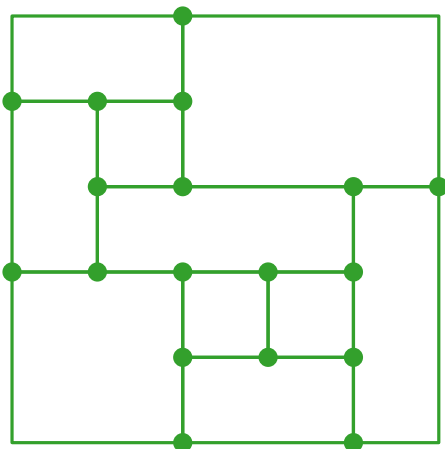


## Lecture 6: Orthogonal Layouts

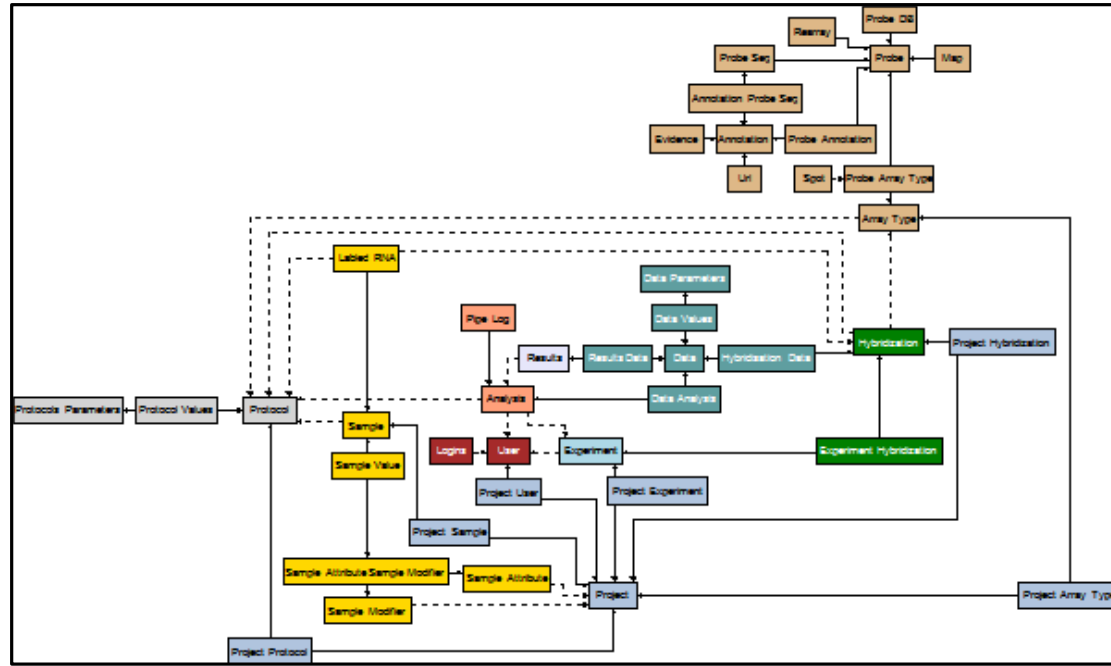


### Part I: Topology – Shape – Metrics

Philipp Kindermann

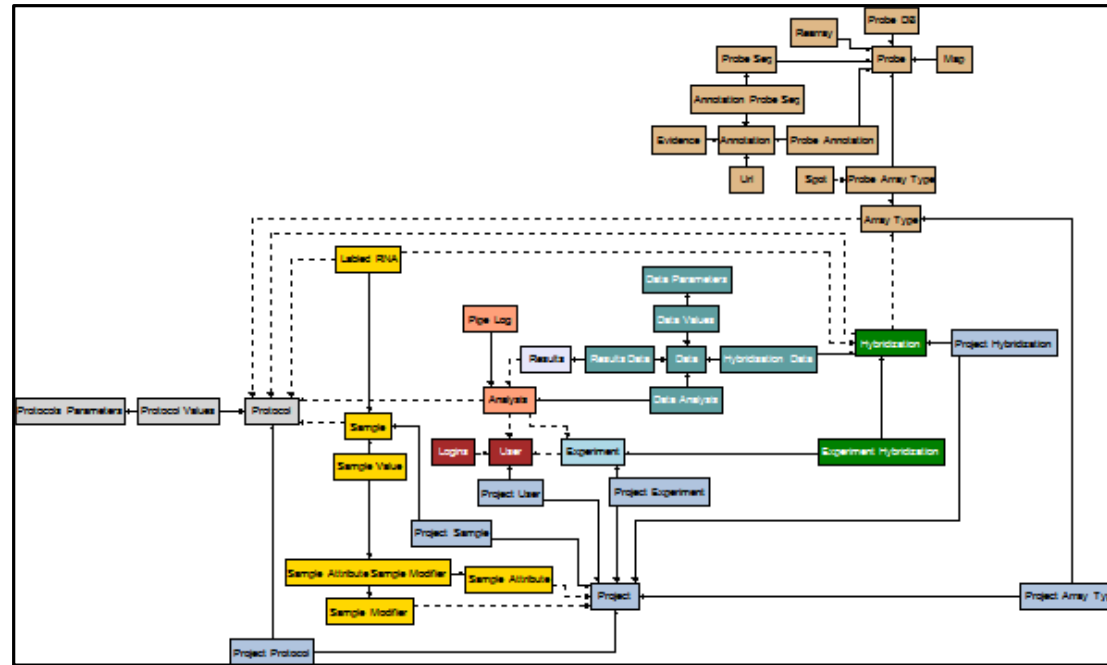


# Orthogonal Layout – Applications

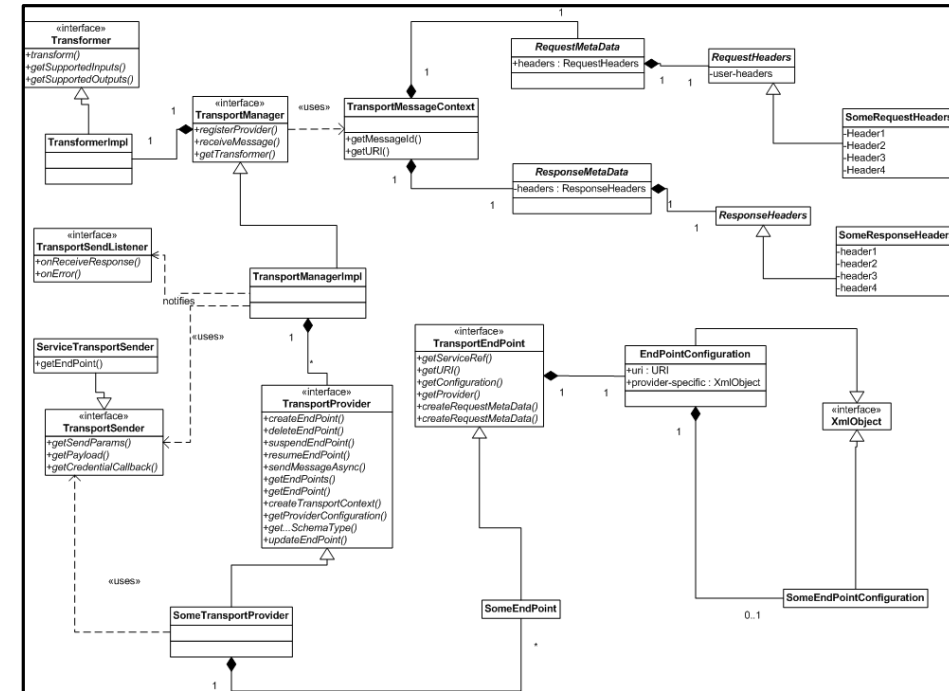


ER diagram in OGDF

# Orthogonal Layout – Applications

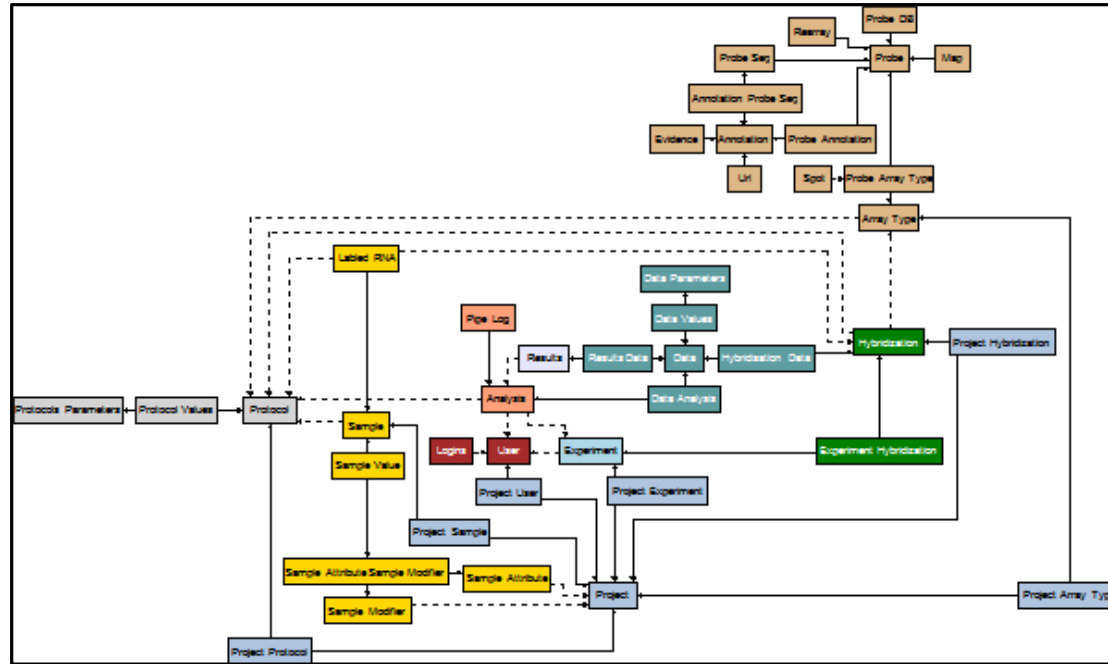


ER diagram in OGDF

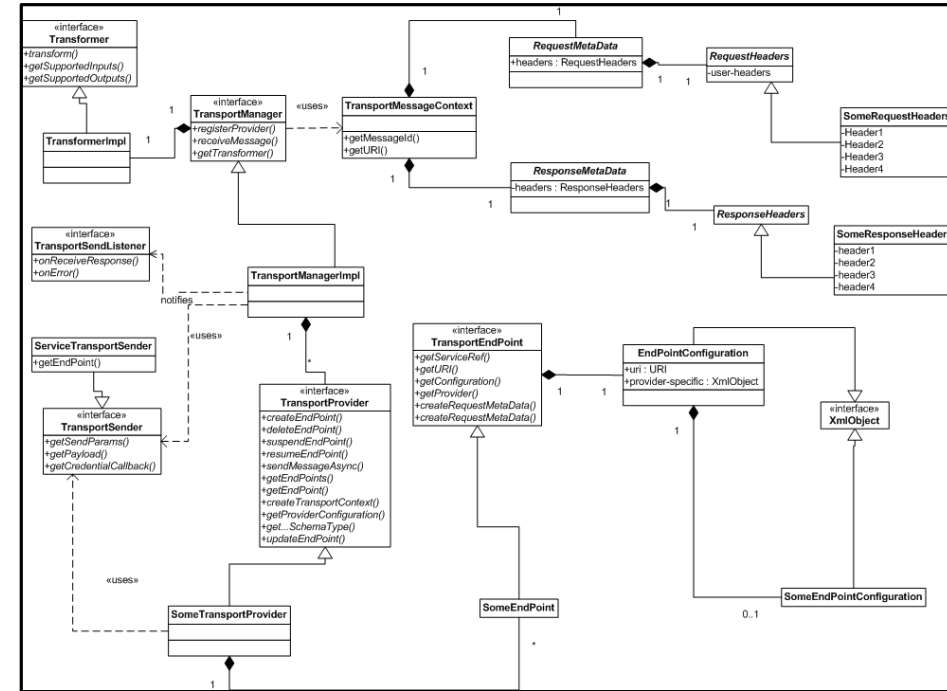


UML diagram by Oracle

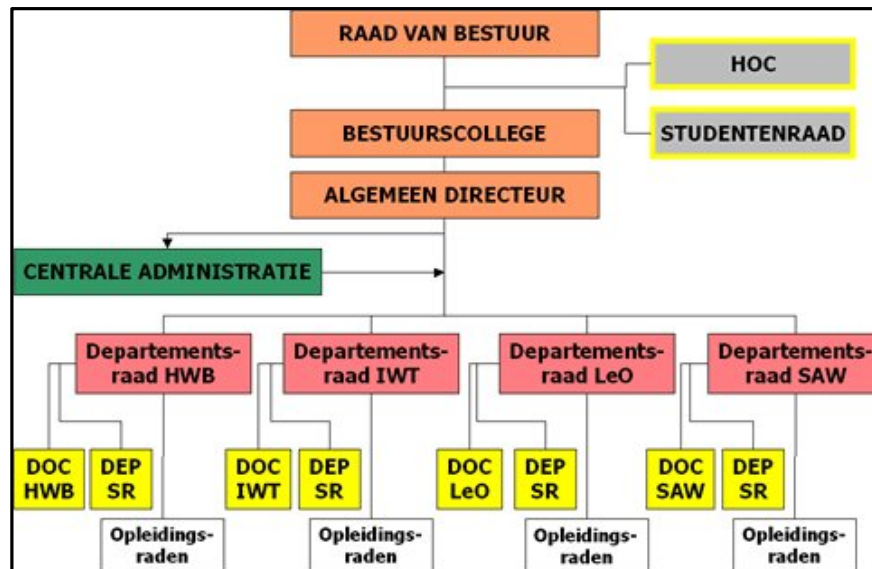
# Orthogonal Layout – Applications



ER diagram in OGDF



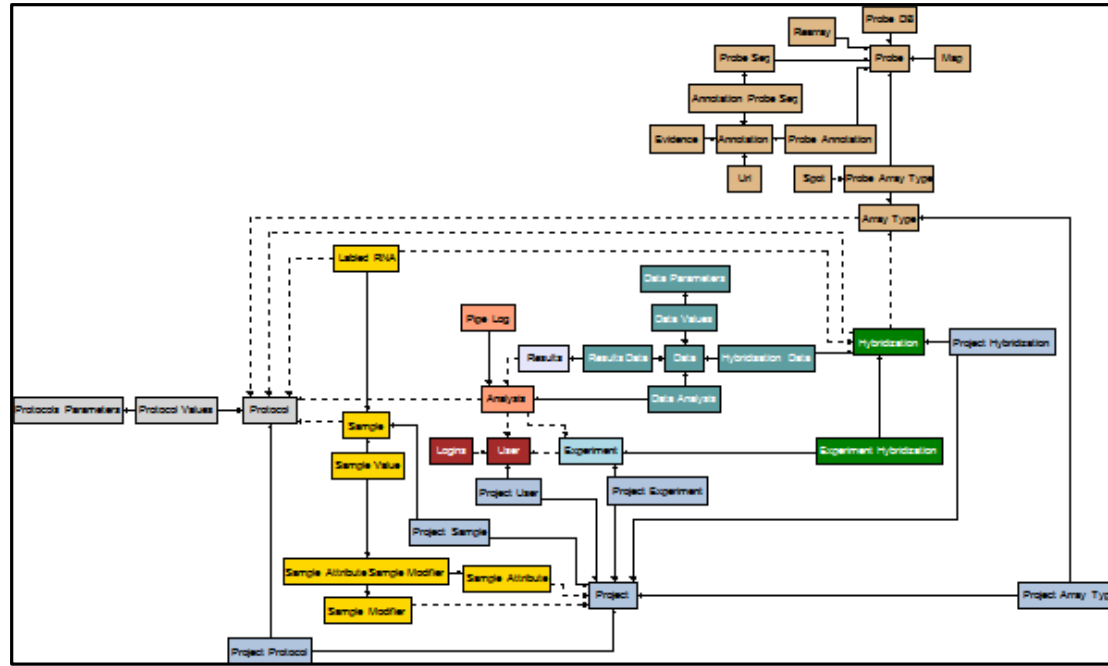
UML diagram by Oracle



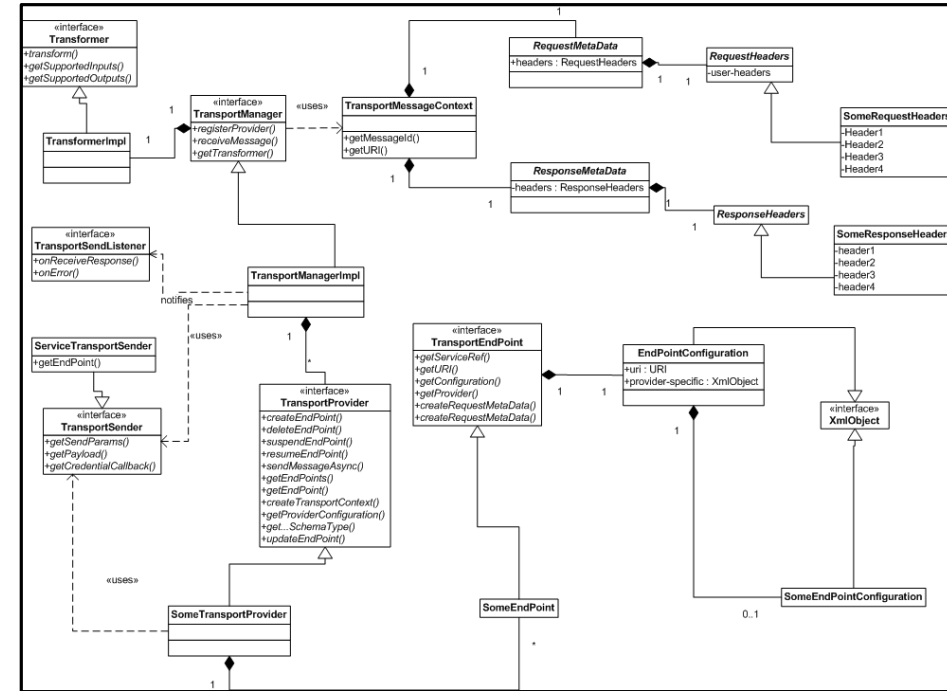
Organigram of HS Limburg



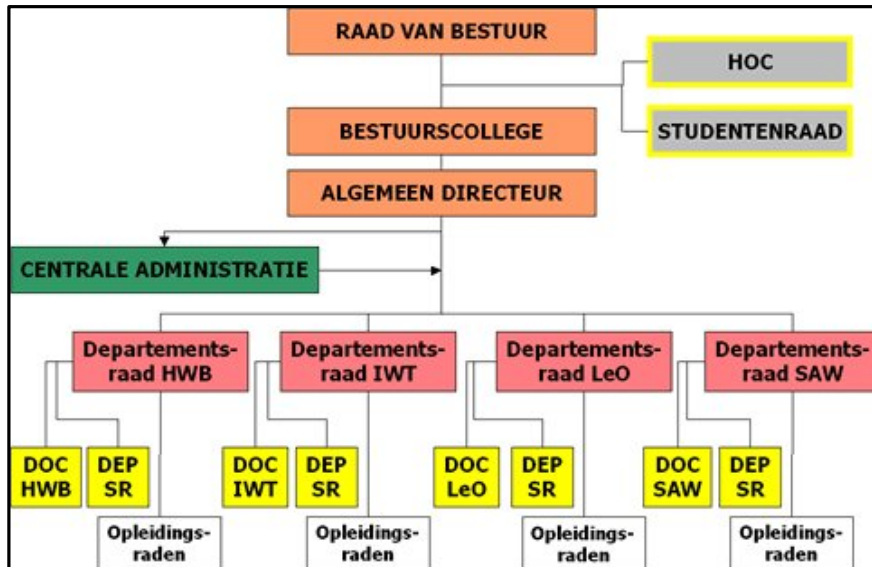
# Orthogonal Layout – Applications



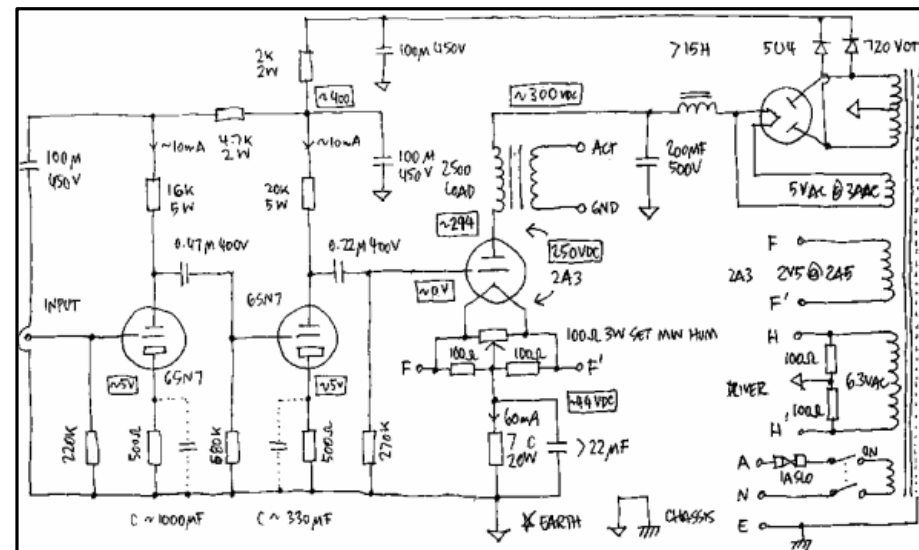
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UML diagram by Oracle



Organigram of HS Limburg



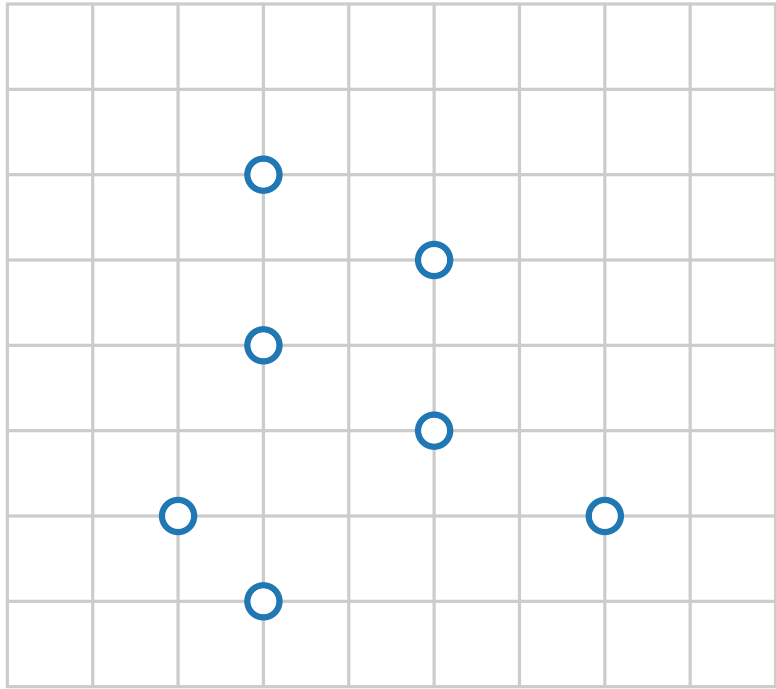
Circuit diagram by Jeff Atwood

# Orthogonal Layout – Definition

**Definition.**

A drawing  $\Gamma$  of a graph  $G = (V, E)$  is called **orthogonal** if

# Orthogonal Layout – Definition

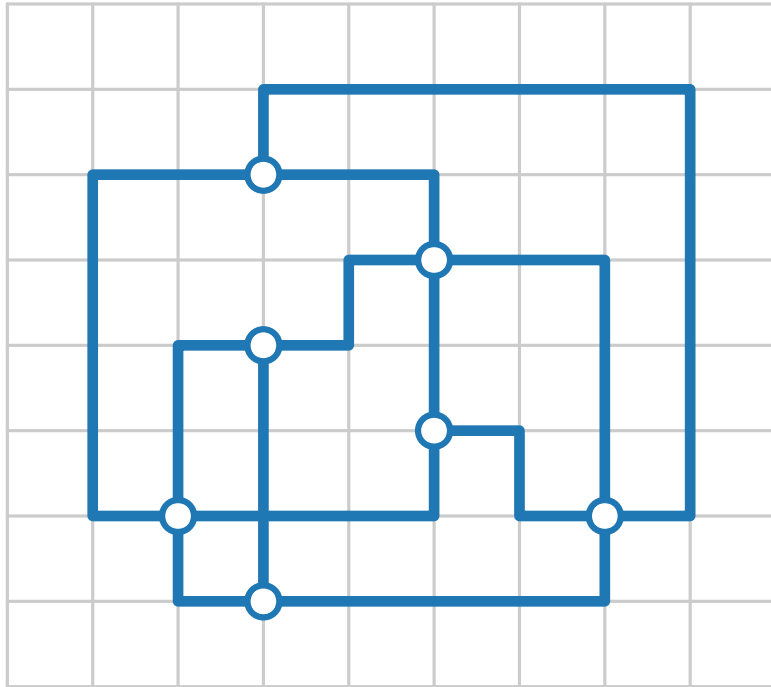


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# Orthogonal Layout – Definition



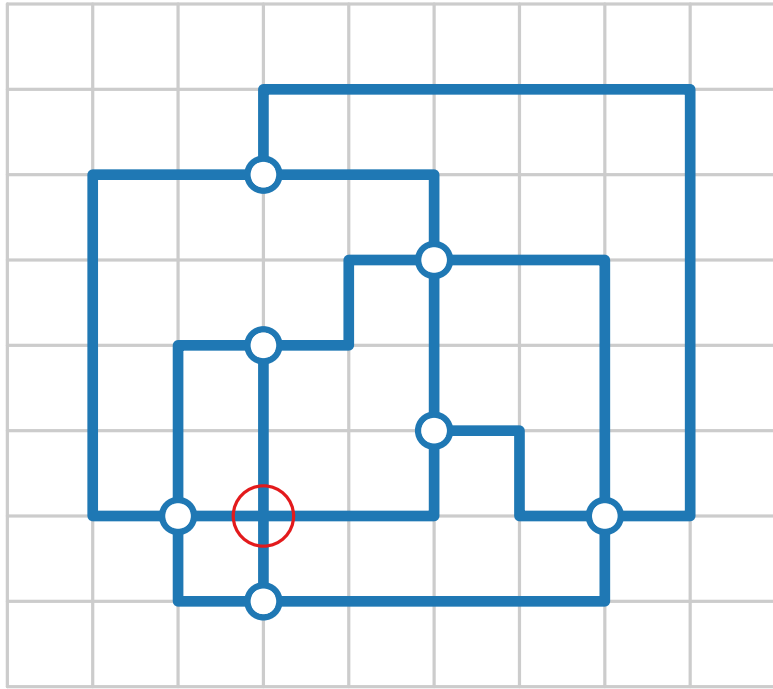
## Definition.

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# Orthogonal Layout – Definition



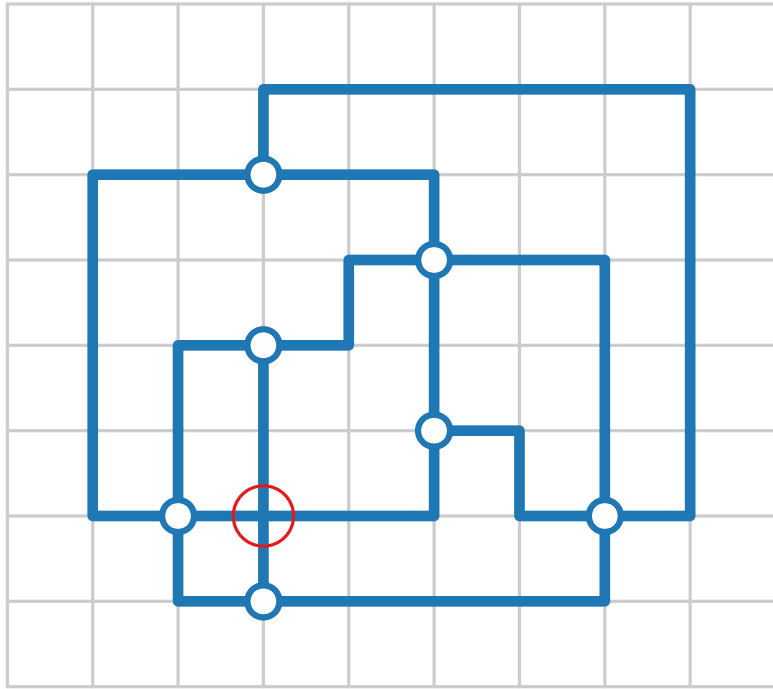
**Observations.**

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# Orthogonal Layout – Definition



## Definition.

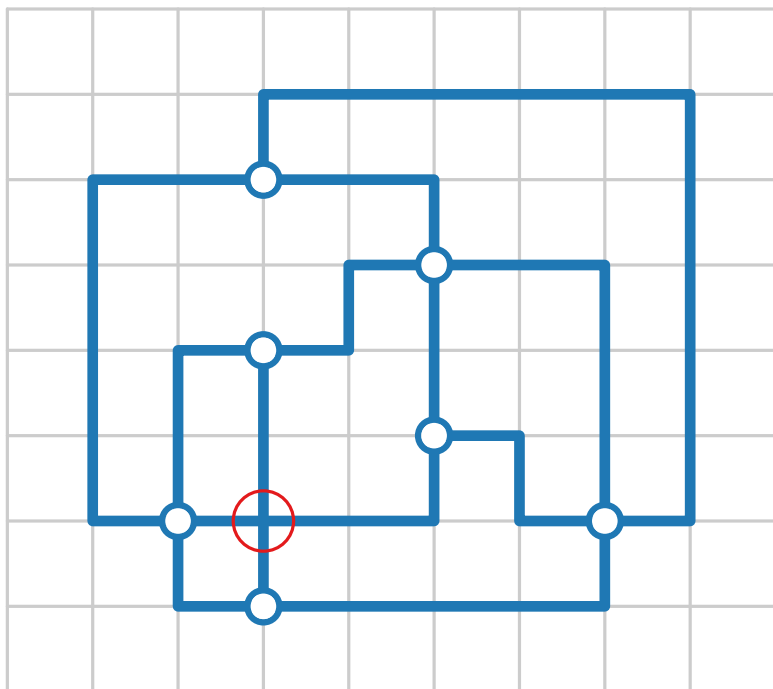
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## Observations.

- Edges lie on grid  $\Rightarrow$   
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# Orthogonal Layout – Definition



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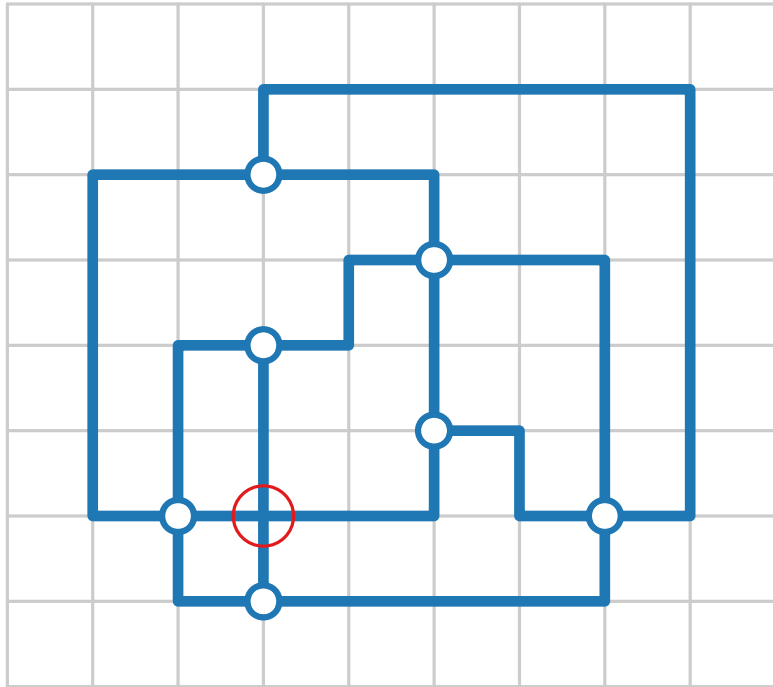
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## Observations.

- Edges lie on grid  $\Rightarrow$  **bends** lie on grid points
- Max degree of each vertex is at most 4



# Orthogonal Layout – Definition

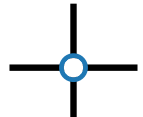


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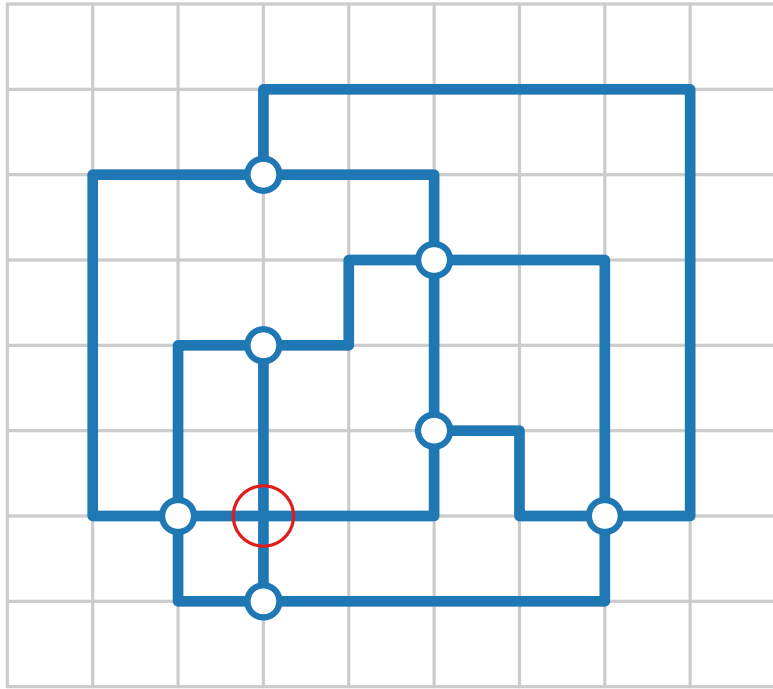
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## Observations.

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# Orthogonal Layout – Definition



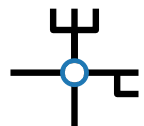
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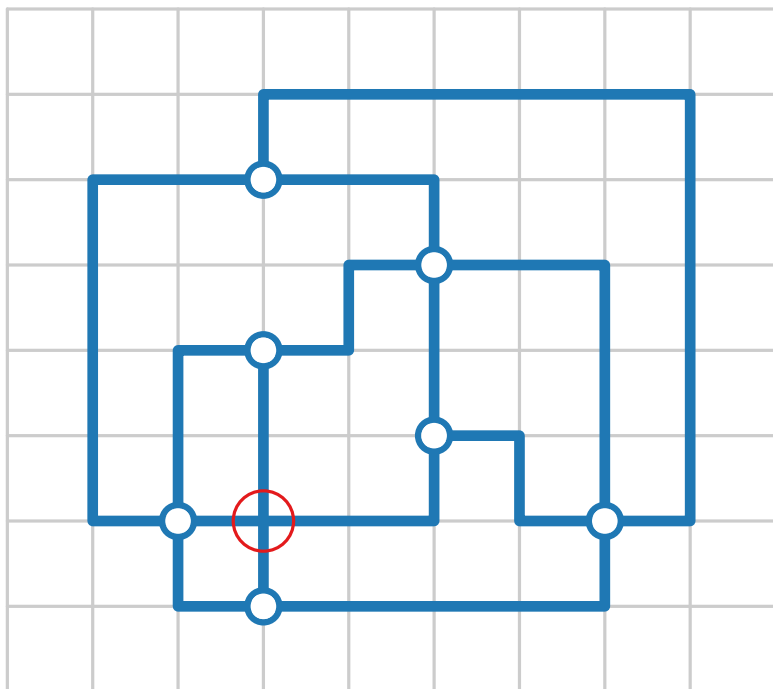
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# Orthogonal Layout – Definition



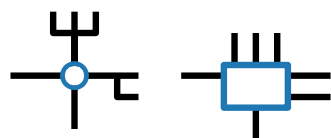
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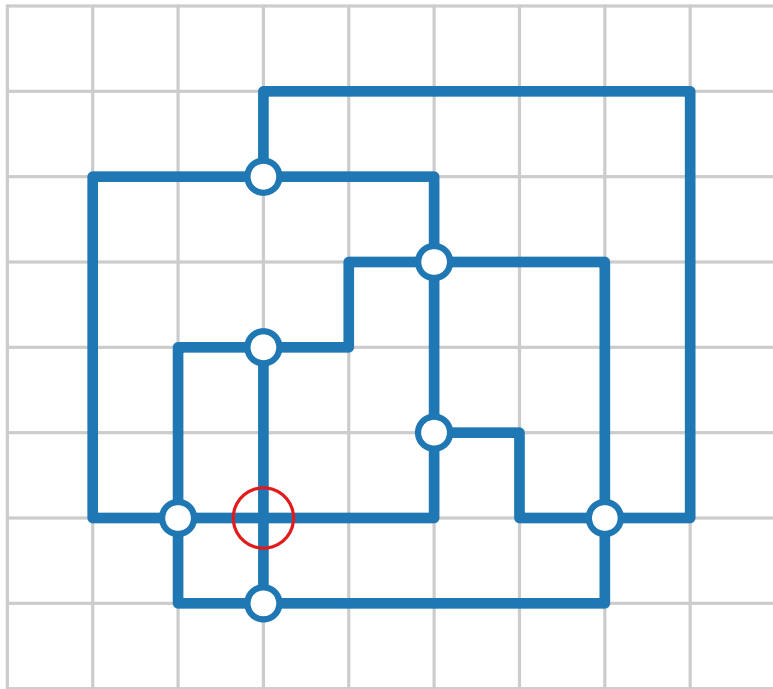
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# Orthogonal Layout – Definition



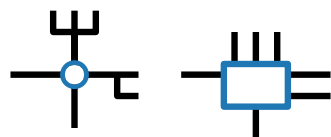
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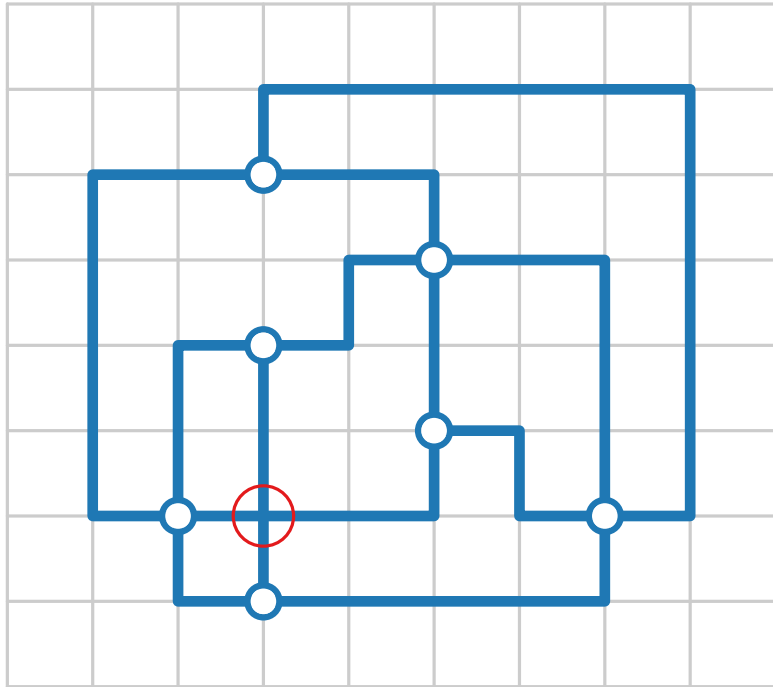
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## Planarization.

# Orthogonal Layout – Definition



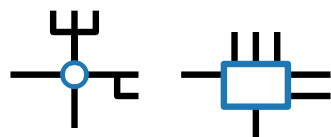
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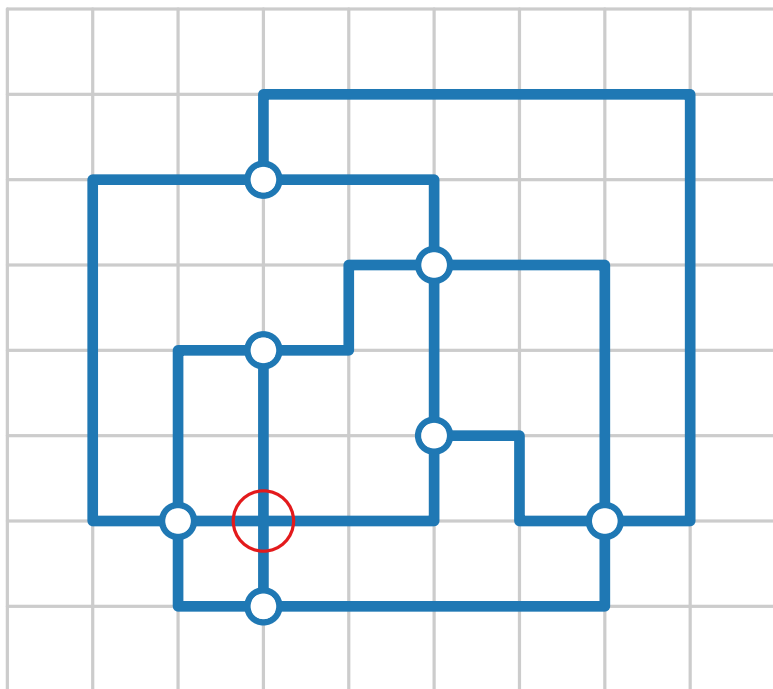
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## Planarization.

- Fix embedding

# Orthogonal Layout – Definition



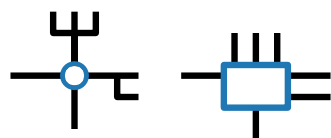
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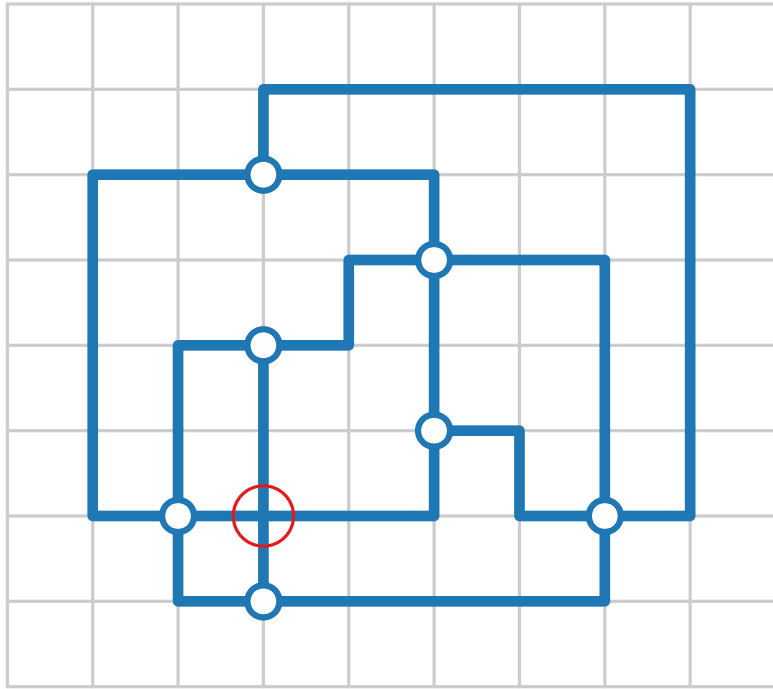
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## Planarization.

- Fix embedding
- Crossings become vertices

# Orthogonal Layout – Definition



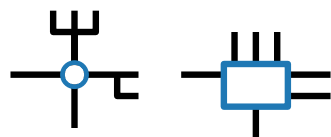
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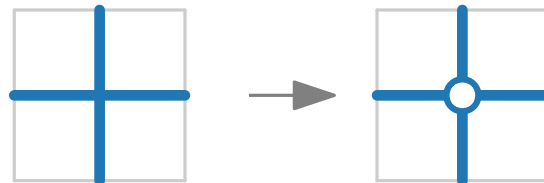
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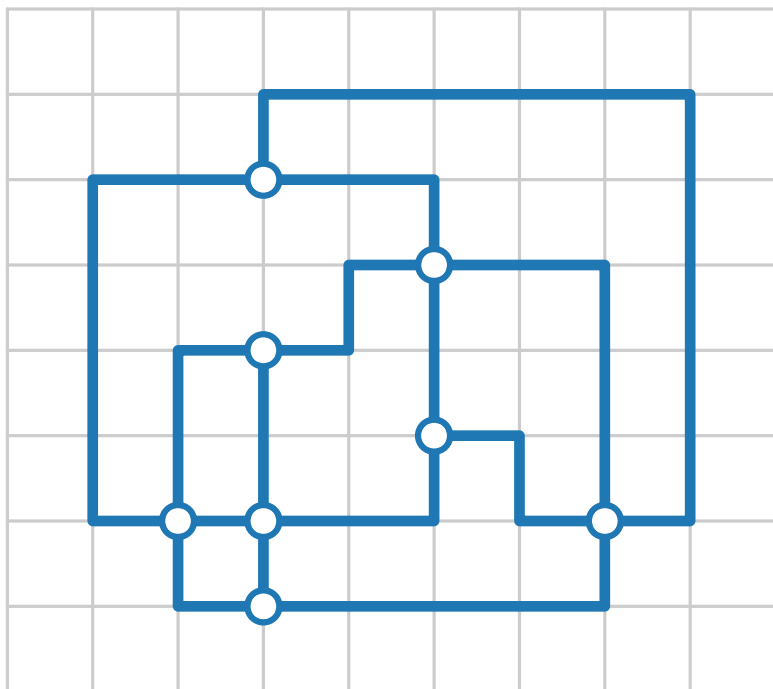


## Planarization.

- Fix embedding
- Crossings become vertices



# Orthogonal Layout – Definition



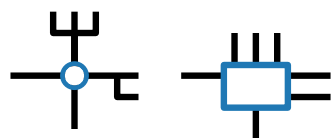
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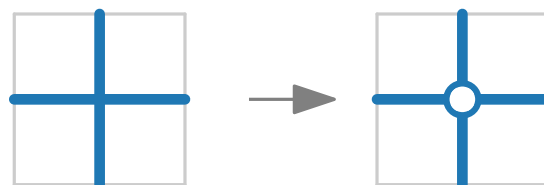
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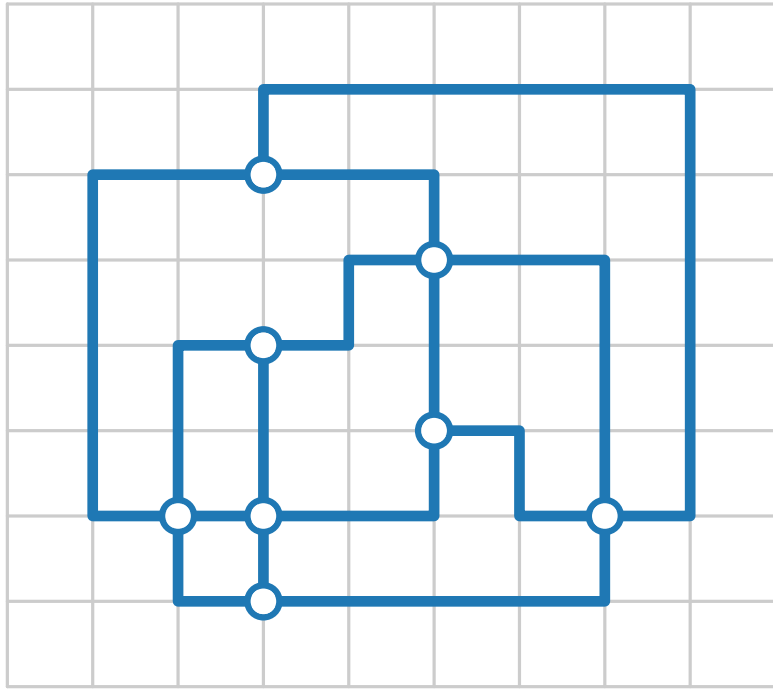
## Planarization.

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# Orthogonal Layout – Definition



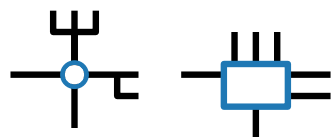
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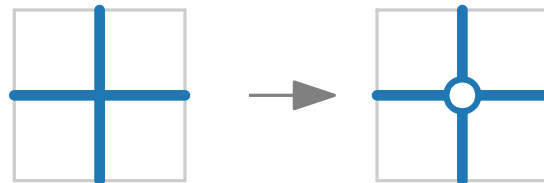
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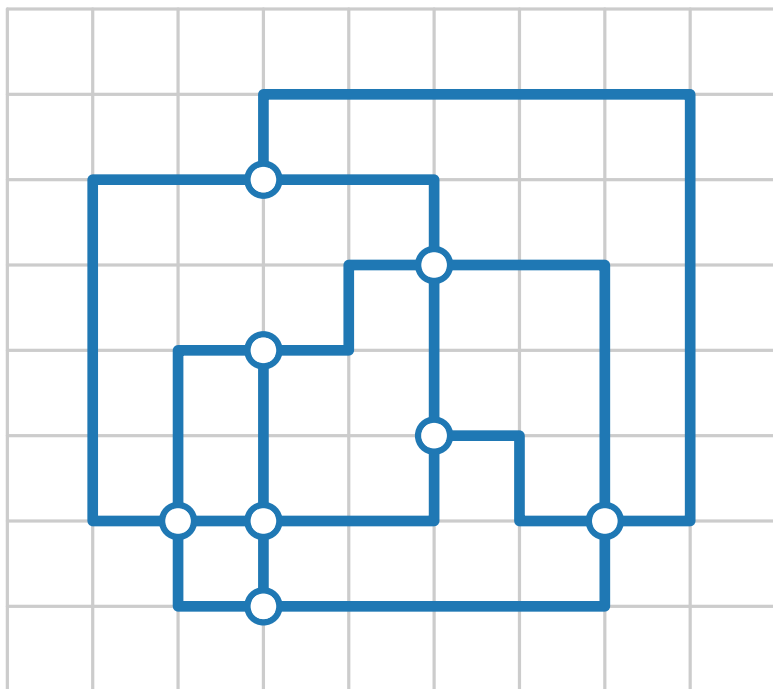
## Planarization.

- Fix embedding
- Crossings become vertices



## Aesthetic criteria.

# Orthogonal Layout – Definition



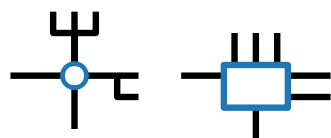
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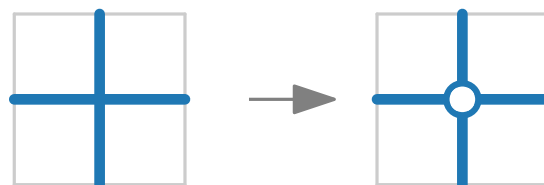
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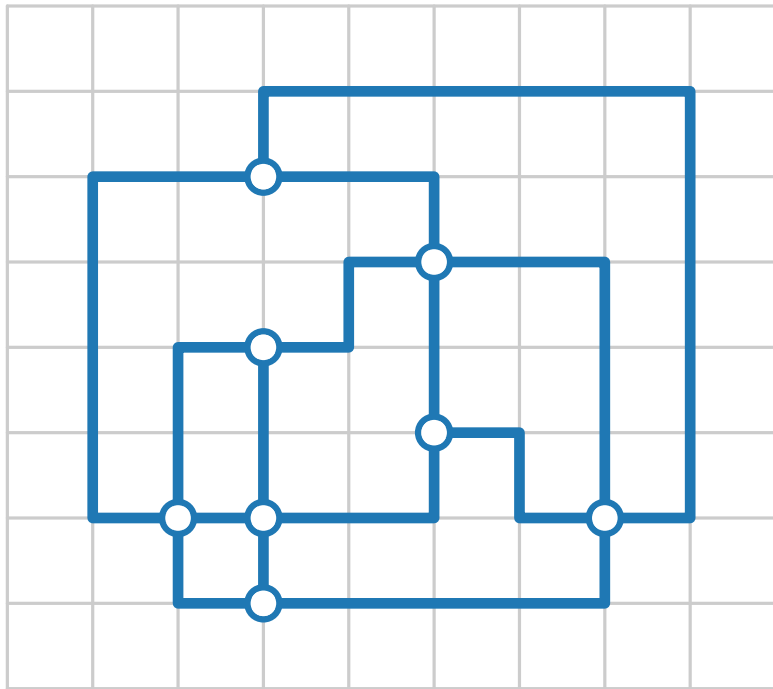
- Fix embedding
- Crossings become vertices



## Aesthetic criteria.

- Number of bends

# Orthogonal Layout – Definition



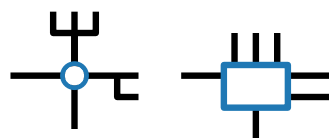
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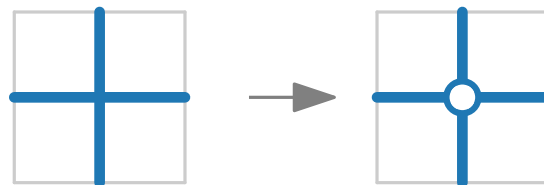
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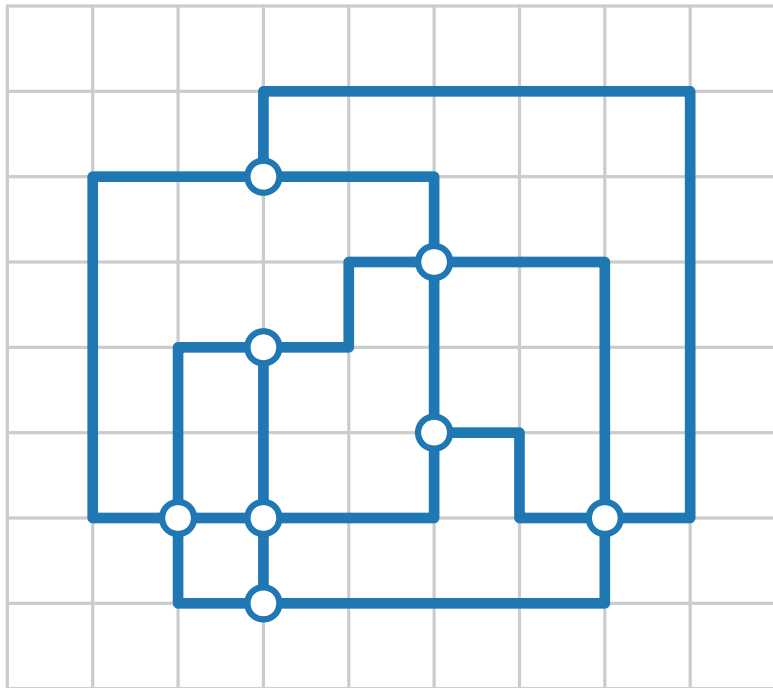
- Fix embedding
- Crossings become vertices



## Aesthetic criteria.

- Number of bends
- Length of edges

# Orthogonal Layout – Definition



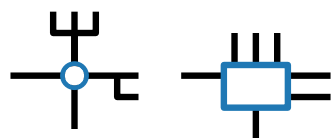
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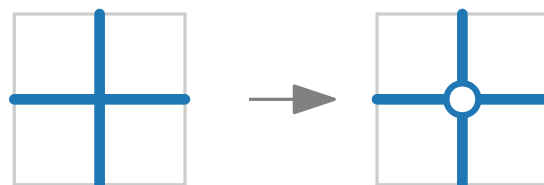
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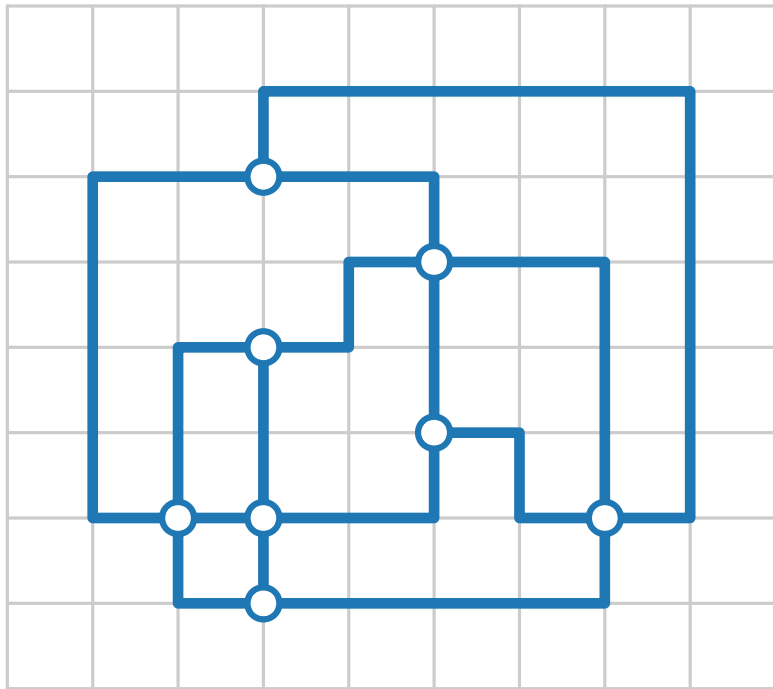
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## Aesthetic criteria.

- Number of bends
- Length of edges
- Width, height, area

# Orthogonal Layout – Definition



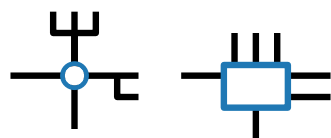
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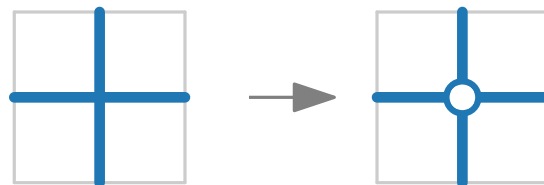
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## Planarization.

- Fix embedding
- Crossings become vertices



## Aesthetic criteria.

- Number of bends
- Length of edges
- Width, height, area
- Monotonicity of edges
- ...

# Topology – Shape – Metrics

Three-step approach:

[Tamassia 1987]

TOPOLOGY

—

SHAPE

—

METRICS

# Topology – Shape – Metrics

Three-step approach:

[Tamassia 1987]

$$V = \{v_1, v_2, v_3, v_4\}$$

$$E = \{v_1v_2, v_1v_3, v_1v_4, v_2v_3, v_2v_4\}$$

TOPOLOGY

—

SHAPE

—

METRICS

# Topology – Shape – Metrics

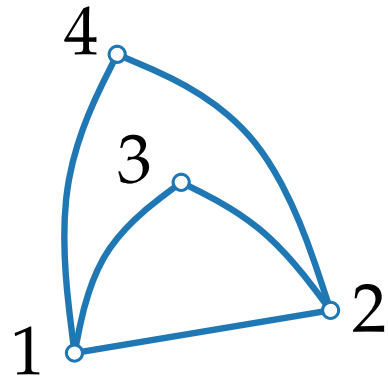
Three-step approach:

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$$V = \{v_1, v_2, v_3, v_4\}$$

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combinatorial  
embedding/  
planarization



TOPOLOGY

—

SHAPE

—

METRICS



# Topology – Shape – Metrics

Three-step approach:

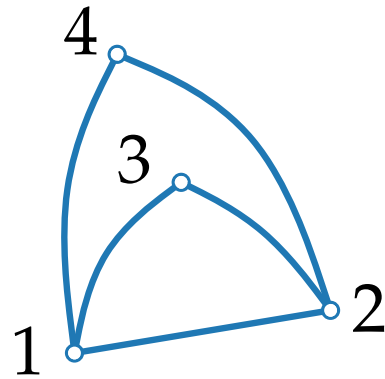
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reduce  
crossings

combinatorial  
embedding/  
planarization



TOPOLOGY

—

SHAPE

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METRICS

# Topology – Shape – Metrics

Three-step approach:

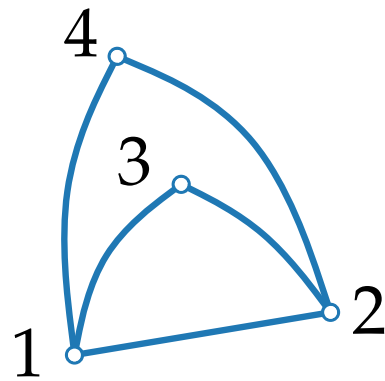
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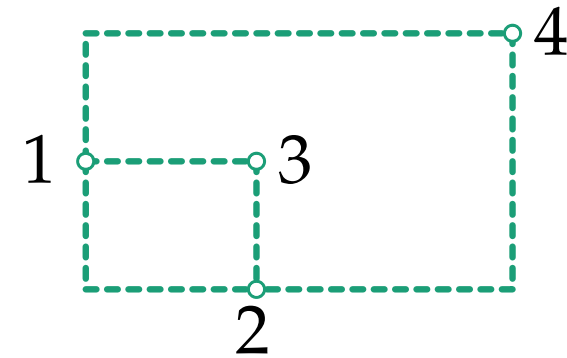
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reduce  
crossings

combinatorial  
embedding/  
planarization



orthogonal  
representation



TOPOLOGY

—

SHAPE

—

METRICS

# Topology – Shape – Metrics

Three-step approach:

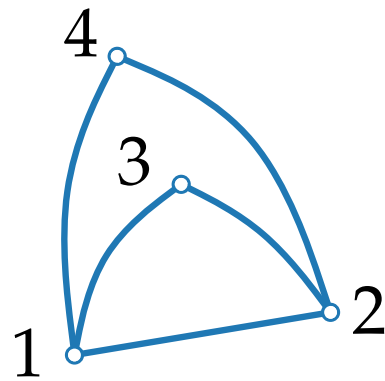
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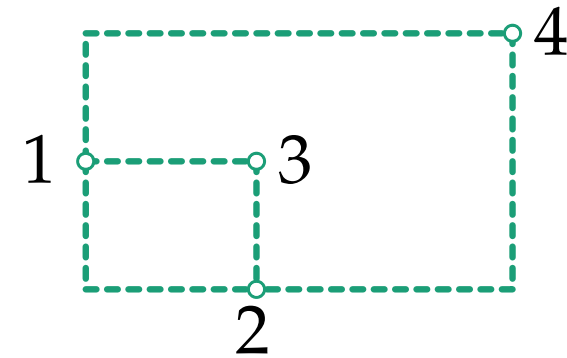
reduce  
crossings

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bend minimization

orthogonal  
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TOPOLOGY

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# Topology – Shape – Metrics

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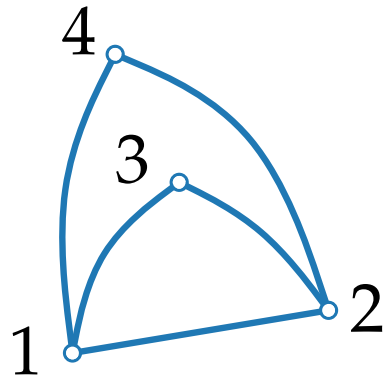
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$$V = \{v_1, v_2, v_3, v_4\}$$

$$E = \{v_1v_2, v_1v_3, v_1v_4, v_2v_3, v_2v_4\}$$

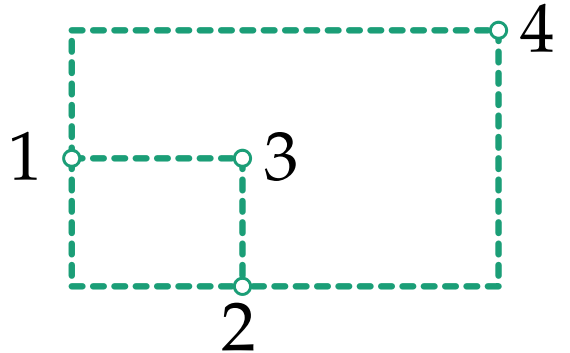
reduce crossings

combinatorial embedding/  
planarization

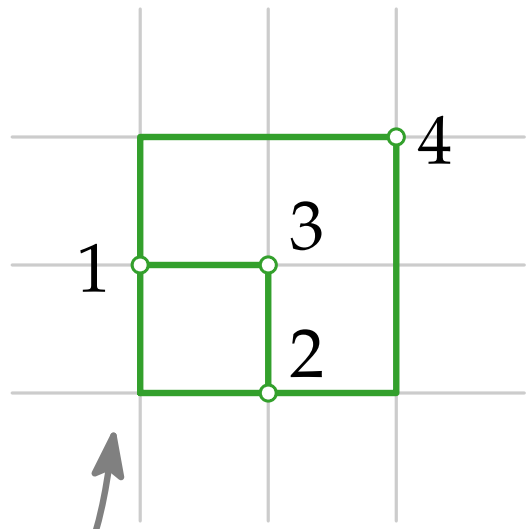


bend minimization

orthogonal representation



planar orthogonal drawing



TOPOLOGY

—

SHAPE

—

METRICS

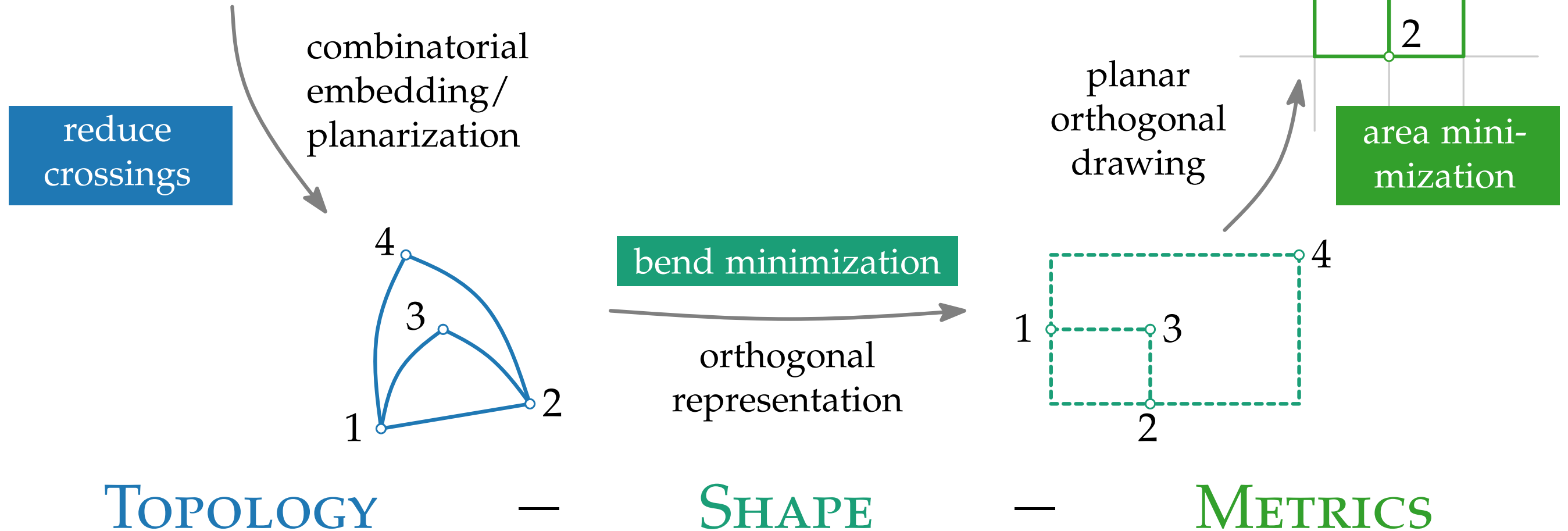
# Topology – Shape – Metrics

Three-step approach:

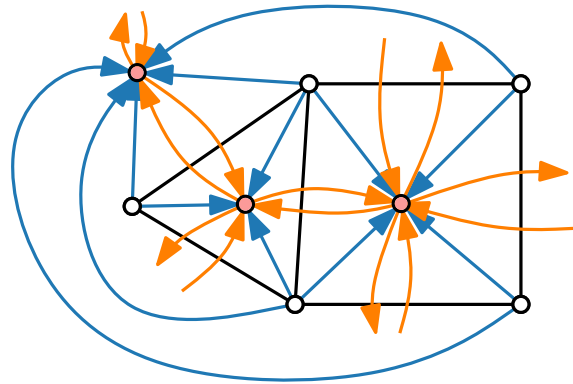
[Tamassia 1987]

$$V = \{v_1, v_2, v_3, v_4\}$$

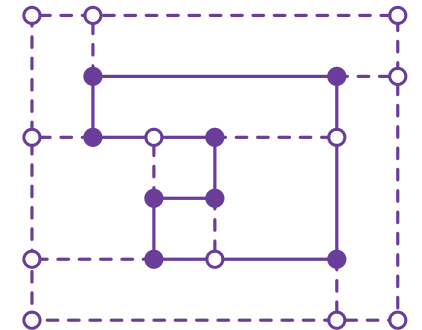
$$E = \{v_1v_2, v_1v_3, v_1v_4, v_2v_3, v_2v_4\}$$



# Visualization of Graphs

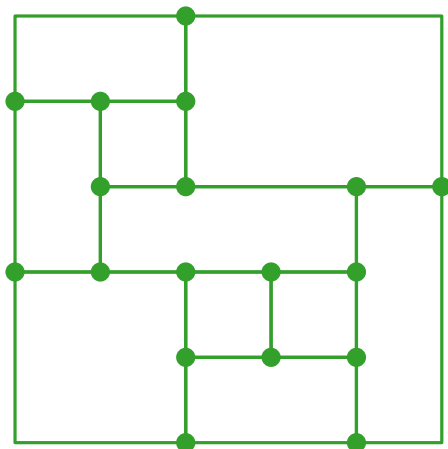


## Lecture 6: Orthogonal Layouts



## Part II: Orthogonal Representation

Philipp Kindermann



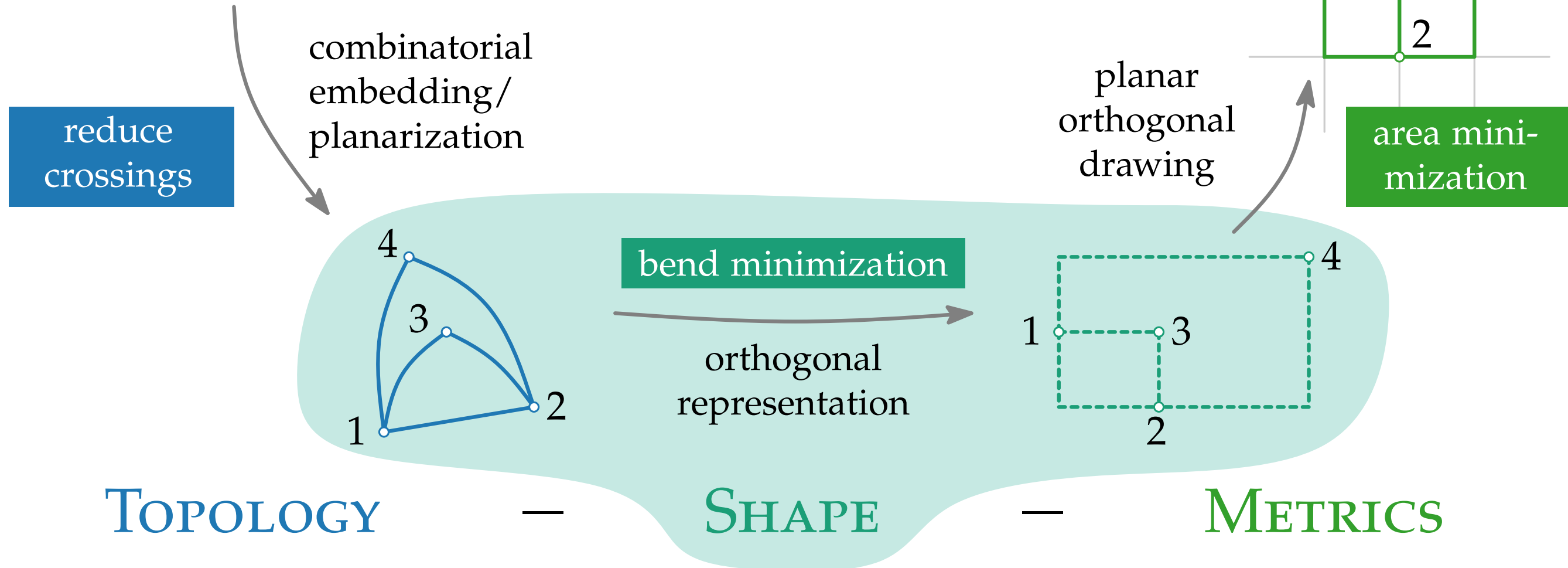
# Topology – Shape – Metrics

Three-step approach:

[Tamassia 1987]

$$V = \{v_1, v_2, v_3, v_4\}$$

$$E = \{v_1v_2, v_1v_3, v_1v_4, v_2v_3, v_2v_4\}$$



# Orthogonal Representation

## Idea.

Describe orthogonal drawing combinatorically.



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Describe orthogonal drawing combinatorically.

## Definitions.

Let  $G = (V, E)$  be a plane graph with faces  $F$  and outer face  $f_0$ .

# Orthogonal Representation

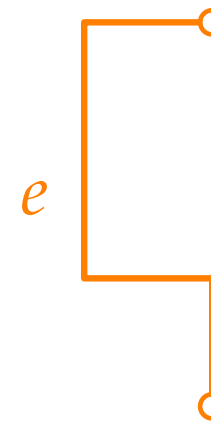
## Idea.

Describe orthogonal drawing combinatorially.

## Definitions.

Let  $G = (V, E)$  be a plane graph with faces  $F$  and outer face  $f_0$ .

- Let  $e$  be an edge



# Orthogonal Representation

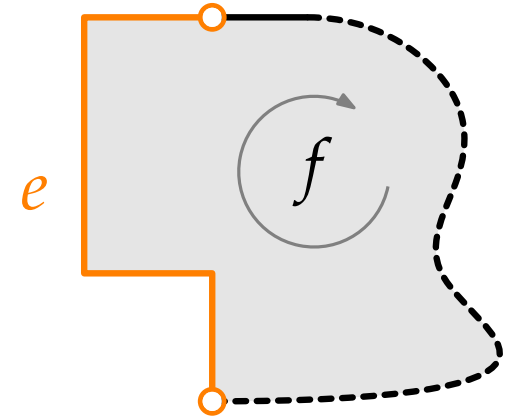
## Idea.

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# Orthogonal Representation

## Idea.

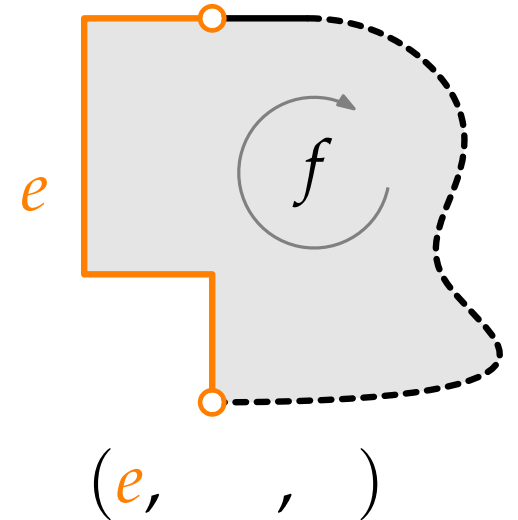
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An **edge description** of  $e$  wrt  $f$  is a triple  $(e, \delta, \alpha)$  where



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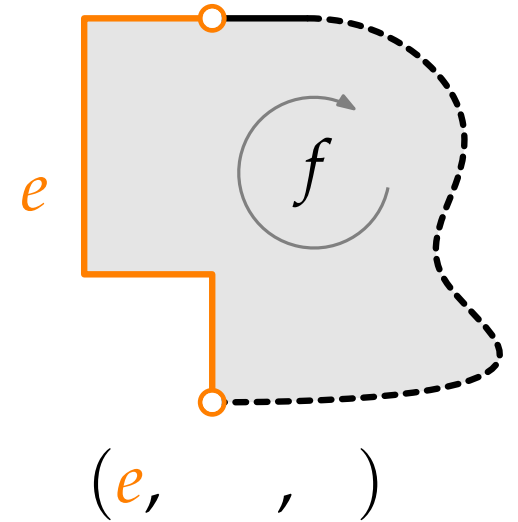
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    - $\delta$  is a sequence of  $\{0, 1\}^*$  (0 = right bend, 1 = left bend)



# Orthogonal Representation

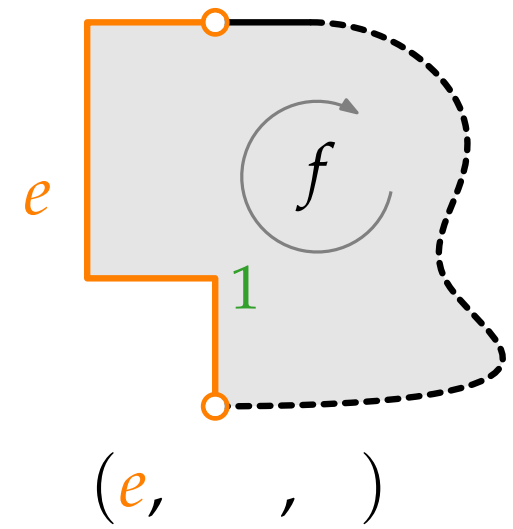
## Idea.

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# Orthogonal Representation

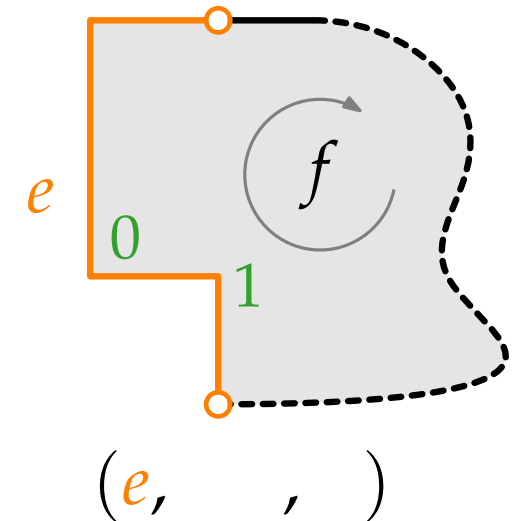
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# Orthogonal Representation

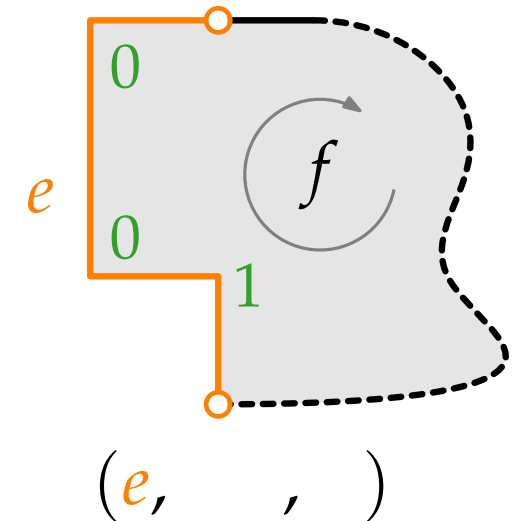
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# Orthogonal Representation

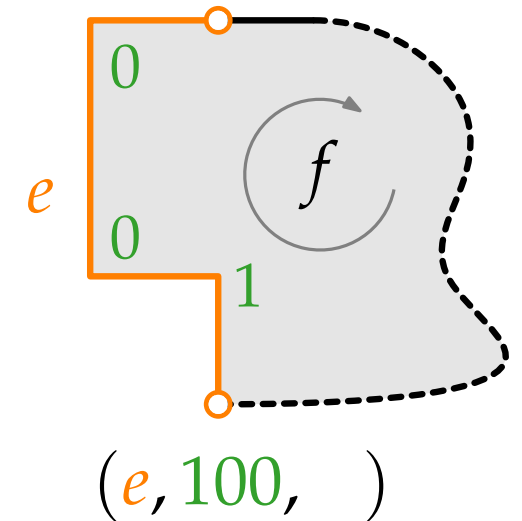
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# Orthogonal Representation

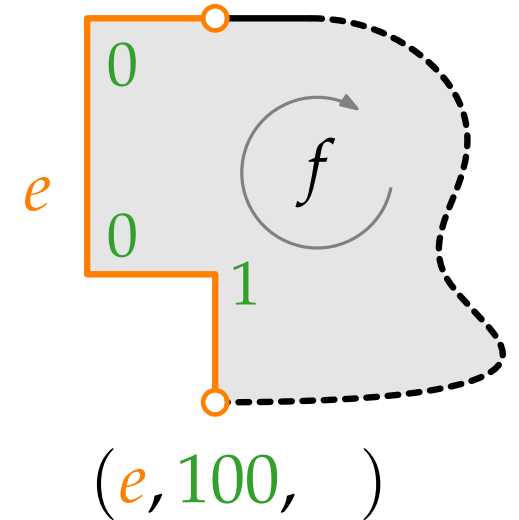
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    - $\alpha$  is angle  $\in \{\frac{\pi}{2}, \pi, \frac{3\pi}{2}, 2\pi\}$  between  $e$  and next edge  $e'$



# Orthogonal Representation

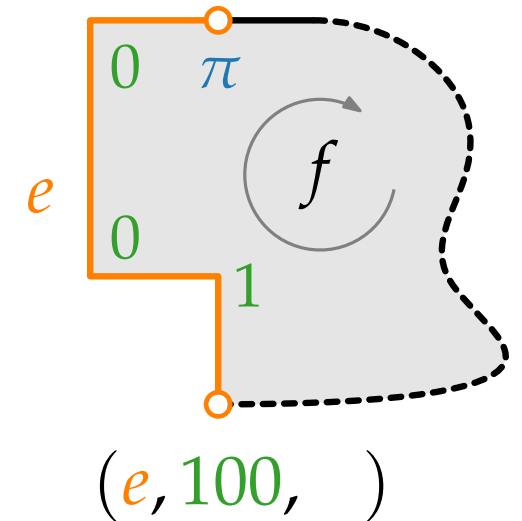
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# Orthogonal Representation

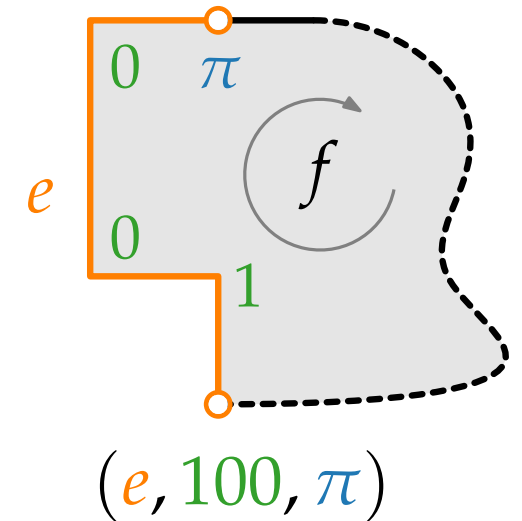
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Describe orthogonal drawing combinatorially.

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# Orthogonal Representation

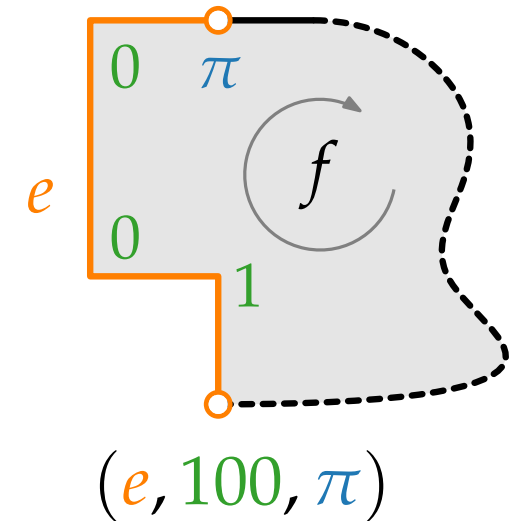
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- A **face representation**  $H(f)$  of  $f$  is a clockwise ordered sequence of edge descriptions  $(e, \delta, \alpha)$ .



# Orthogonal Representation

## Idea.

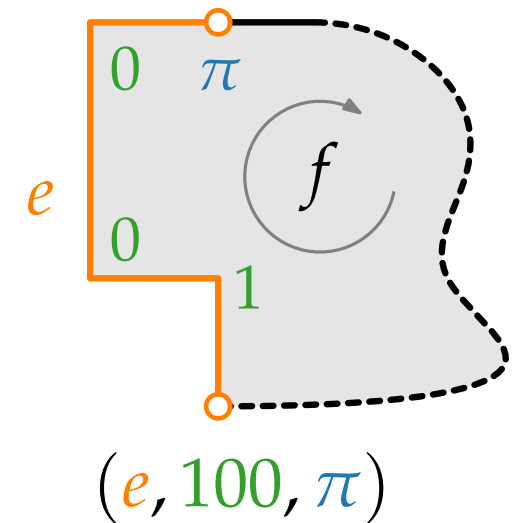
Describe orthogonal drawing combinatorially.

## Definitions.

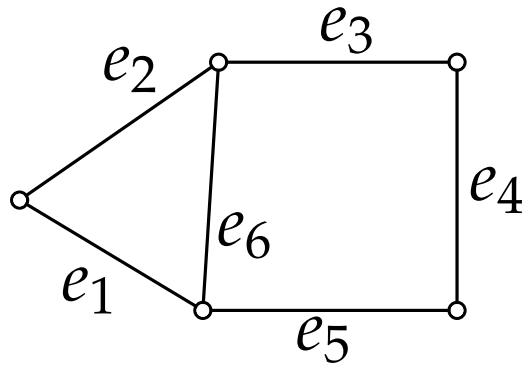
Let  $G = (V, E)$  be a plane graph with faces  $F$  and outer face  $f_0$ .

- Let  $e$  be an edge with the face  $f$  to the right.
  - An **edge description** of  $e$  wrt  $f$  is a triple  $(e, \delta, \alpha)$  where
    - $\delta$  is a sequence of  $\{0, 1\}^*$  (0 = right bend, 1 = left bend)
    - $\alpha$  is angle  $\in \{\frac{\pi}{2}, \pi, \frac{3\pi}{2}, 2\pi\}$  between  $e$  and next edge  $e'$
- A **face representation**  $H(f)$  of  $f$  is a clockwise ordered sequence of edge descriptions  $(e, \delta, \alpha)$ .
- An **orthogonal representation**  $H(G)$  of  $G$  is defined as

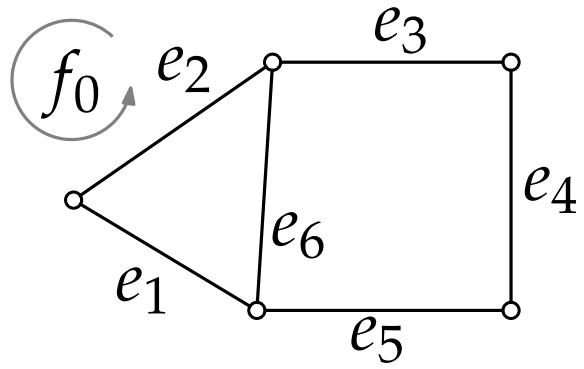
$$H(G) = \{H(f) \mid f \in F\}.$$



# Orthogonal Representation – Example

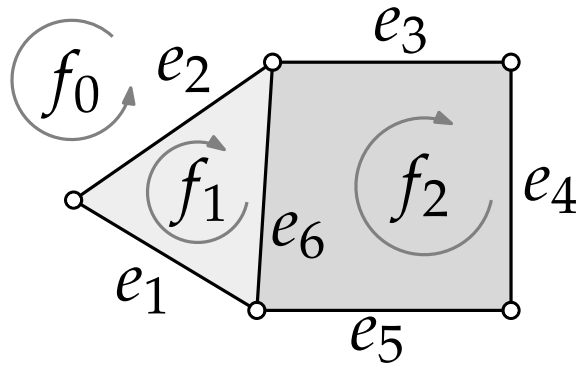


# Orthogonal Representation – Example





# Orthogonal Representation – Example

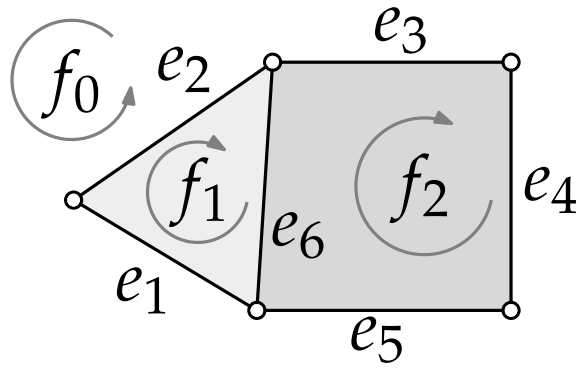


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

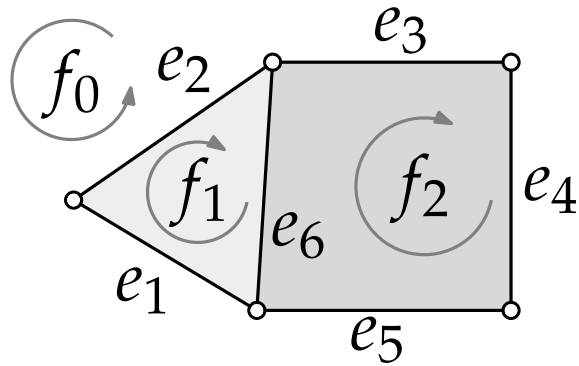


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

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$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$



Combinatorial “drawing” of  $H(G)$ ?

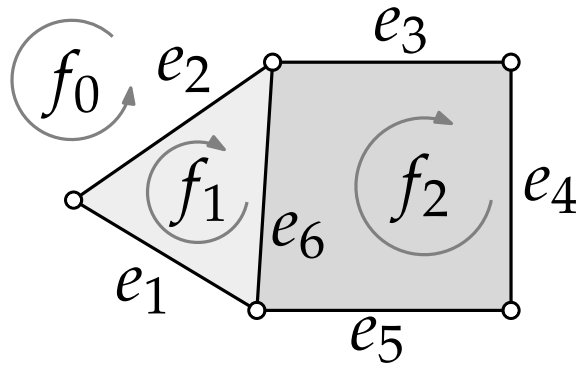
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$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

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$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

$f_0$

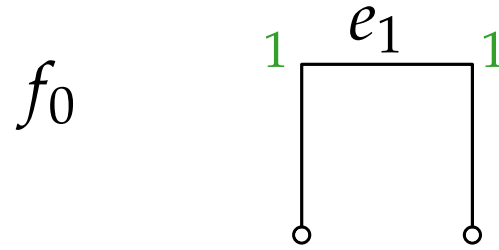
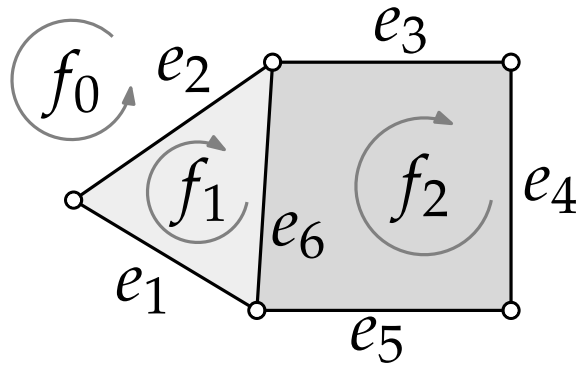


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

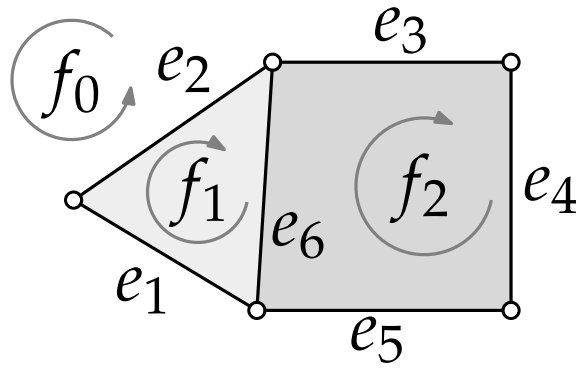


# Orthogonal Representation – Example

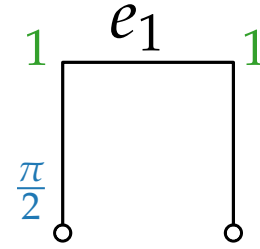
$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

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$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$



$f_0$

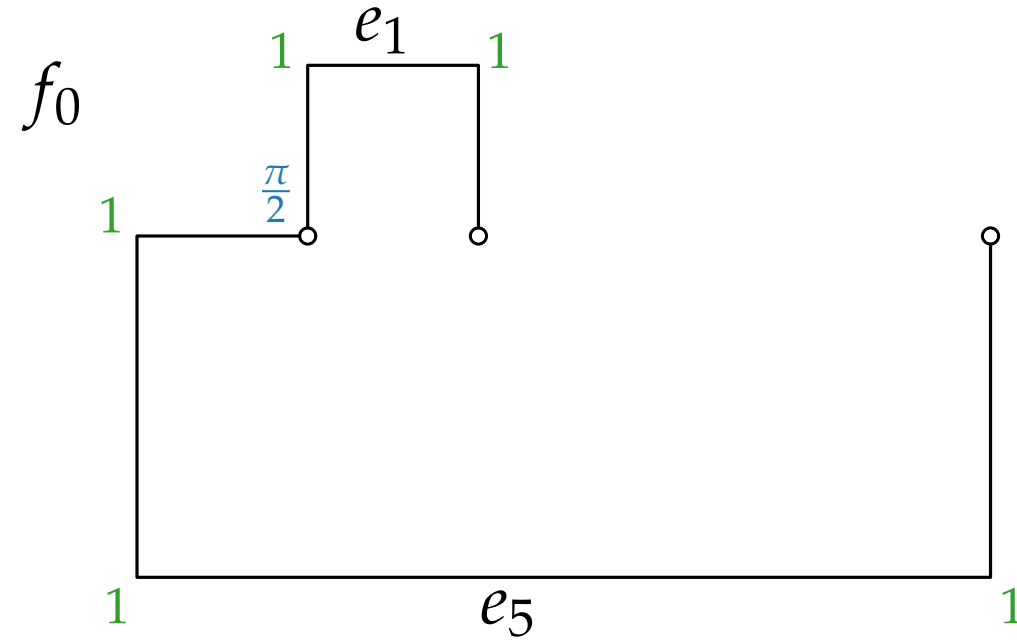
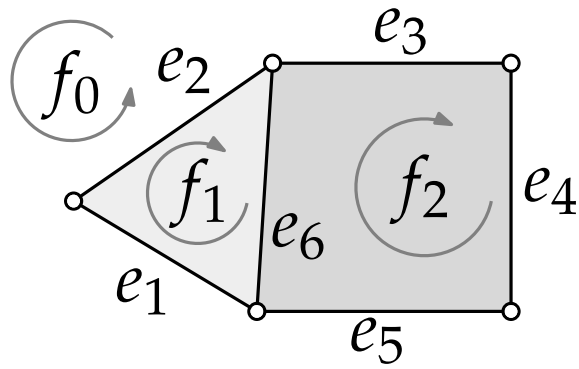


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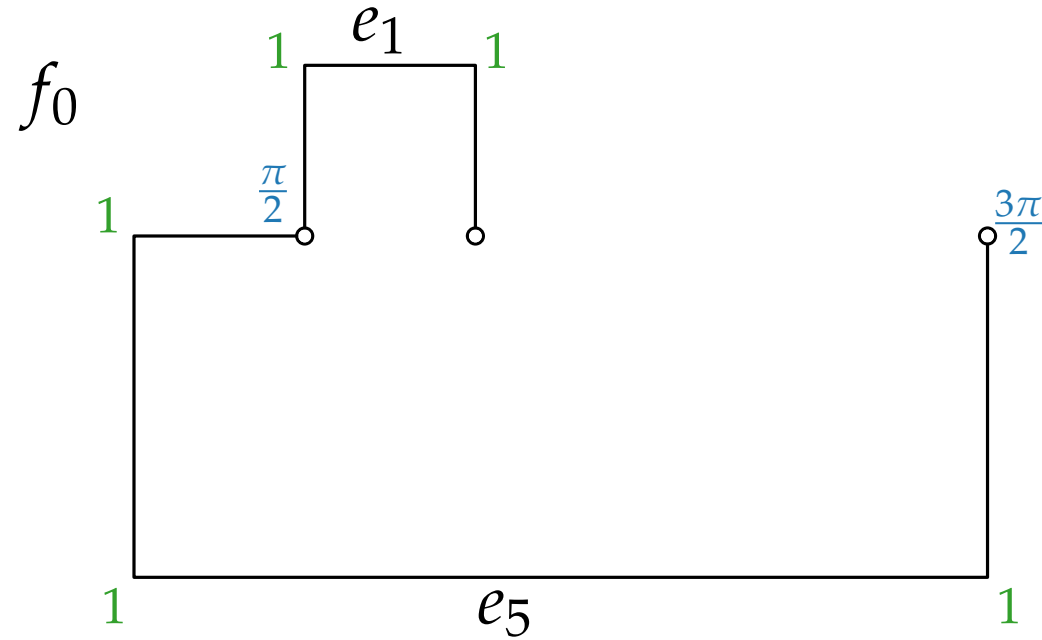
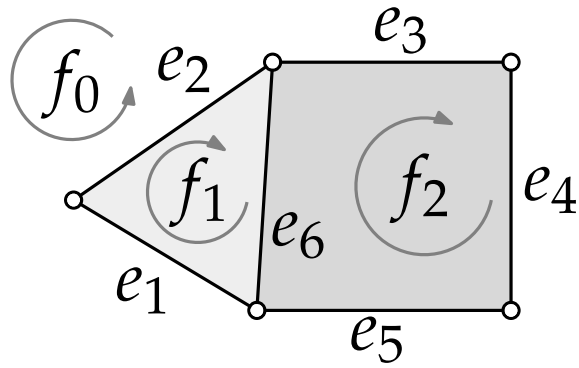


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$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

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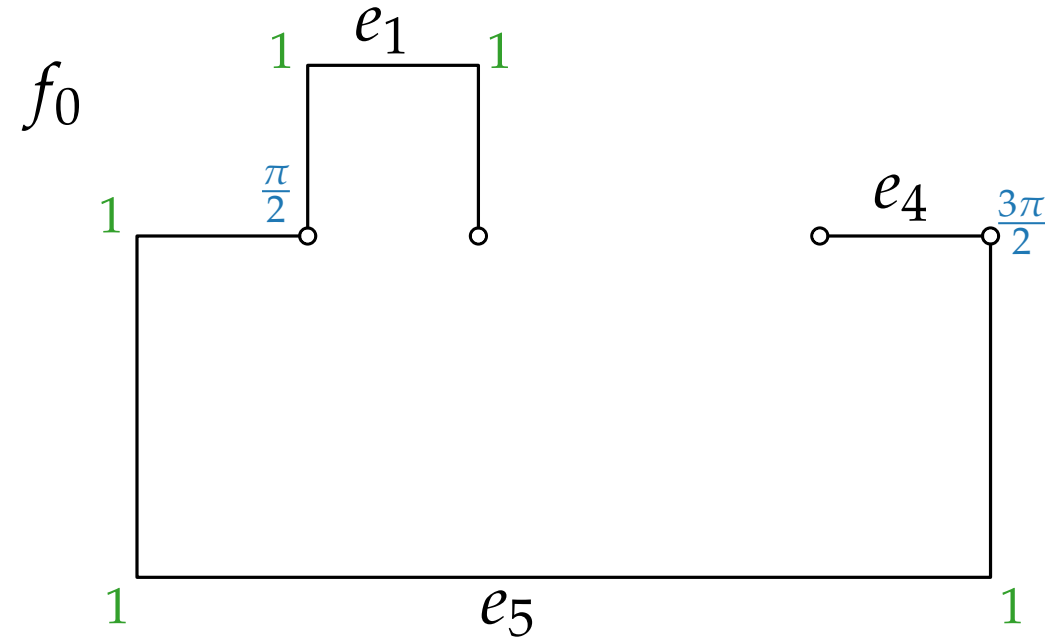
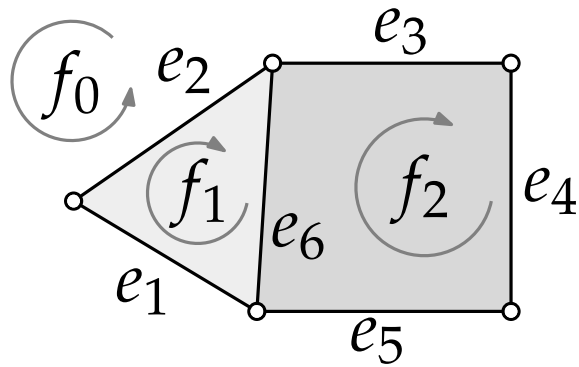


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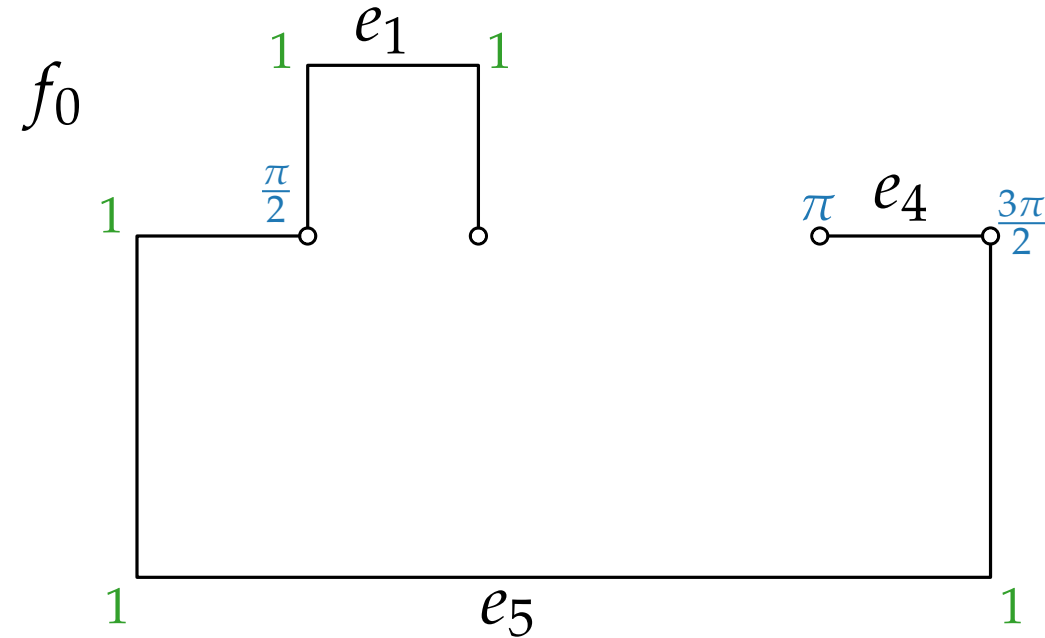
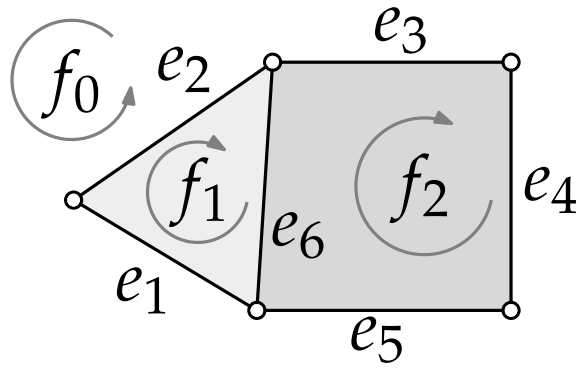


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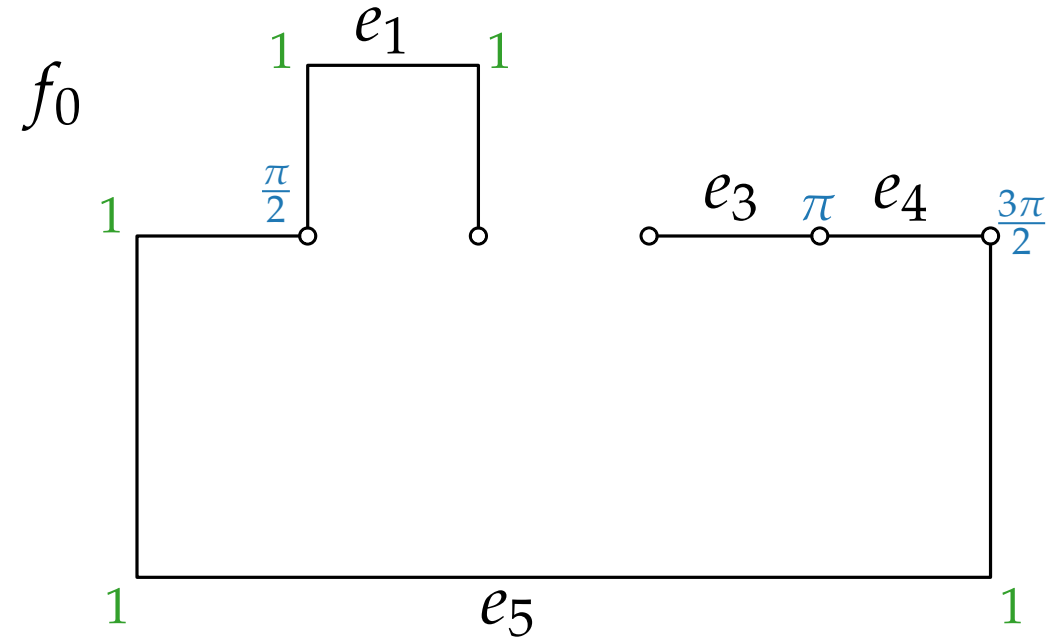
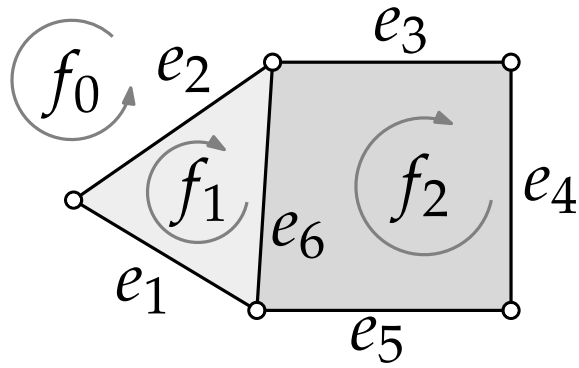


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$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

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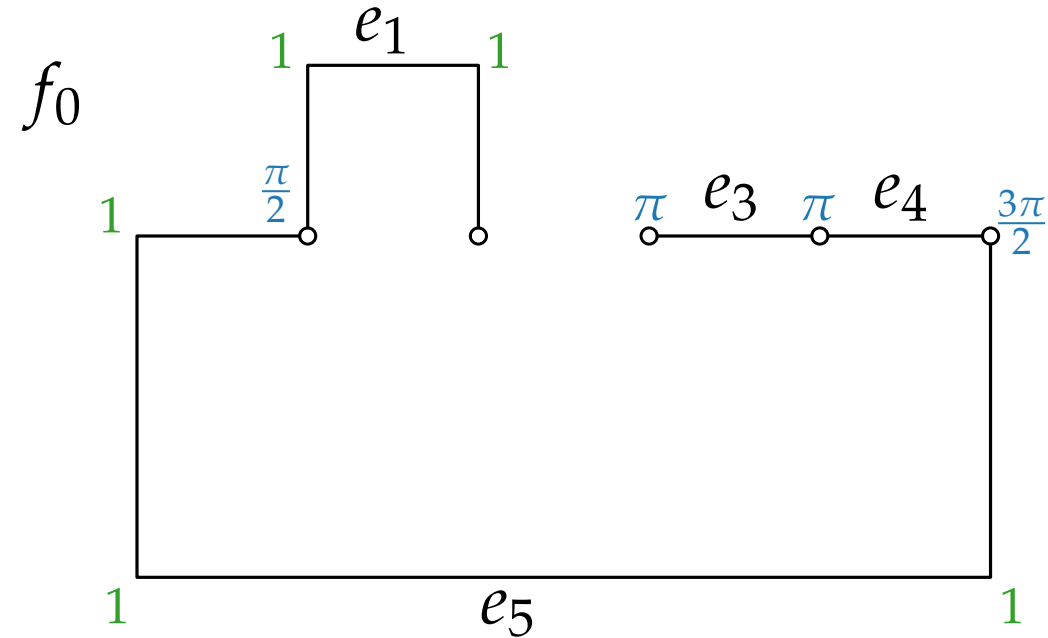
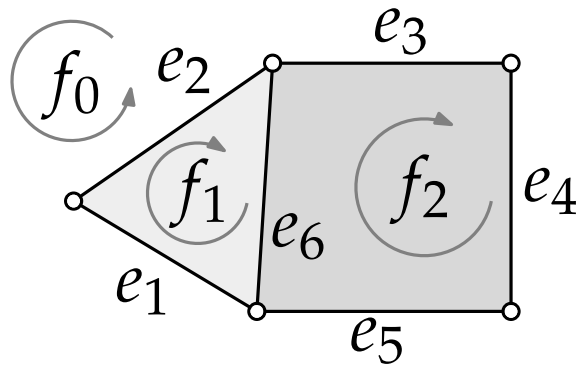


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

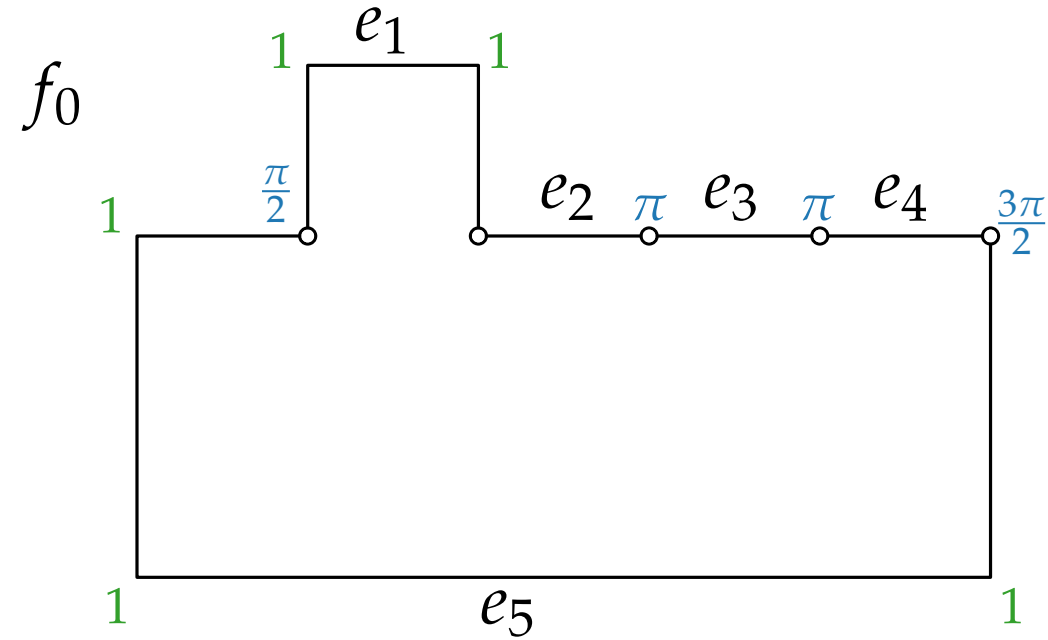
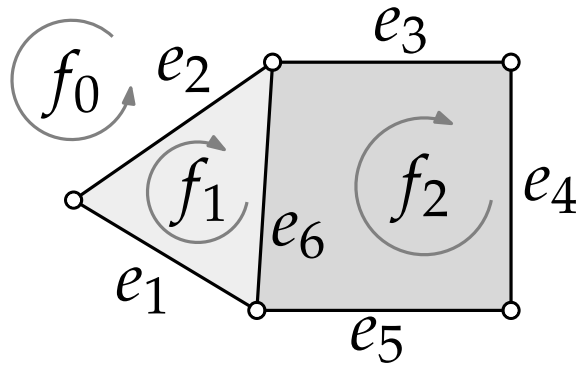


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

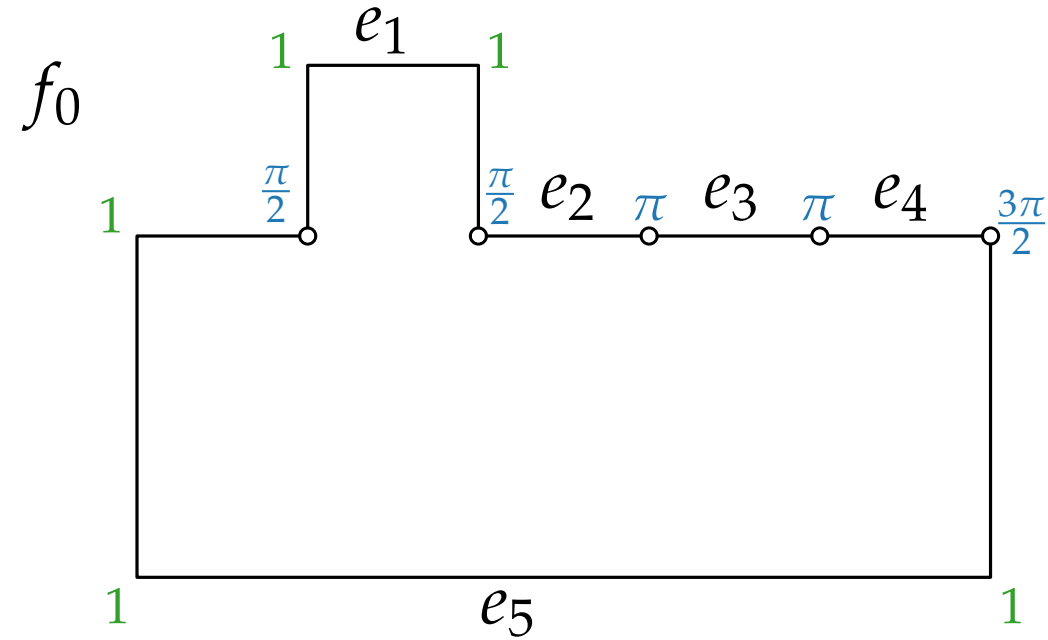
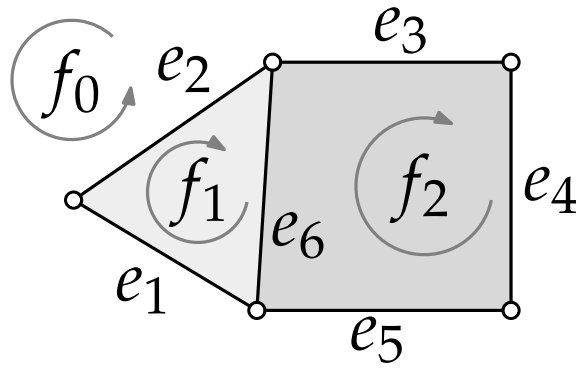


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

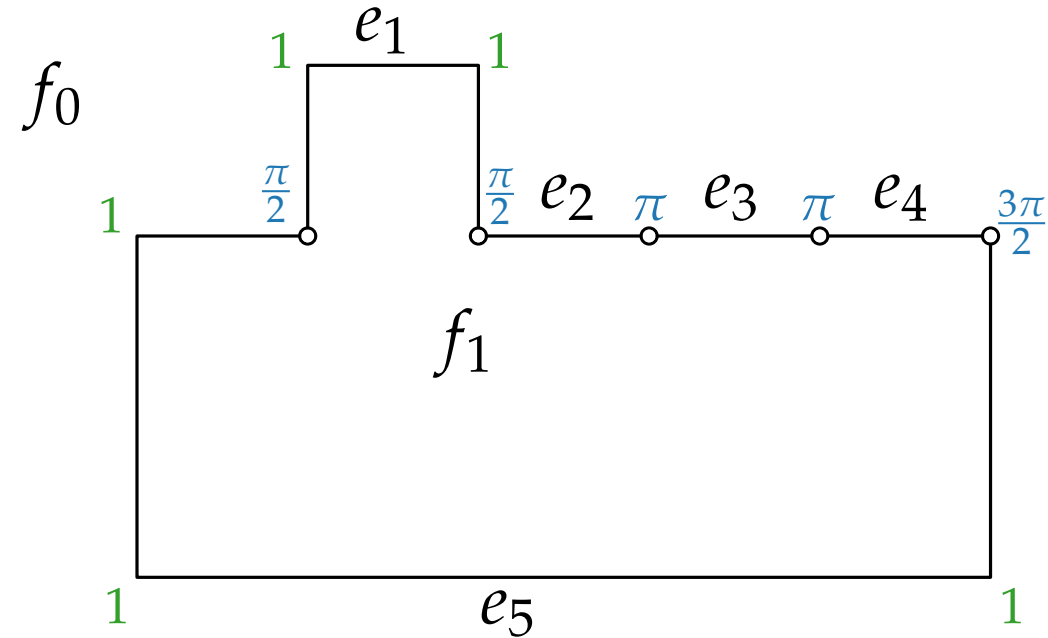
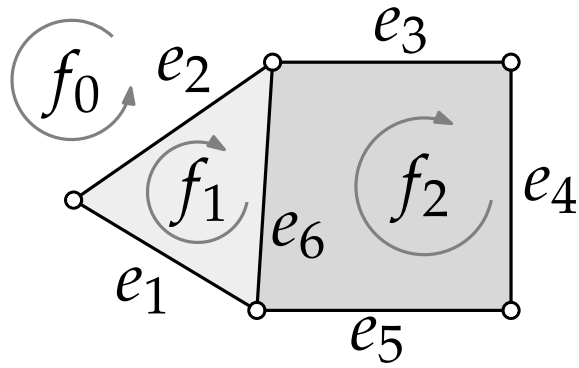


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

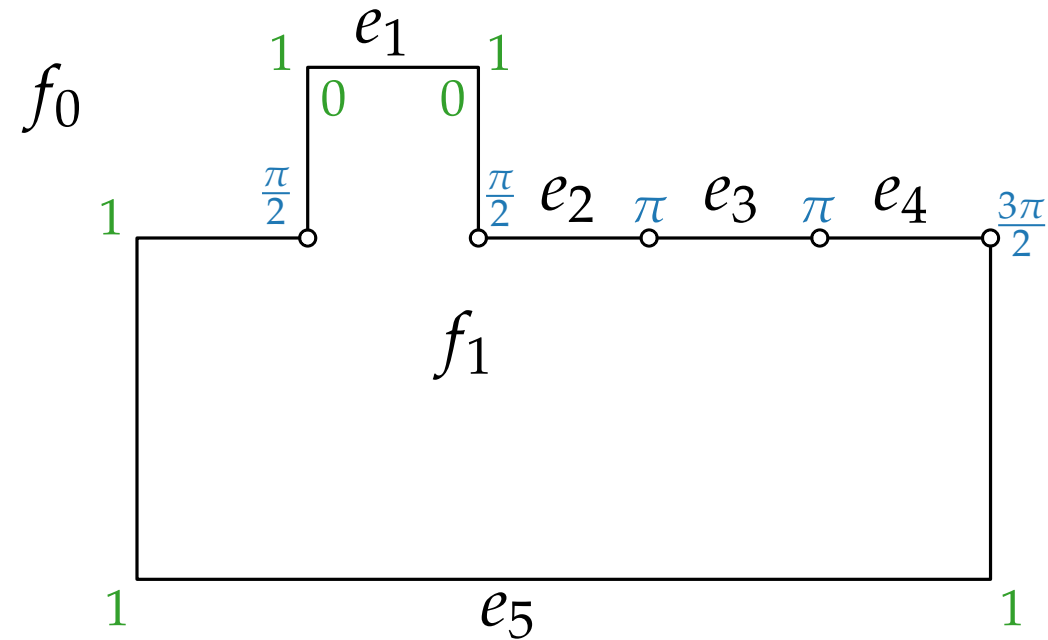
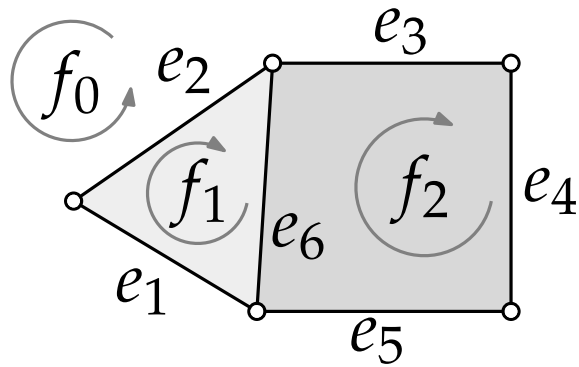


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$



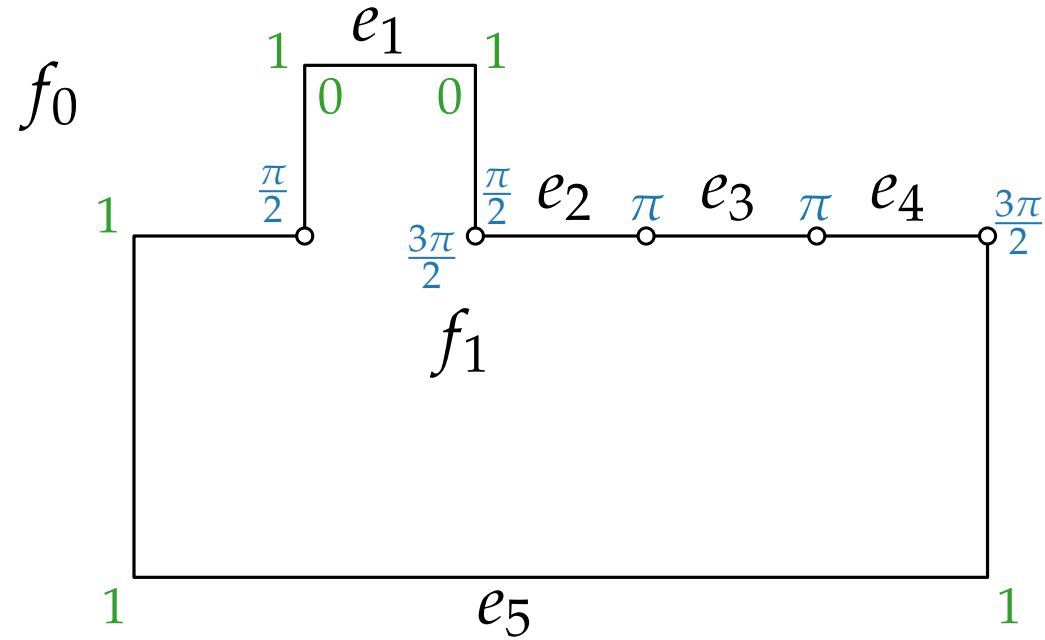
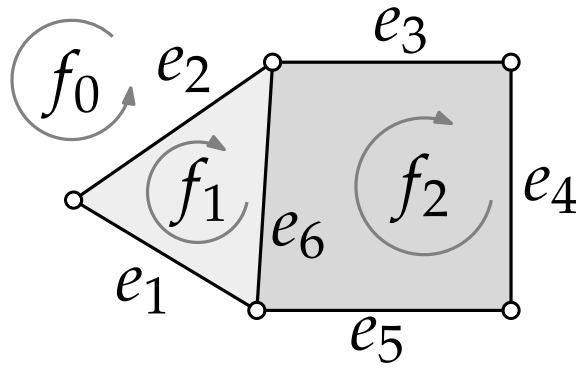


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

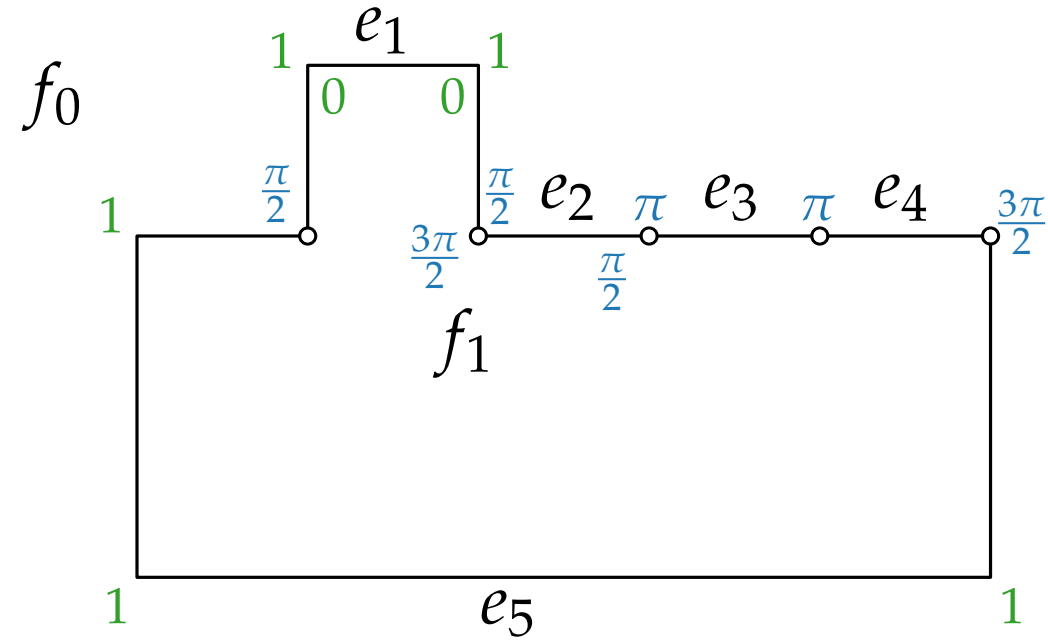
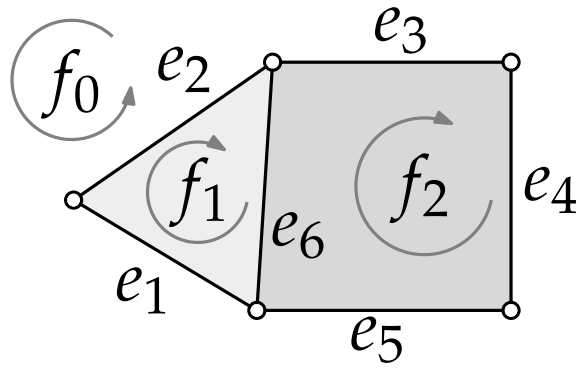


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

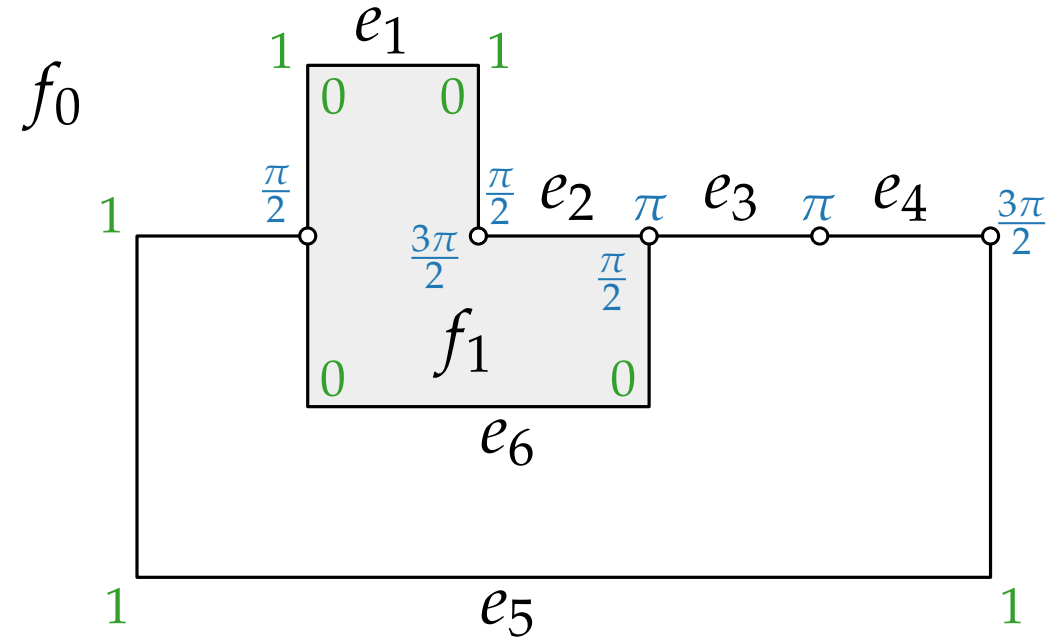
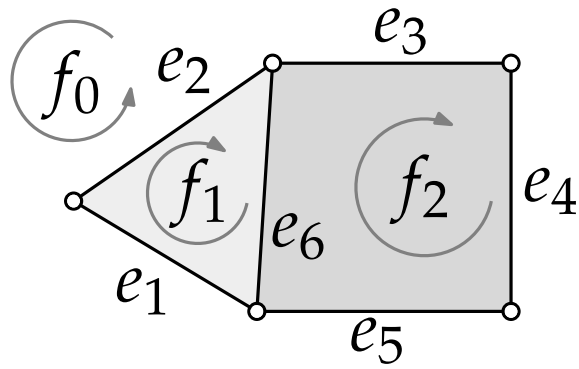


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

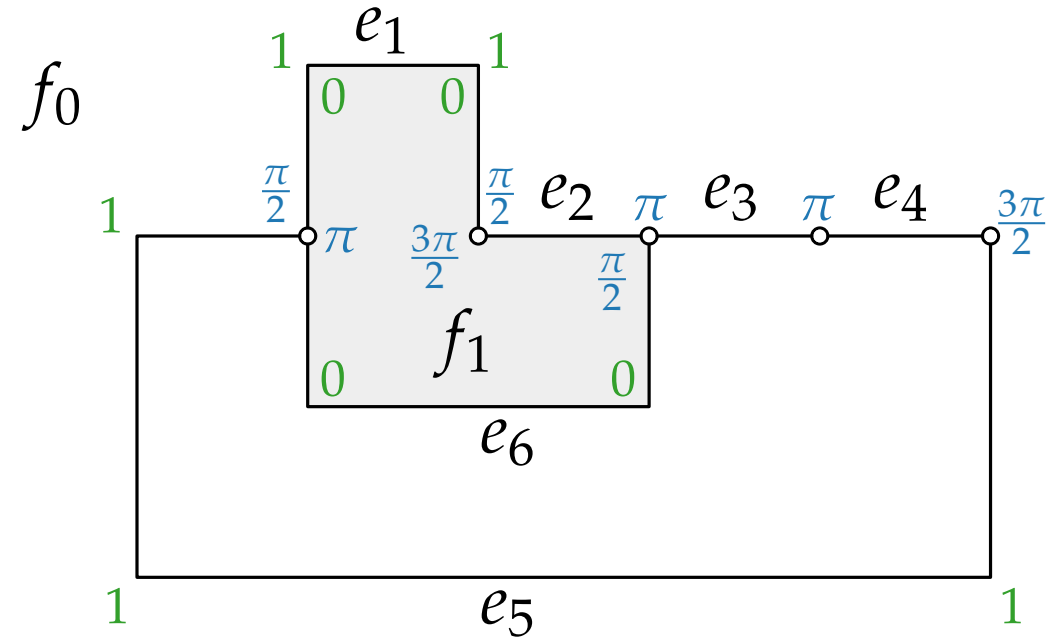
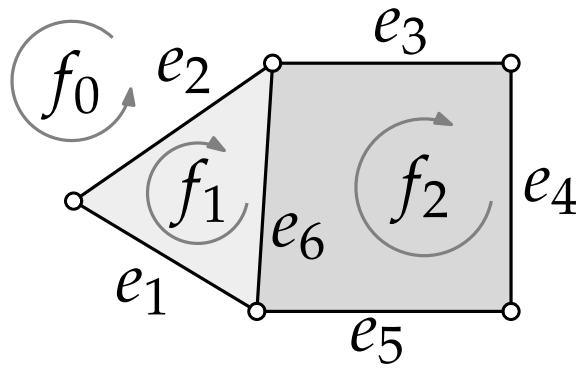


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

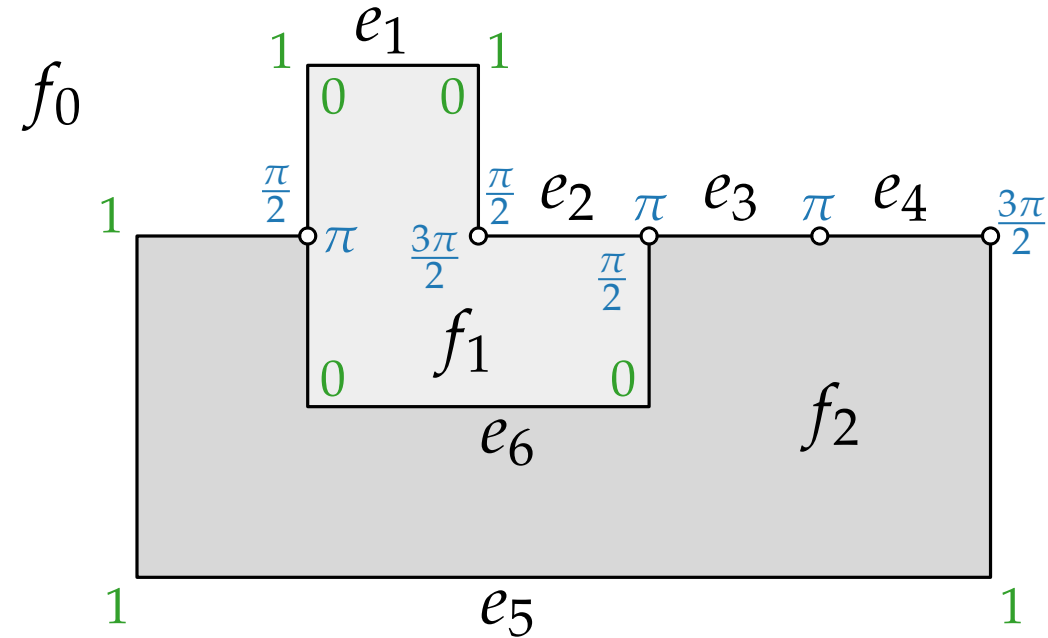
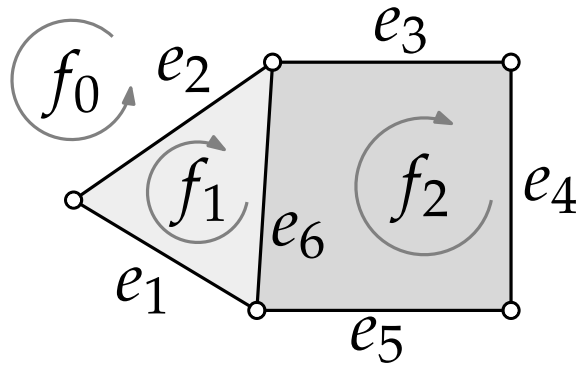


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

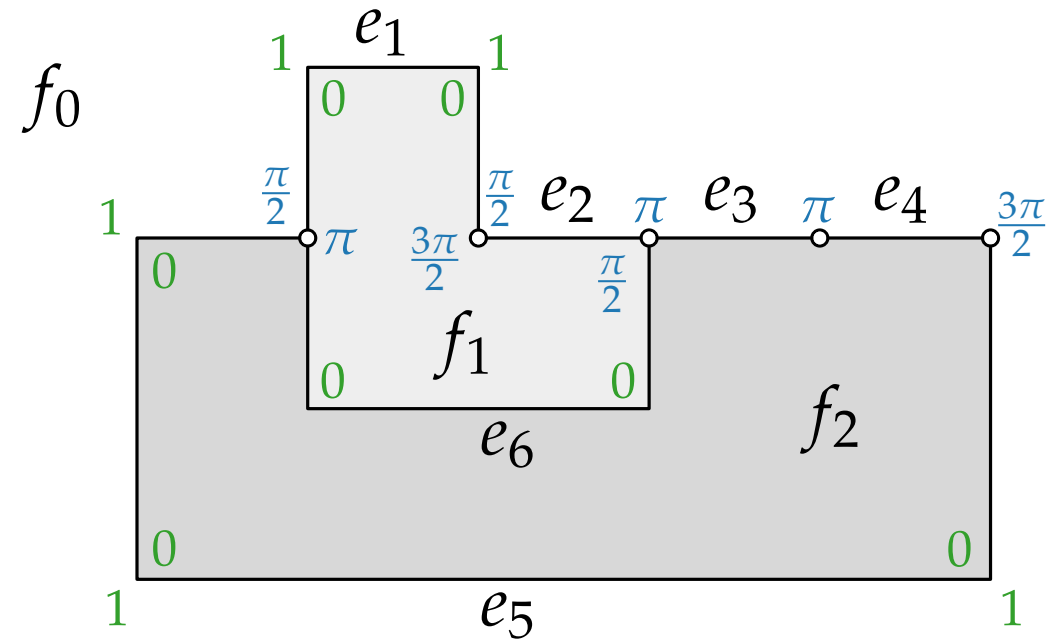
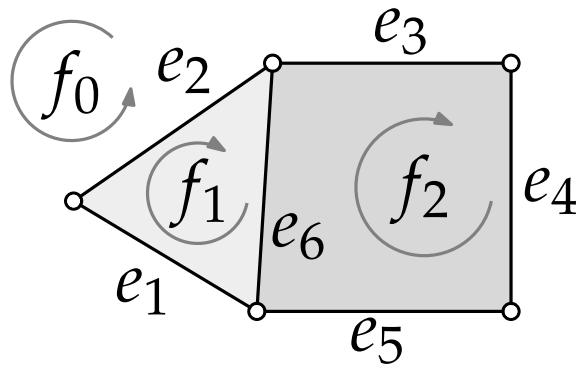


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

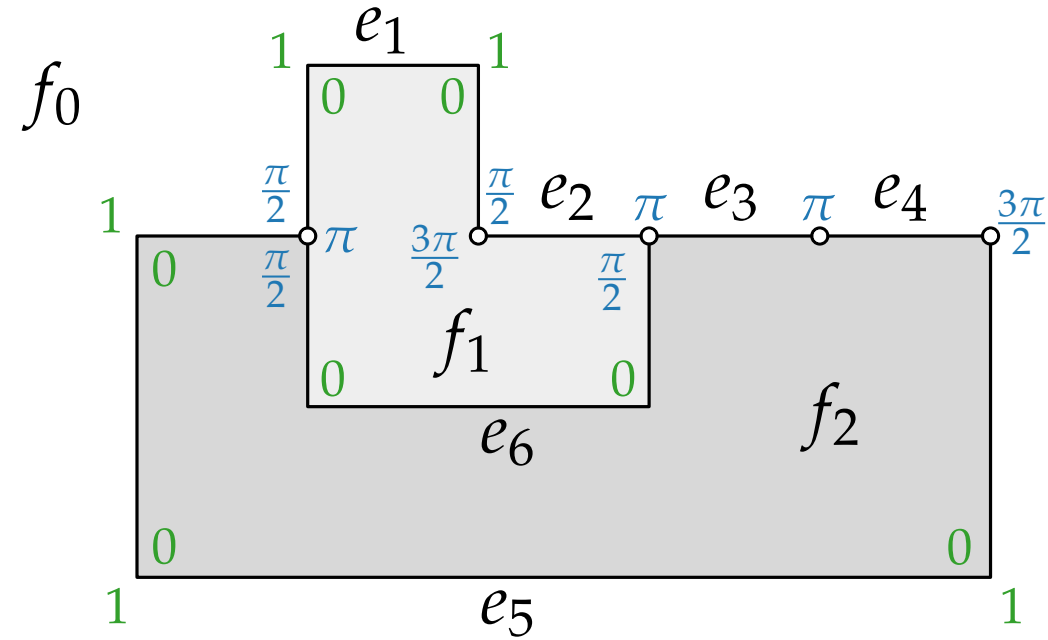
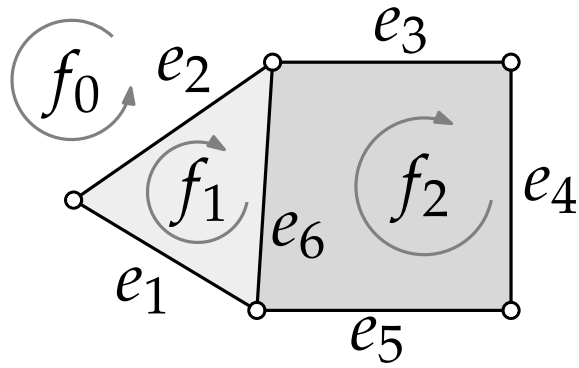


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

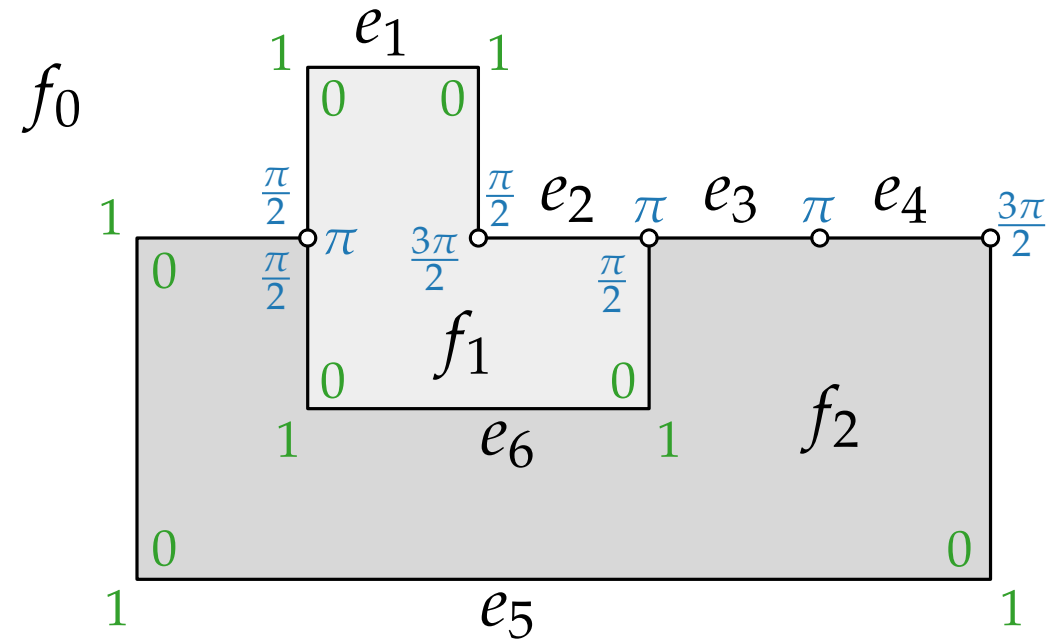
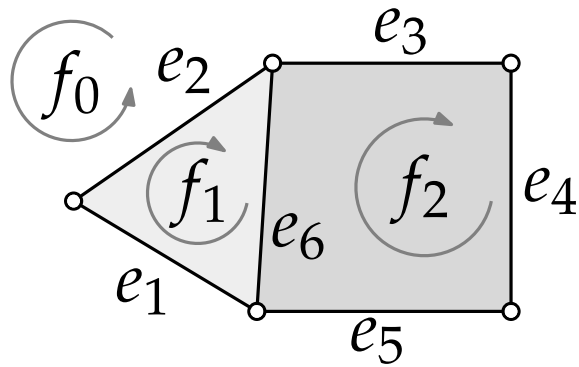


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$



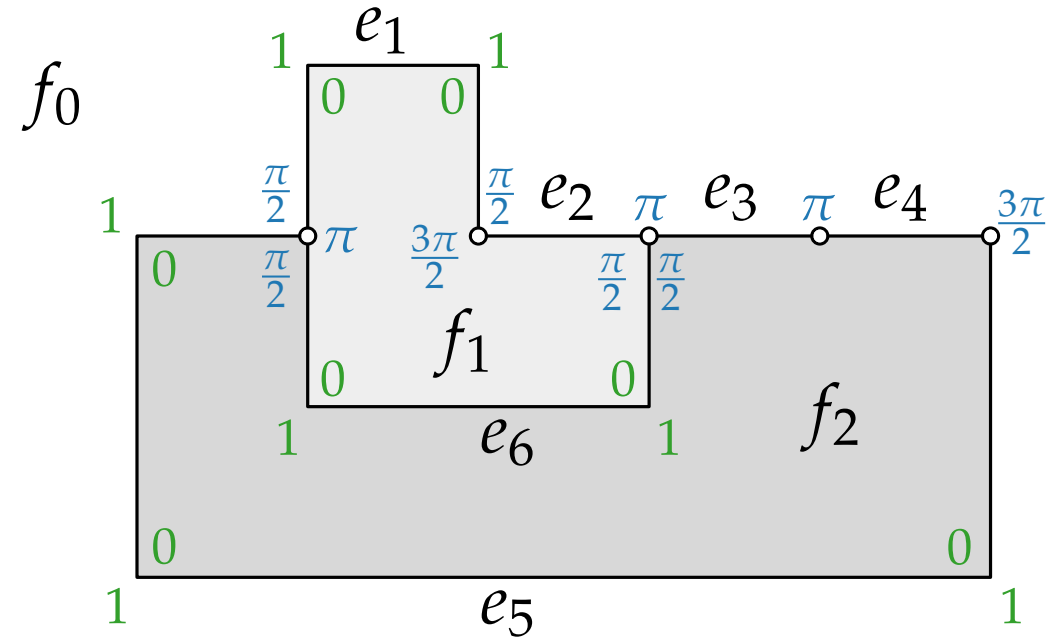
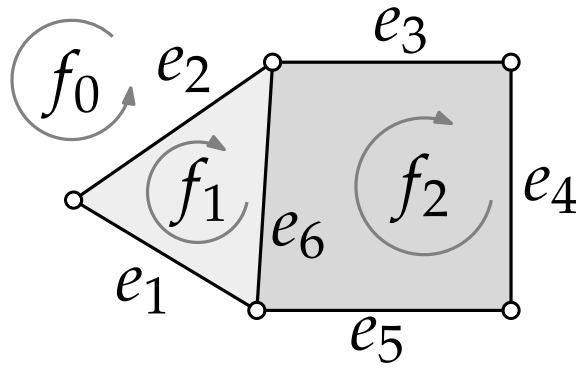


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

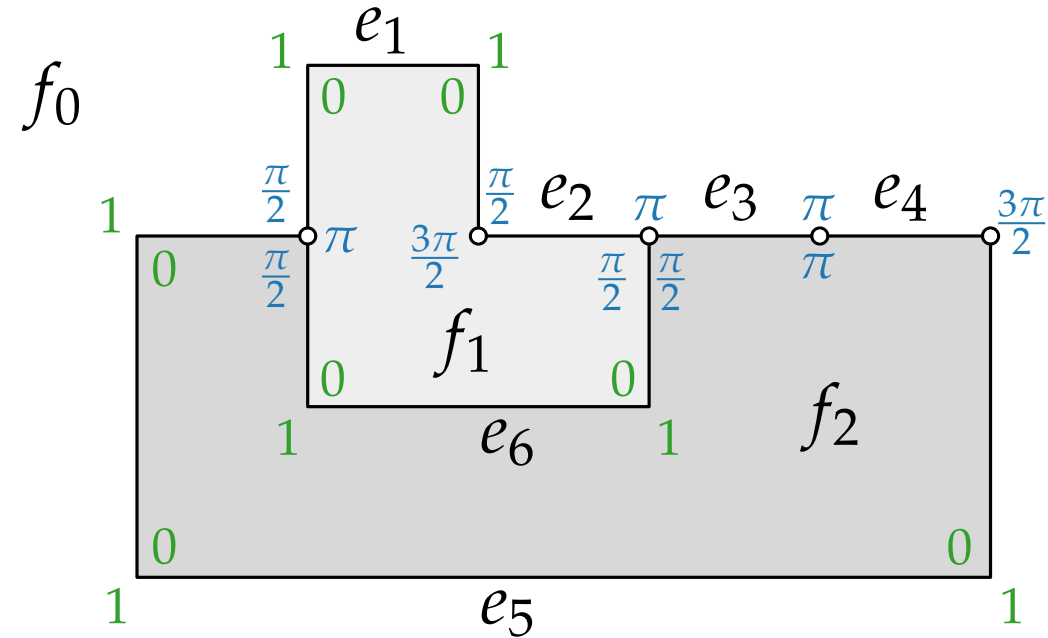
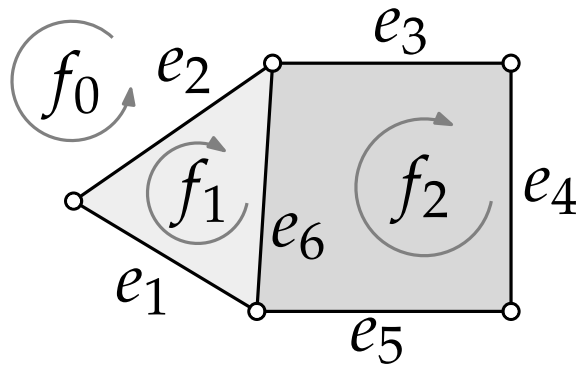


# Orthogonal Representation – Example

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$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

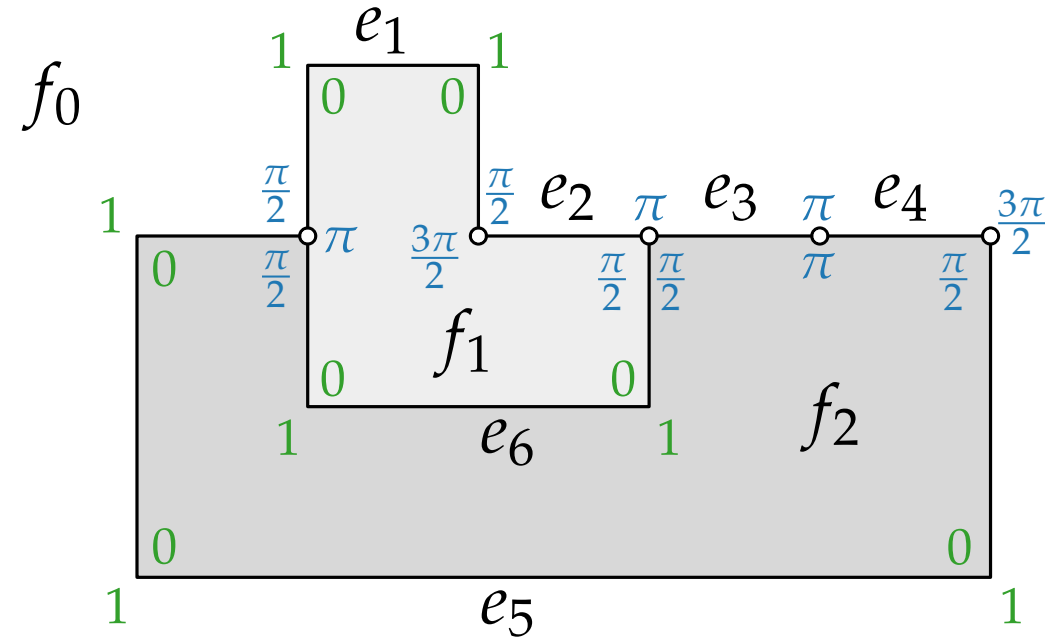
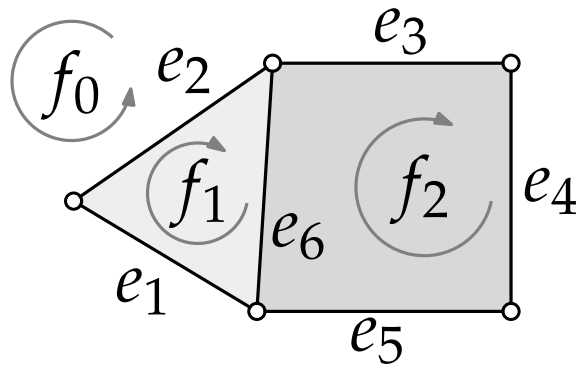


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

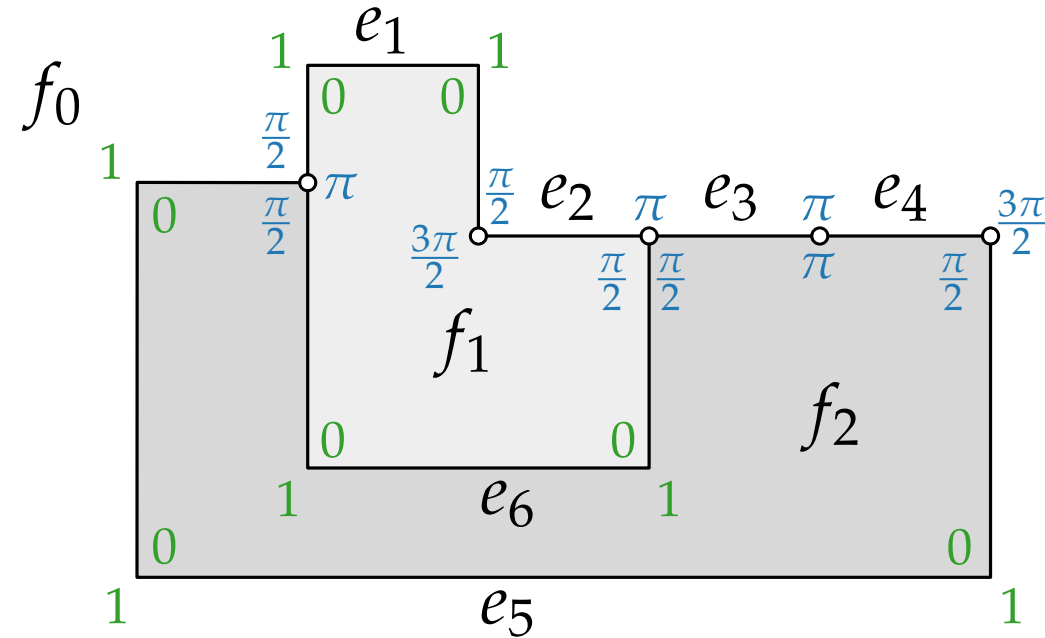
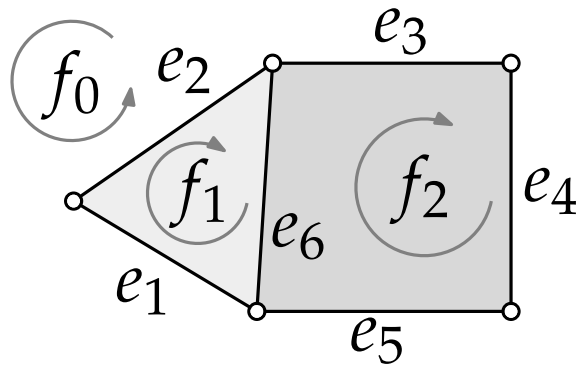


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$

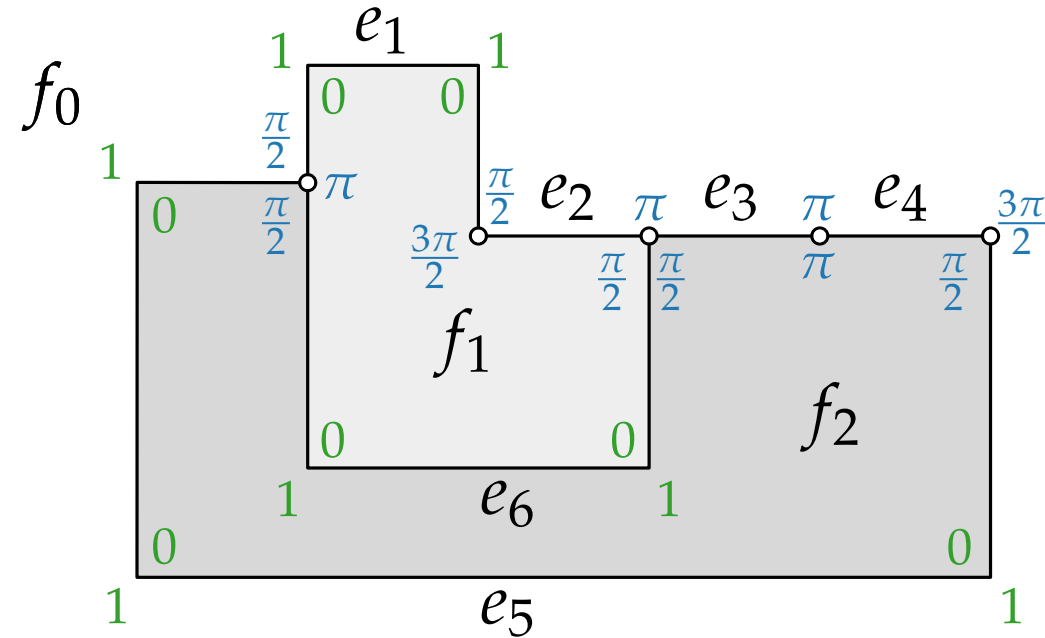
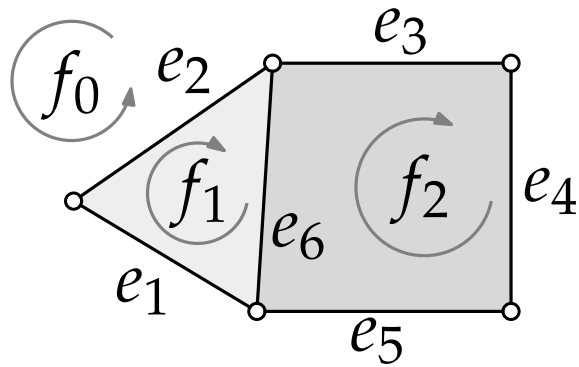


# Orthogonal Representation – Example

$$H(f_0) = ((e_1, 11, \frac{\pi}{2}), (e_5, 111, \frac{3\pi}{2}), (e_4, \emptyset, \pi), (e_3, \emptyset, \pi), (e_2, \emptyset, \frac{\pi}{2}))$$

$$H(f_1) = ((e_1, 00, \frac{3\pi}{2}), (e_2, \emptyset, \frac{\pi}{2}), (e_6, 00, \pi))$$

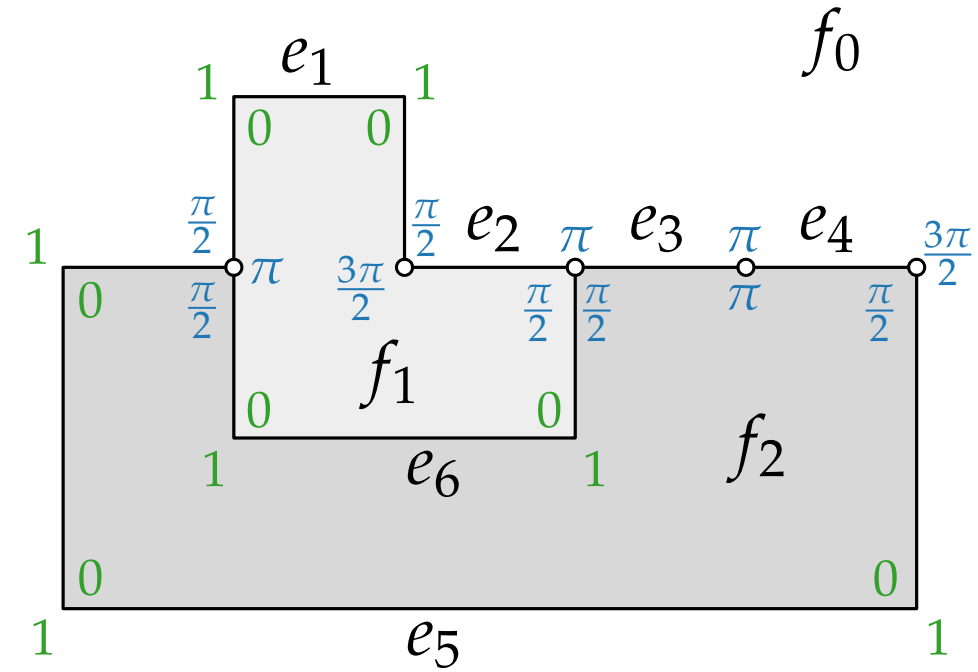
$$H(f_2) = ((e_5, 000, \frac{\pi}{2}), (e_6, 11, \frac{\pi}{2}), (e_3, \emptyset, \pi), (e_4, \emptyset, \frac{\pi}{2}))$$



Concrete coordinates are not fixed yet!

# Correctness of an Orthogonal Representation

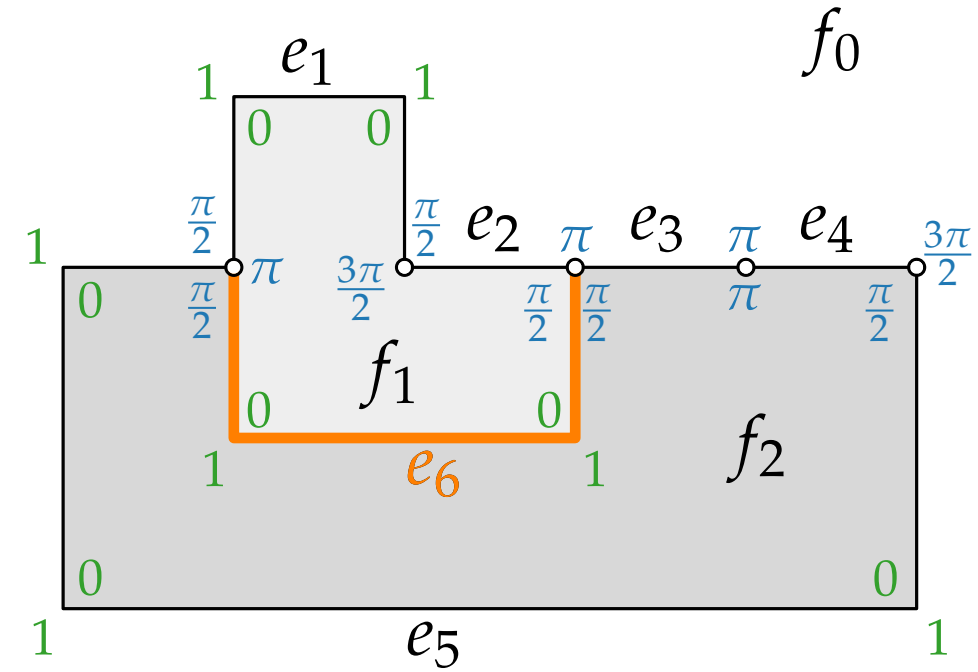
(H1)  $H(G)$  corresponds to  $F, f_0$ .



# Correctness of an Orthogonal Representation

(H1)  $H(G)$  corresponds to  $F, f_0$ .

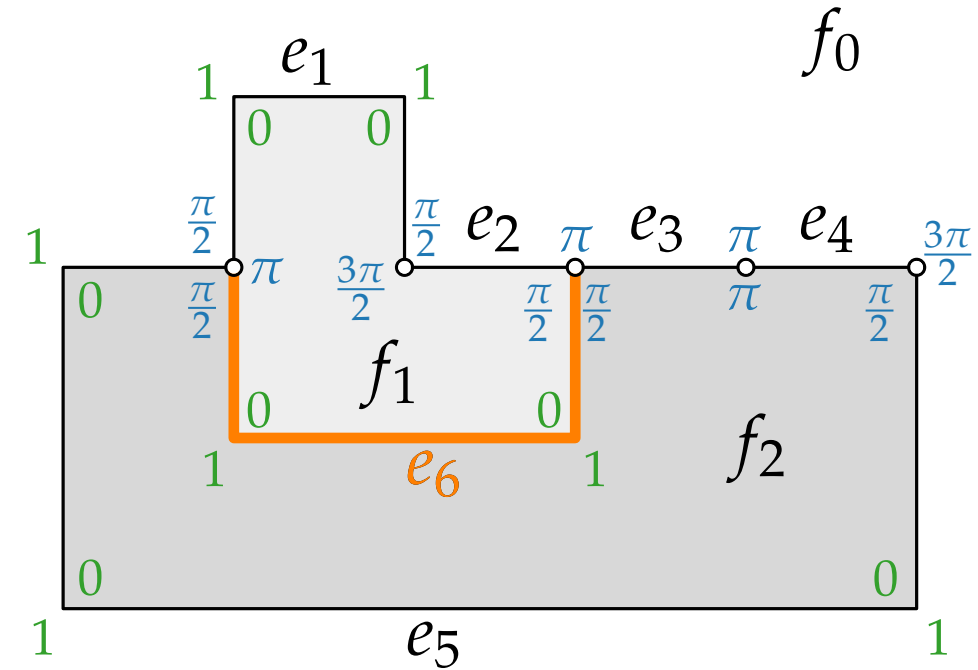
(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$



# Correctness of an Orthogonal Representation

(H1)  $H(G)$  corresponds to  $F, f_0$ .

(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$  with  
 $((u, v), \delta_1, \alpha_1) \in H(f)$  and  $((v, u), \delta_2, \alpha_2) \in H(g)$

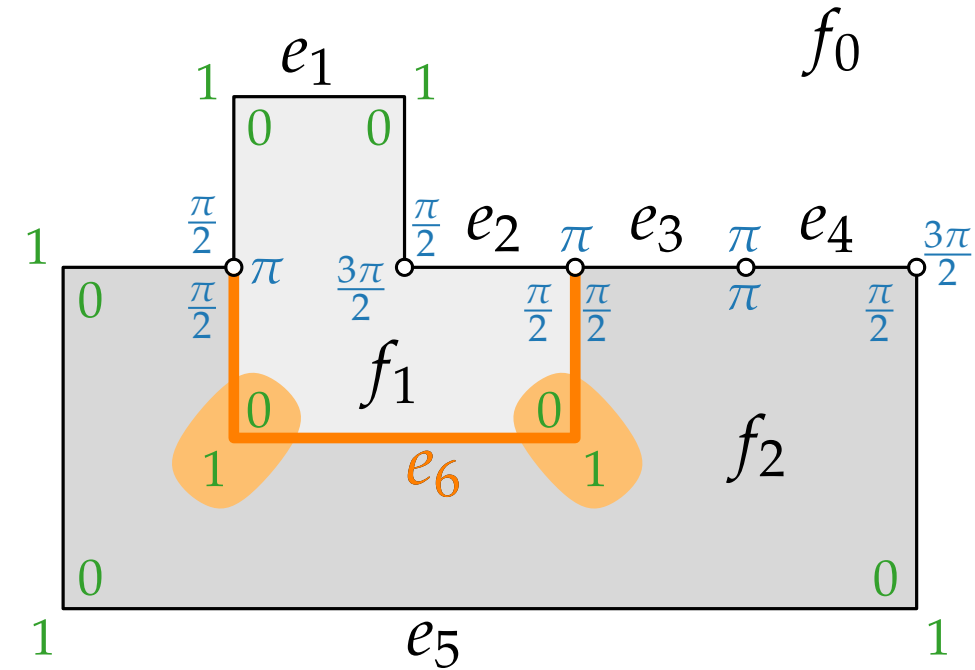




# Correctness of an Orthogonal Representation

(H1)  $H(G)$  corresponds to  $F, f_0$ .

(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$  with  $((u, v), \delta_1, \alpha_1) \in H(f)$  and  $((v, u), \delta_2, \alpha_2) \in H(g)$  sequence  $\delta_1$  is reversed and inverted  $\delta_2$ .

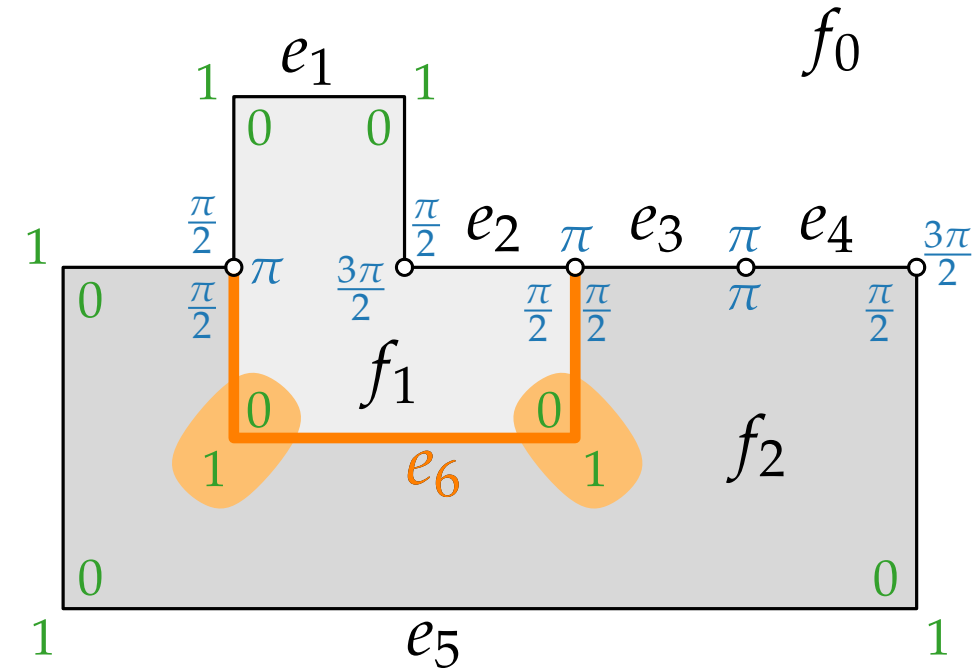


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(H1)  $H(G)$  corresponds to  $F, f_0$ .

(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$  with  $((u, v), \delta_1, \alpha_1) \in H(f)$  and  $((v, u), \delta_2, \alpha_2) \in H(g)$  sequence  $\delta_1$  is reversed and inverted  $\delta_2$ .

(H3) Let  $|\delta|_0$  (resp.  $|\delta|_1$ ) be the number of zeros (resp. ones) in  $\delta$  and  $r = (e, \delta, \alpha)$ .

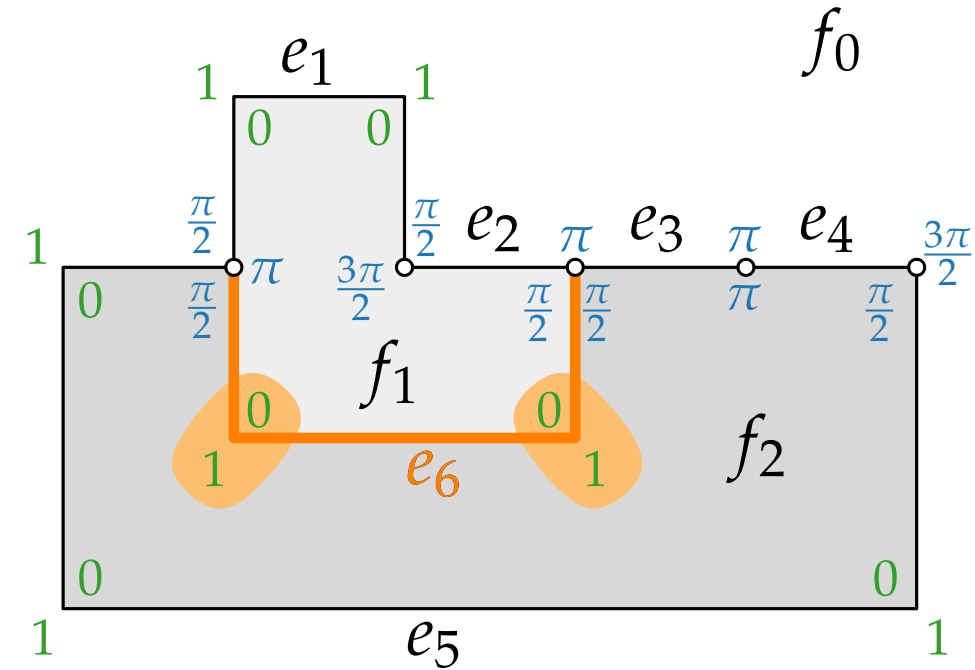


# Correctness of an Orthogonal Representation

(H1)  $H(G)$  corresponds to  $F, f_0$ .

(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$  with  $((u, v), \delta_1, \alpha_1) \in H(f)$  and  $((v, u), \delta_2, \alpha_2) \in H(g)$  sequence  $\delta_1$  is reversed and inverted  $\delta_2$ .

(H3) Let  $|\delta|_0$  (resp.  $|\delta|_1$ ) be the number of zeros (resp. ones) in  $\delta$  and  $r = (e, \delta, \alpha)$ .  
Let  $C(r) := |\delta|_0 - |\delta|_1 + 2 - \alpha \cdot 2/\pi$ .



# Correctness of an Orthogonal Representation

(H1)  $H(G)$  corresponds to  $F, f_0$ .

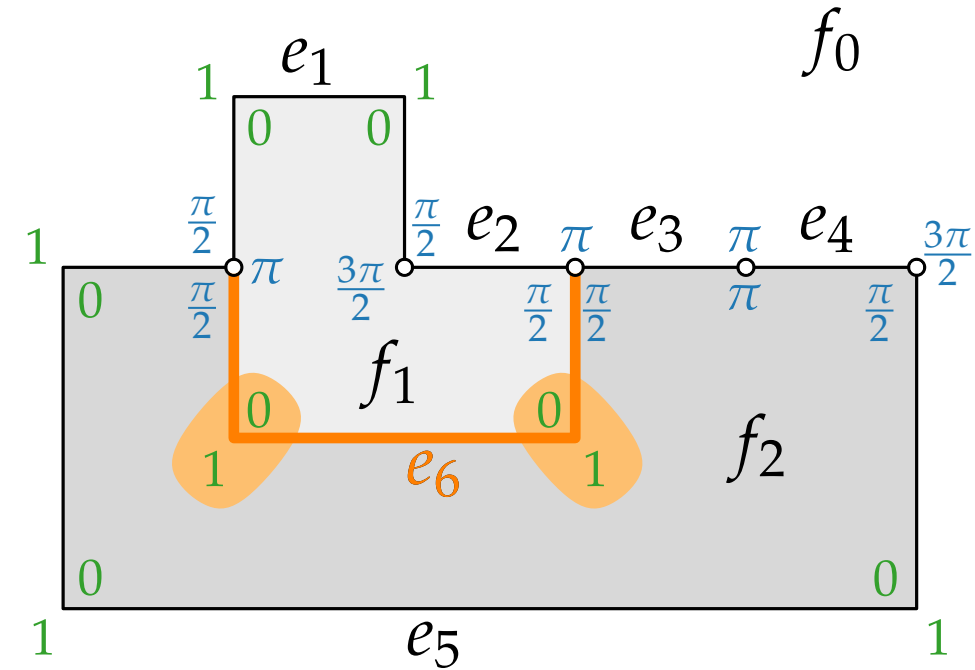
(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$  with  $((u, v), \delta_1, \alpha_1) \in H(f)$  and  $((v, u), \delta_2, \alpha_2) \in H(g)$  sequence  $\delta_1$  is reversed and inverted  $\delta_2$ .

(H3) Let  $|\delta|_0$  (resp.  $|\delta|_1$ ) be the number of zeros (resp. ones) in  $\delta$  and  $r = (e, \delta, \alpha)$ .

Let  $C(r) := |\delta|_0 - |\delta|_1 + 2 - \alpha \cdot 2/\pi$ .

For each **face**  $f$  it holds that:

$$\sum_{r \in H(f)} C(r) = \begin{cases} -4 & \text{if } f = f_0 \\ +4 & \text{otherwise.} \end{cases}$$



# Correctness of an Orthogonal Representation

(H1)  $H(G)$  corresponds to  $F, f_0$ .

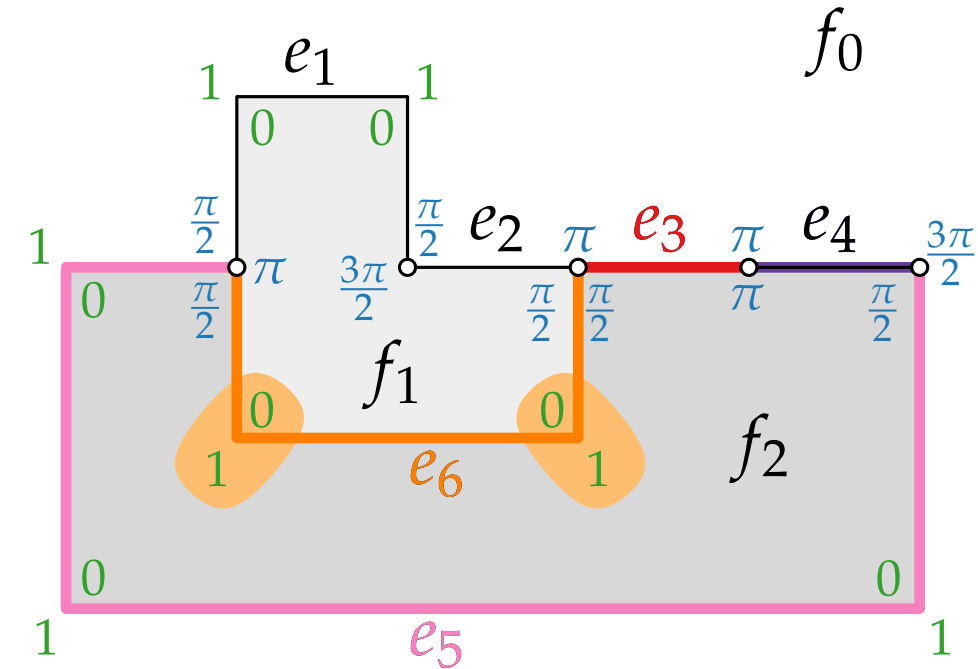
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$$C(e_3) = - + 2 - =$$

$$C(e_4) = - + 2 - =$$

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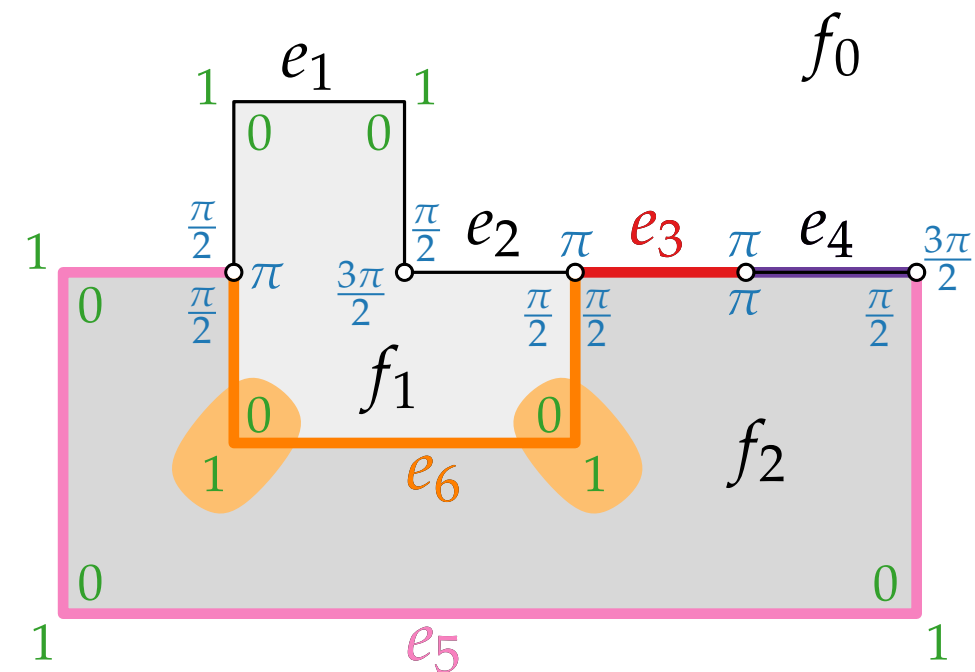
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$$\begin{aligned} C(e_3) &= 0 - + 2 - = \\ C(e_4) &= - + 2 - = \\ C(e_5) &= - + 2 - = \\ C(e_6) &= - + 2 - = \end{aligned}$$

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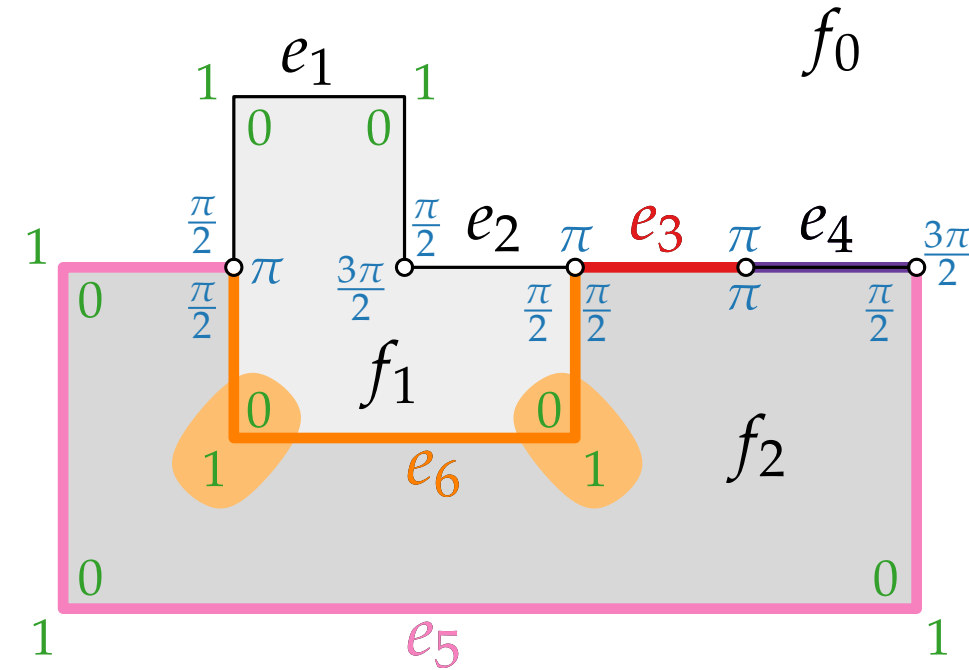
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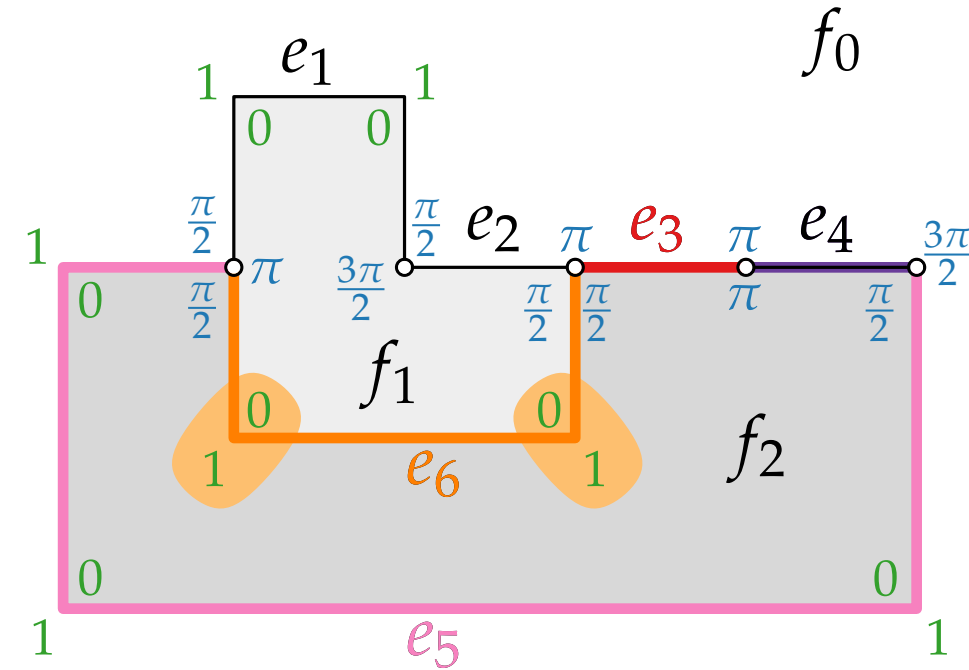
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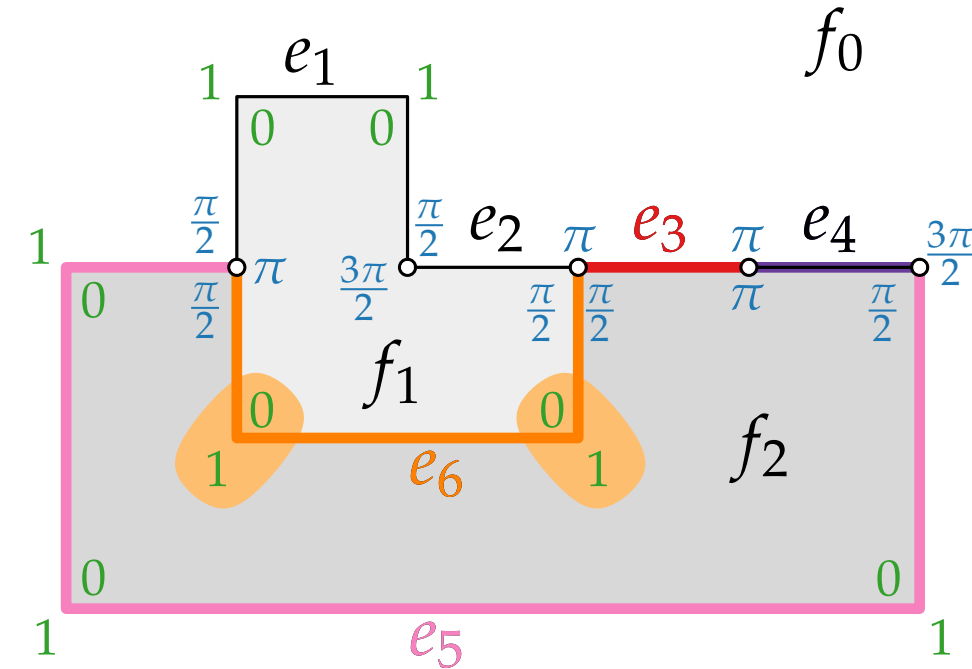
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$$C(e_3) = 0 - 0 + 2 - \pi \cdot \frac{2}{\pi} = 0$$

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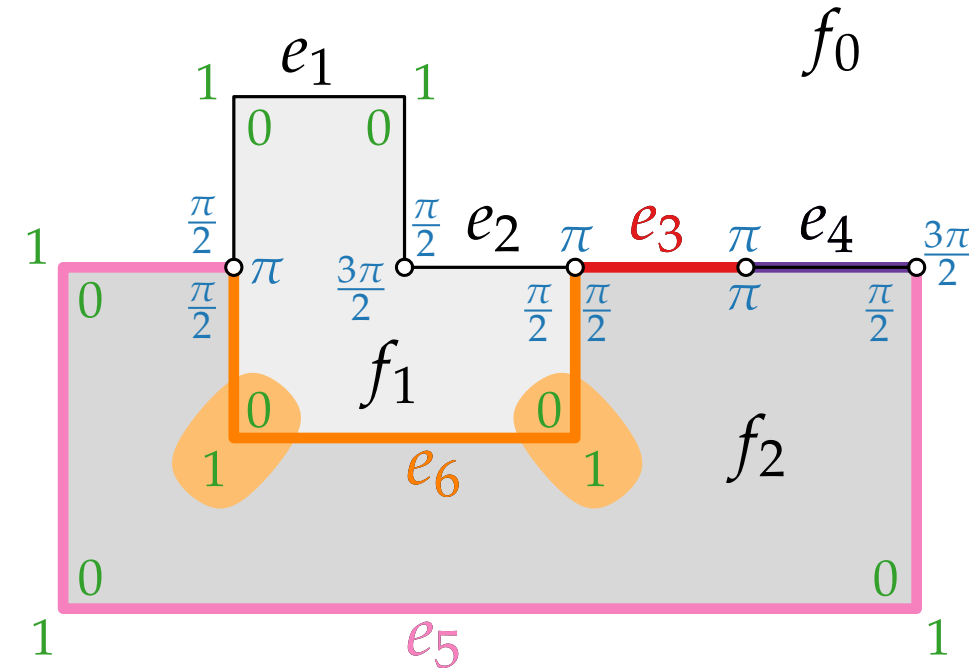
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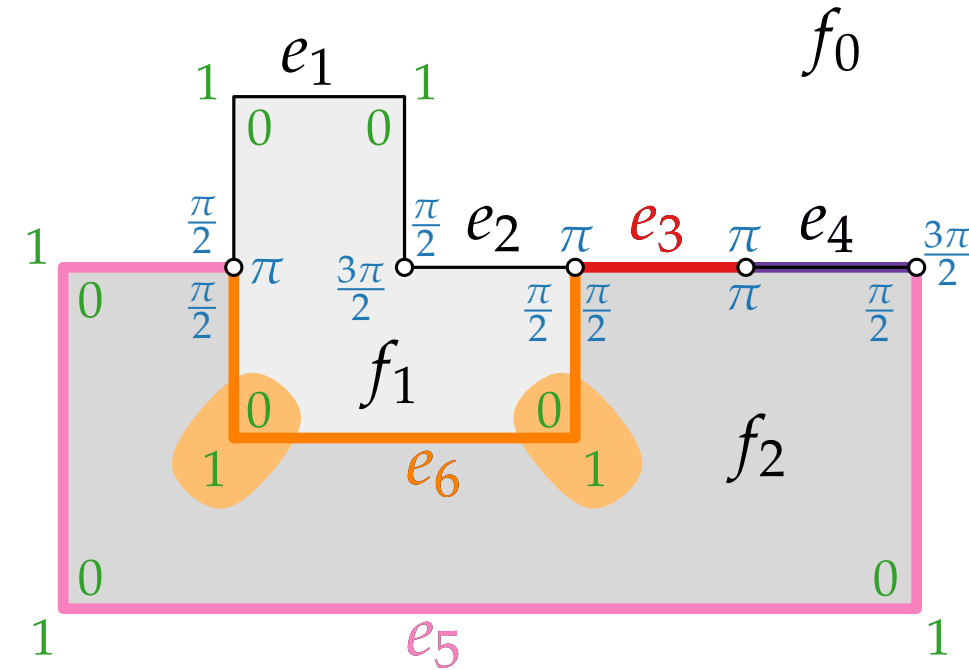
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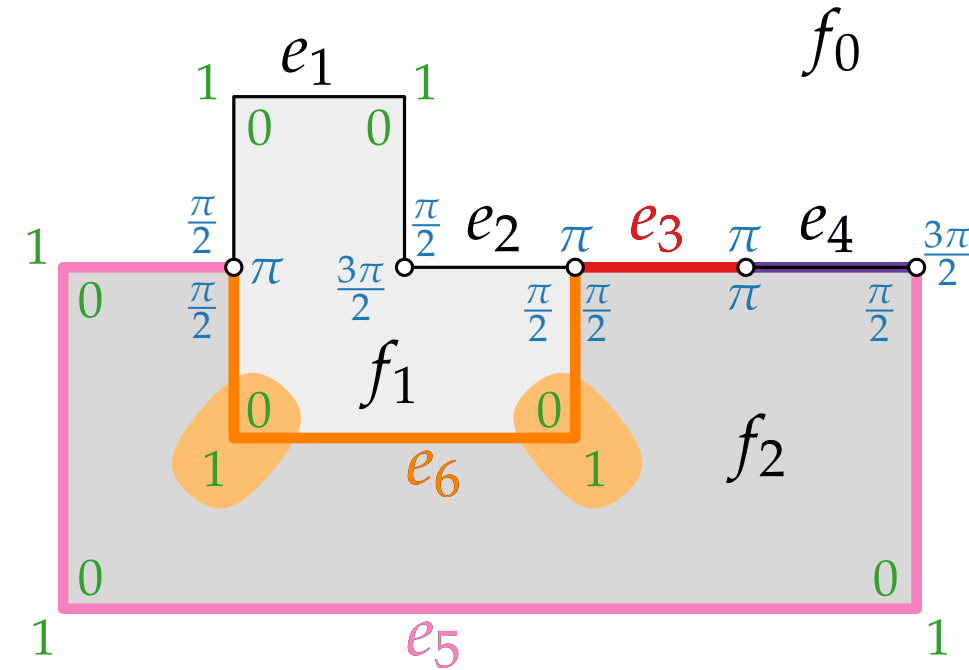
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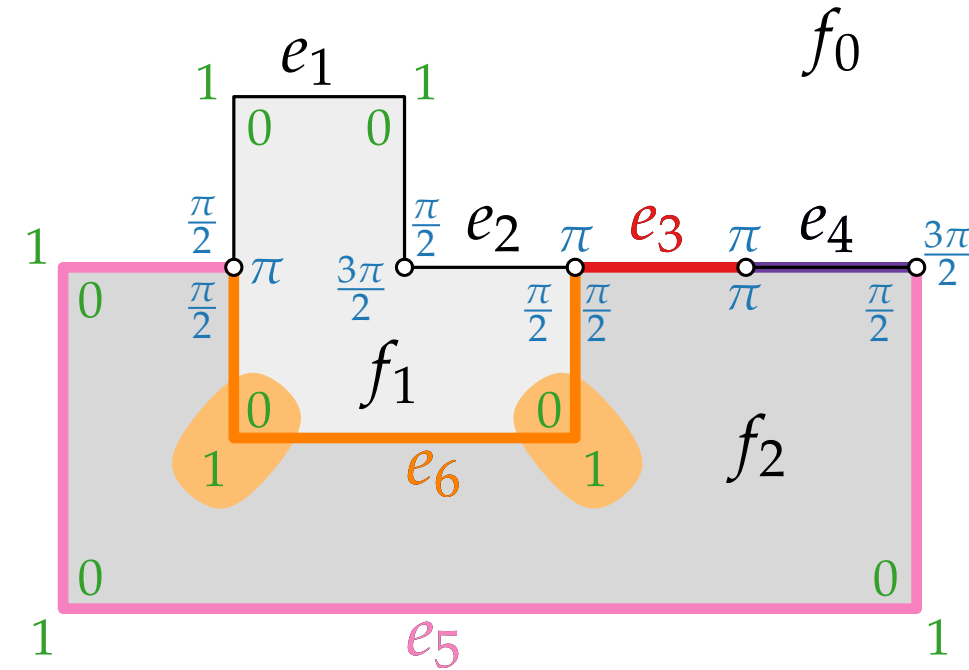
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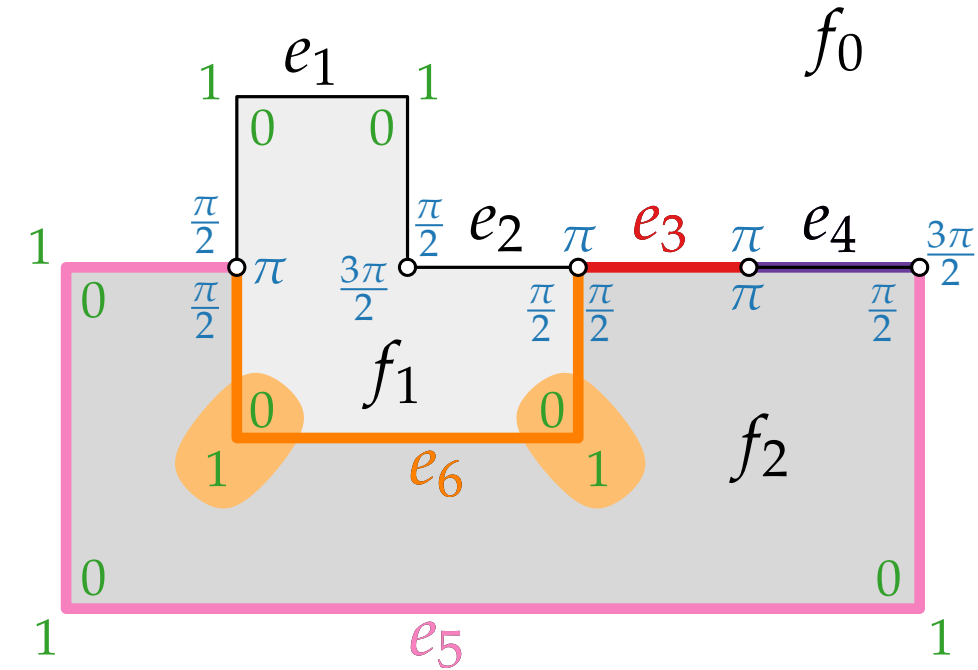
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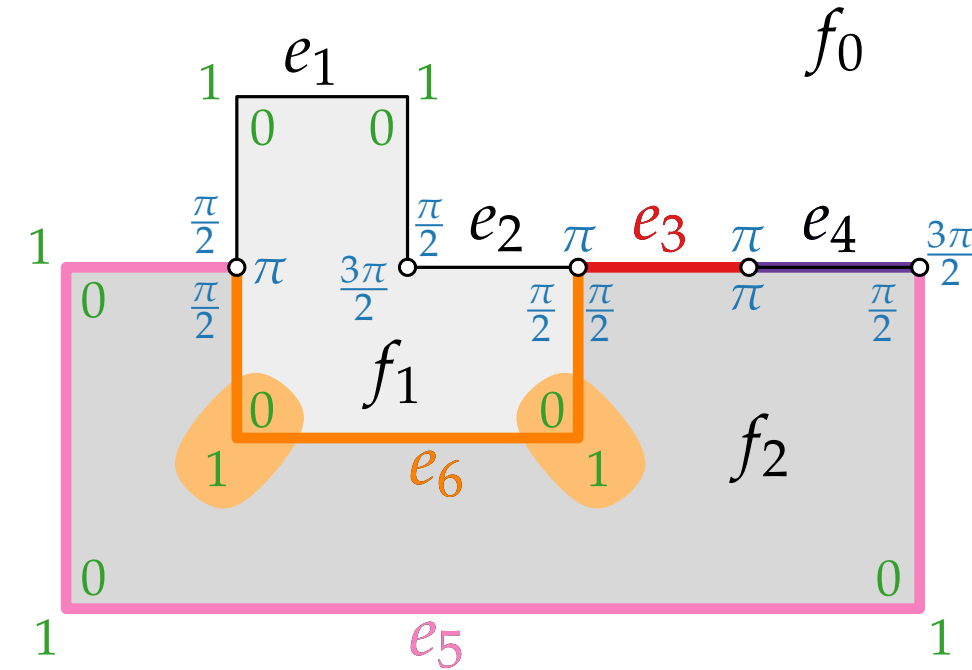
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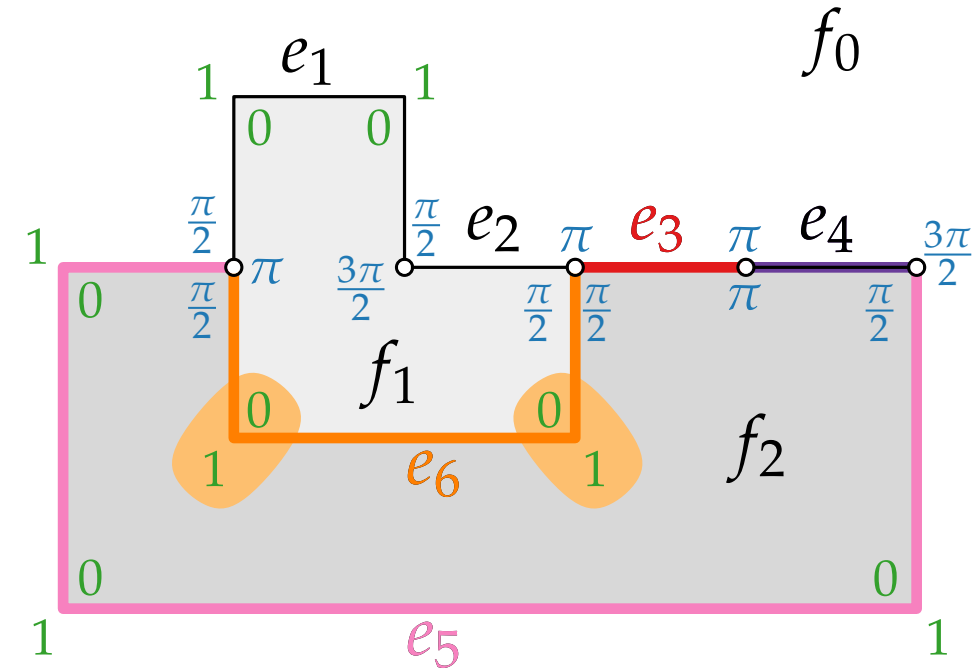
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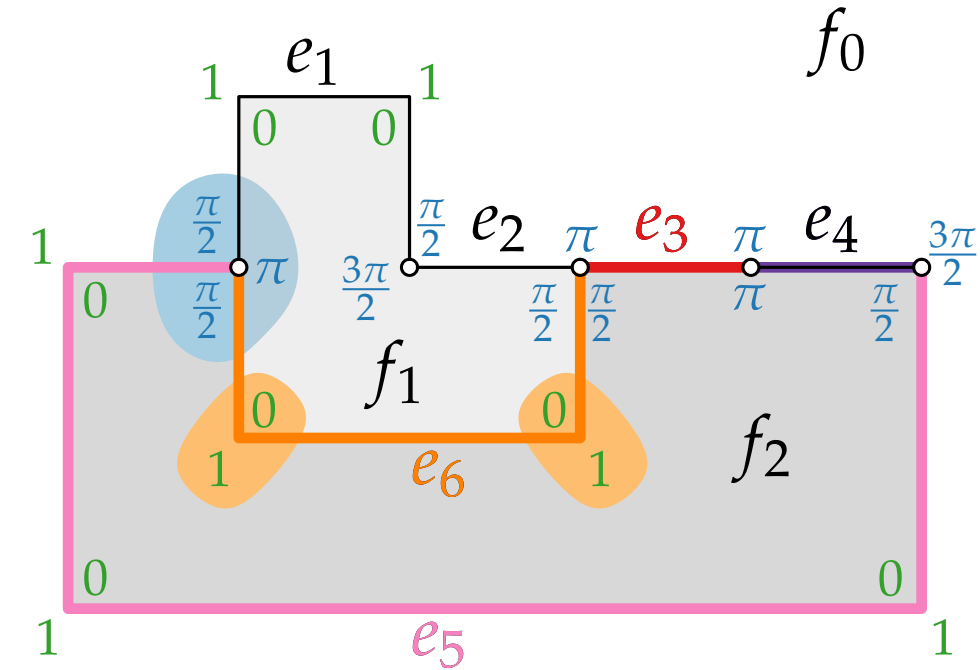
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(H4) For each **vertex**  $v$  the sum of incident angles is  $2\pi$ .



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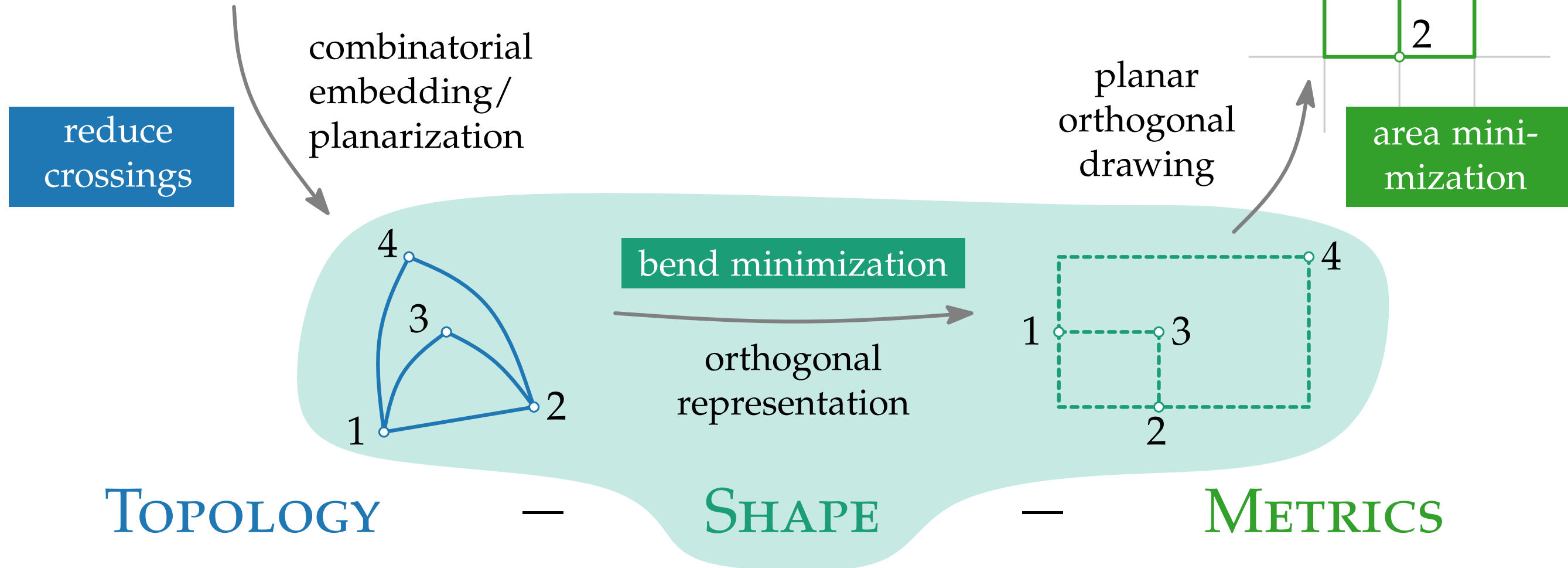
# Topology – Shape – Metrics

Three-step approach:

[Tamassia 1987]

$$V = \{v_1, v_2, v_3, v_4\}$$

$$E = \{v_1v_2, v_1v_3, v_1v_4, v_2v_3, v_2v_4\}$$

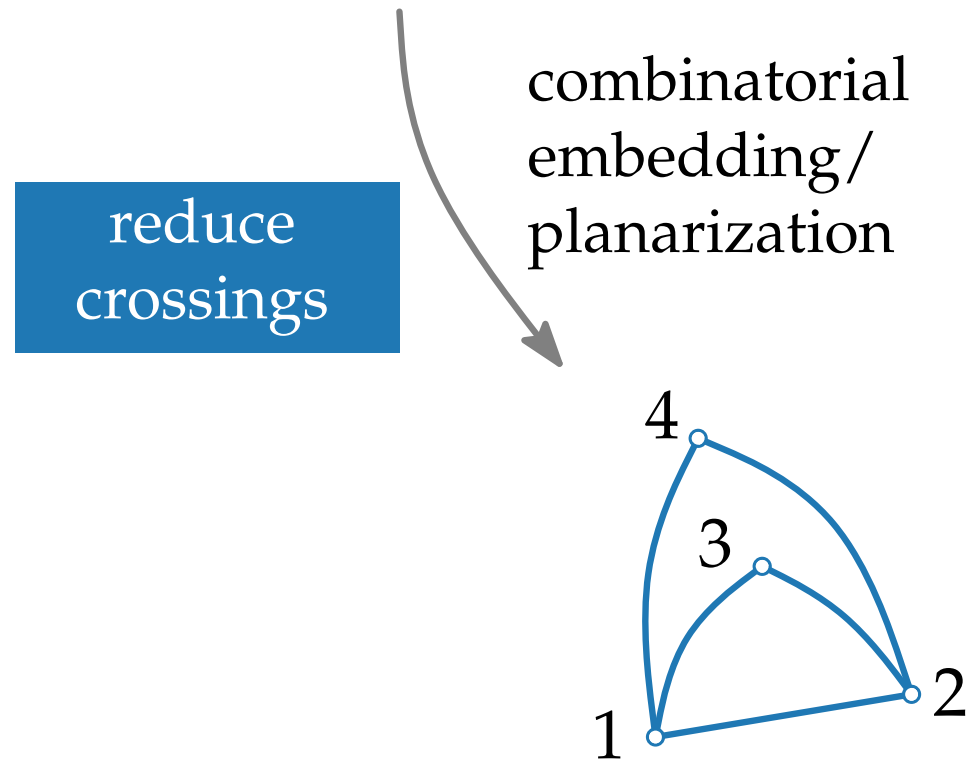


# Topology – Shape – Metrics

Three-step approach:

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$$E = \{v_1v_2, v_1v_3, v_1v_4, v_2v_3, v_2v_4\}$$



TOPOLOGY

—

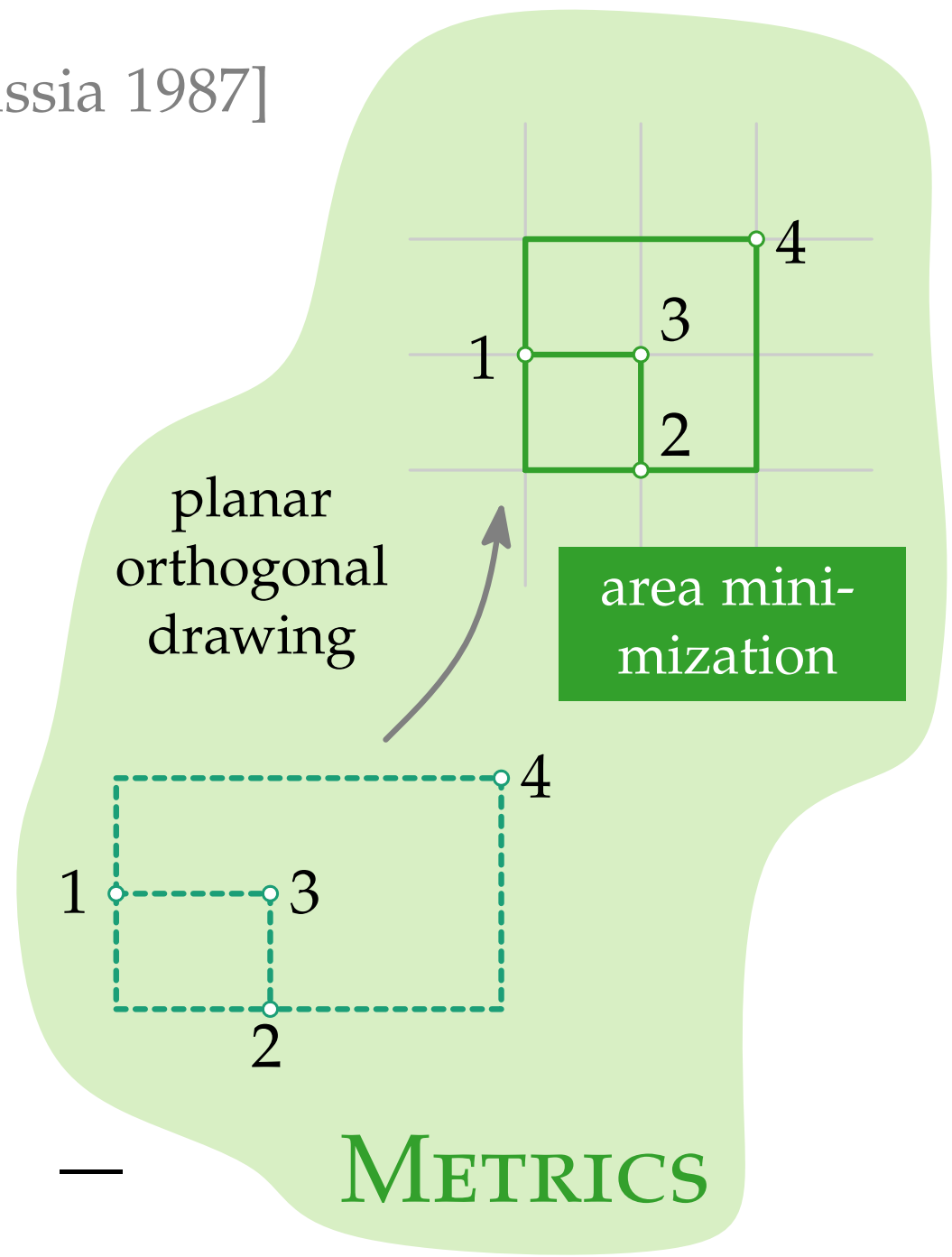
bend minimization

orthogonal representation

SHAPE

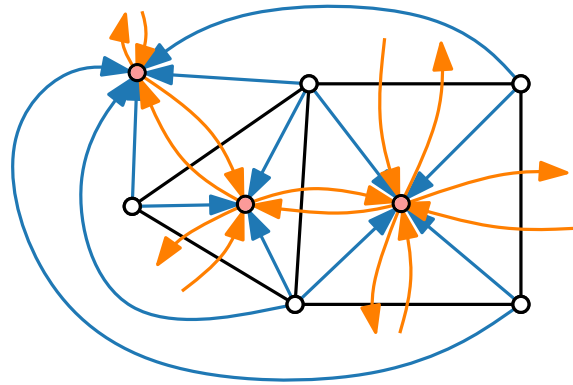
—

[Tamassia 1987]

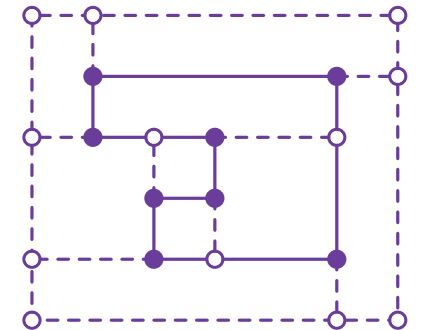


METRICS

# Visualization of Graphs

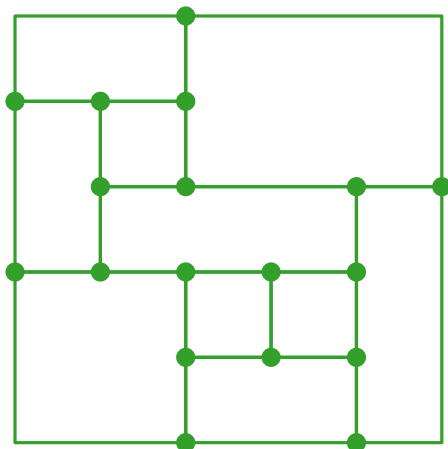


## Lecture 6: Orthogonal Layouts

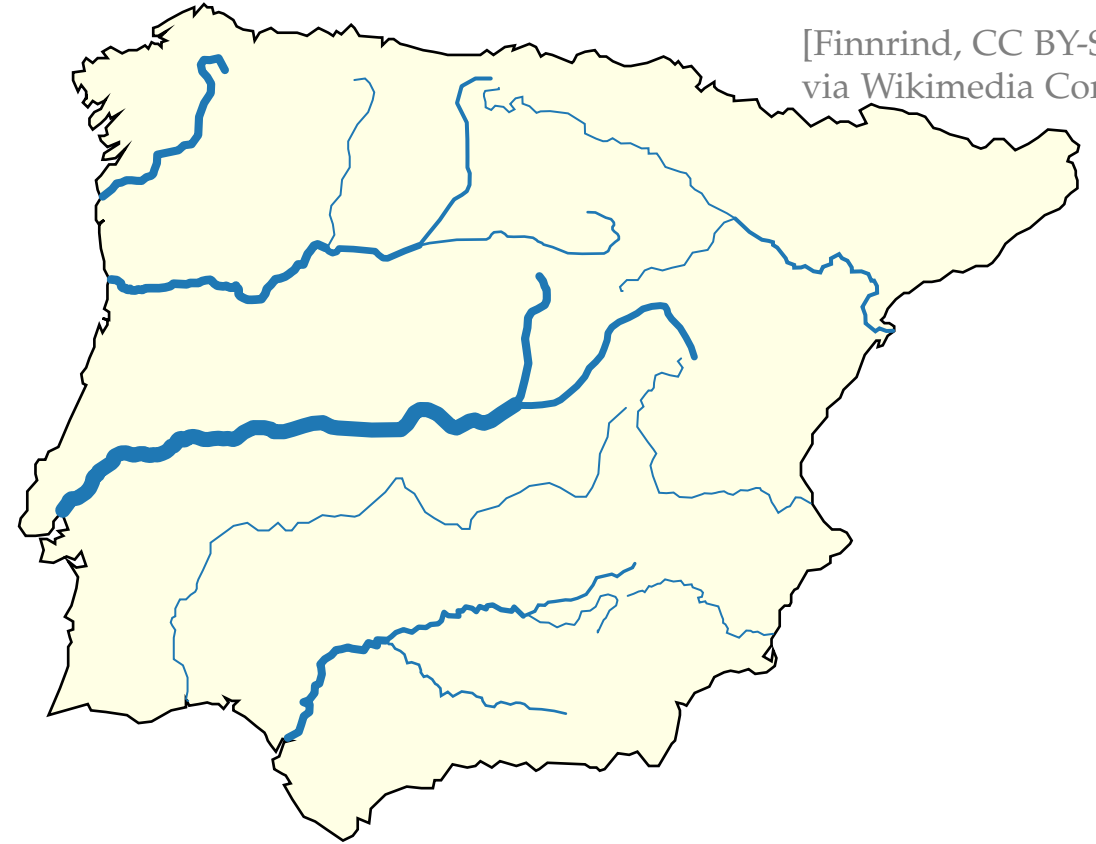


## Part III: Flow Networks

Philipp Kindermann



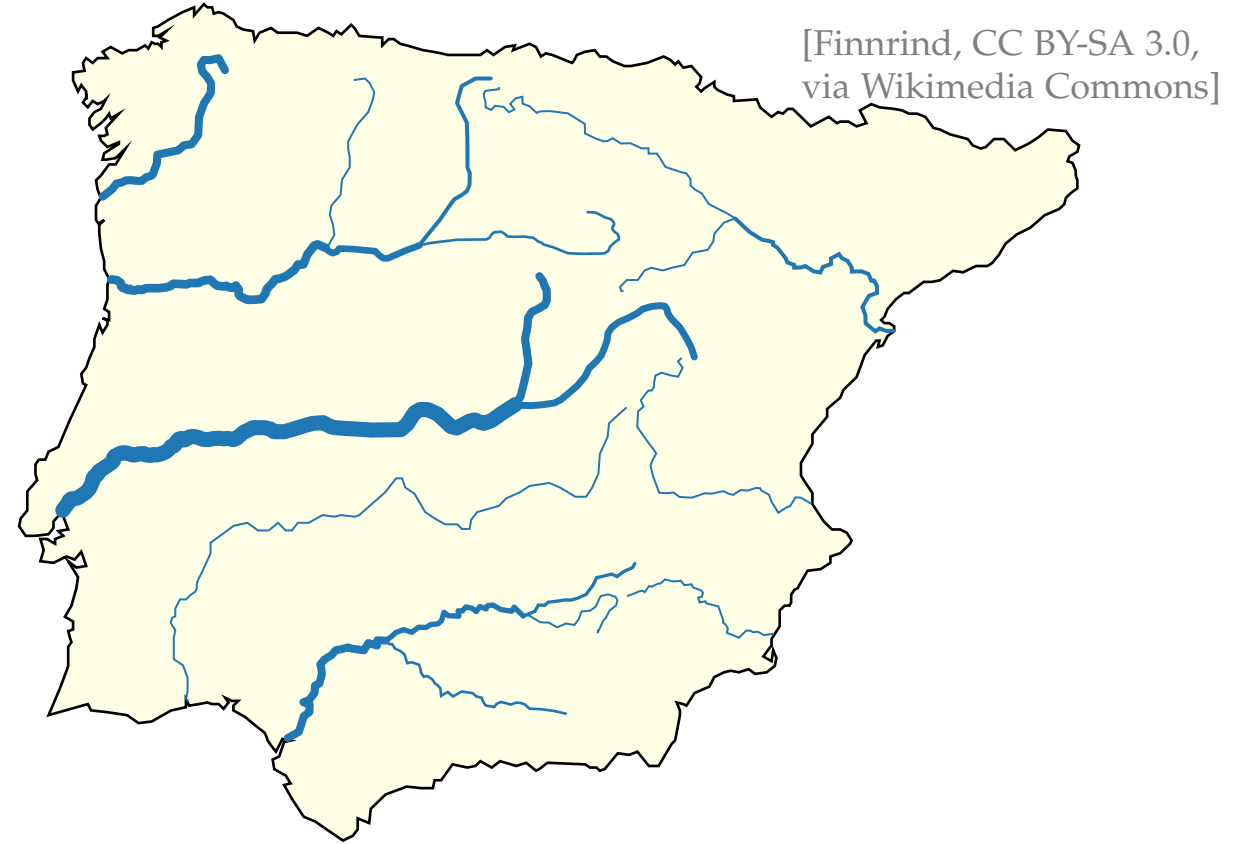
# Flow Networks



[Finnrind, CC BY-SA 3.0,  
via Wikimedia Commons]

# Flow Networks

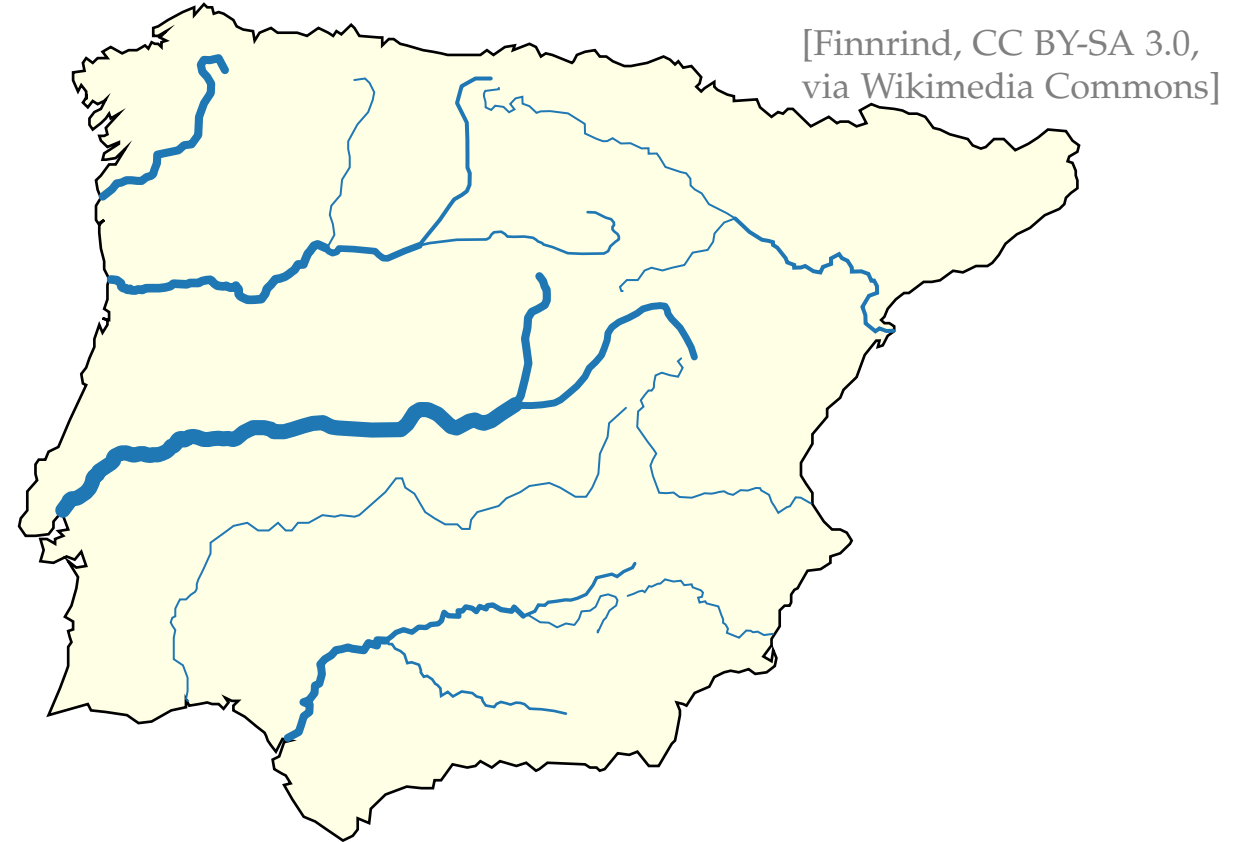
**Flow network**  $(G = (V, E); S, T; u)$  with



# Flow Networks

**Flow network**  $(G = (V, E); S, T; u)$  with

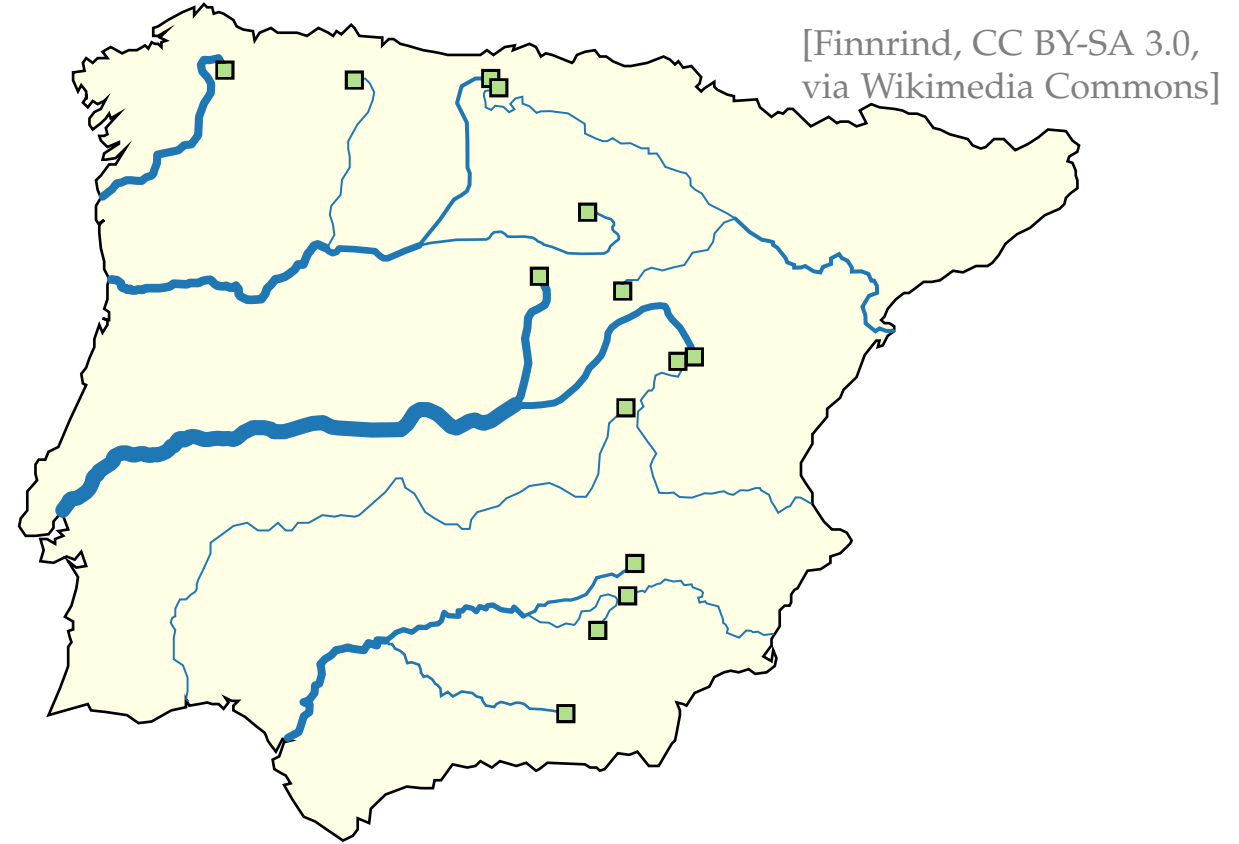
- directed graph  $G = (V, E)$



# Flow Networks

**Flow network**  $(G = (V, E); S, T; u)$  with

- directed graph  $G = (V, E)$
- *sources*  $S \subseteq V$

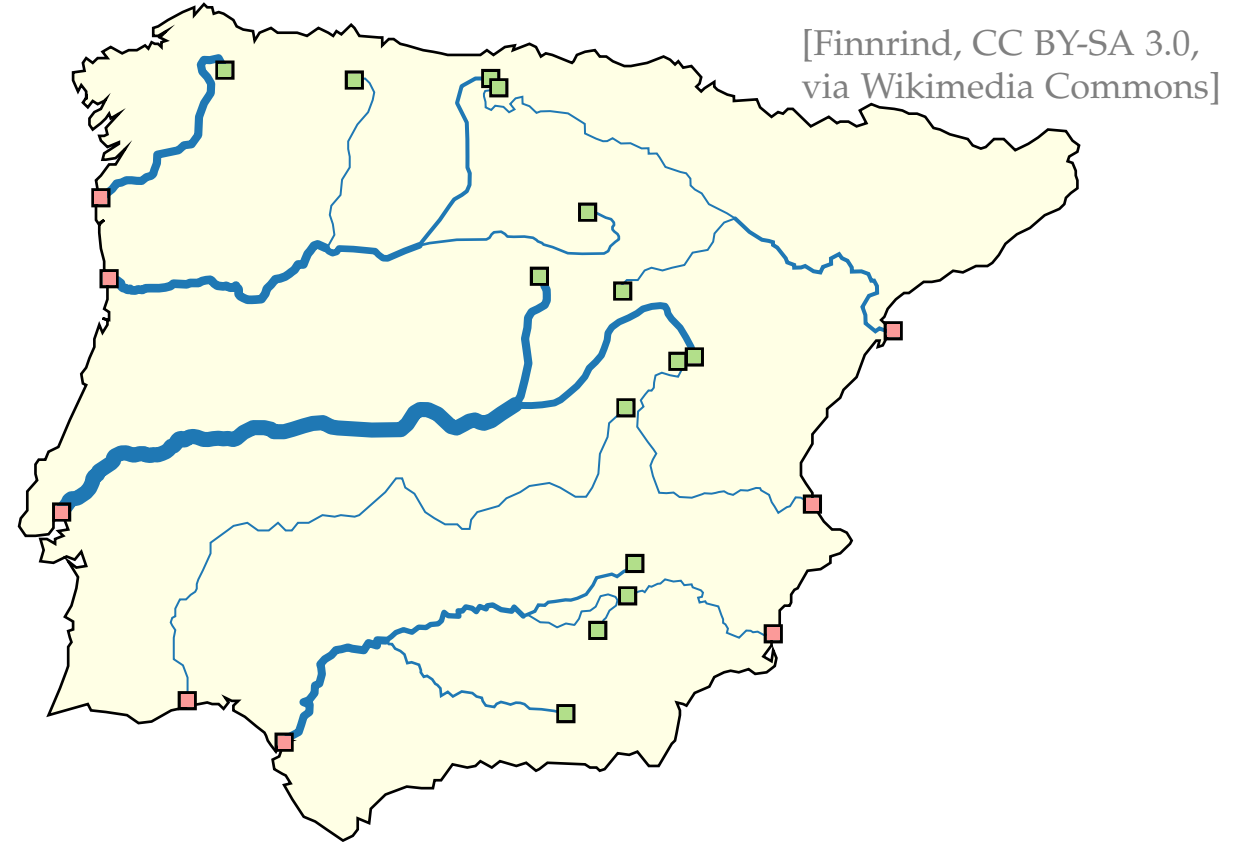




# Flow Networks

**Flow network**  $(G = (V, E); S, T; u)$  with

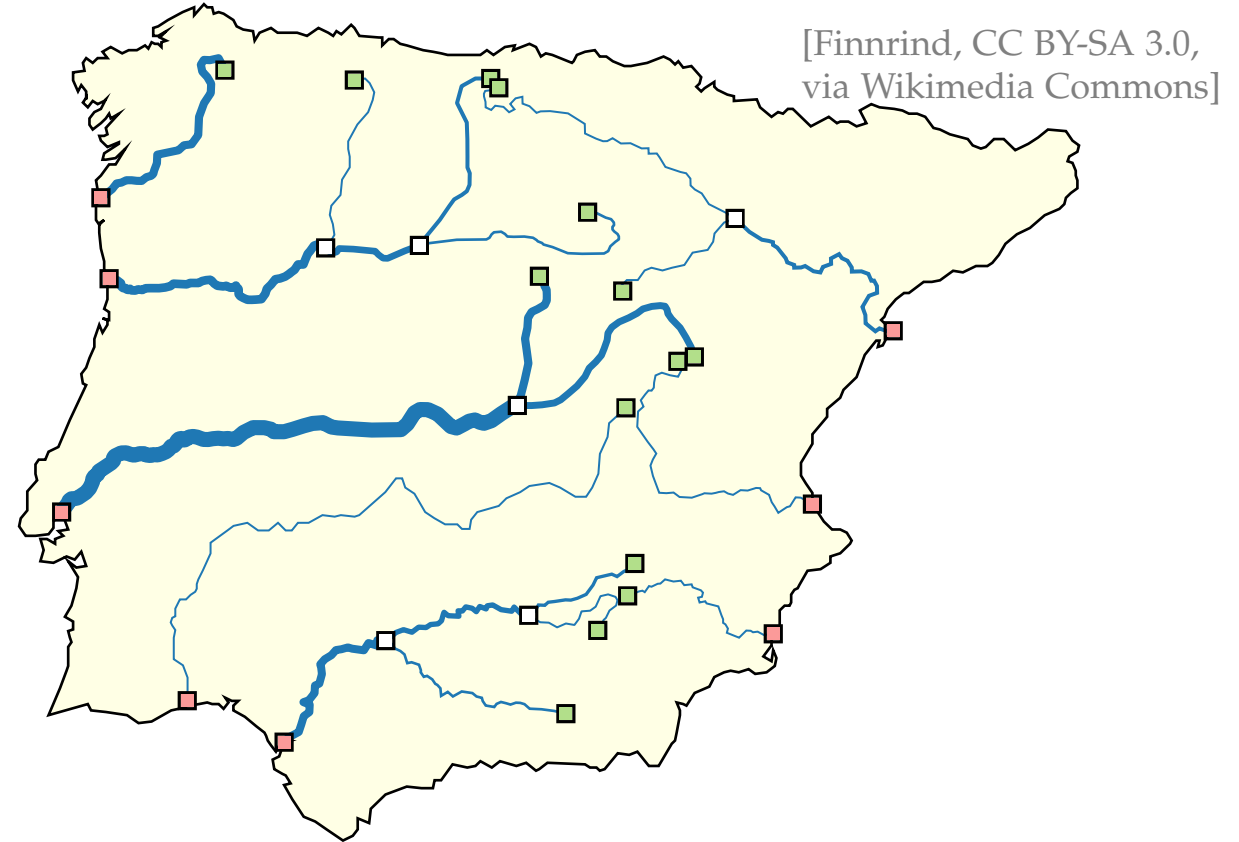
- directed graph  $G = (V, E)$
- *sources*  $S \subseteq V$ , *sinks*  $T \subseteq V$



# Flow Networks

**Flow network**  $(G = (V, E); S, T; u)$  with

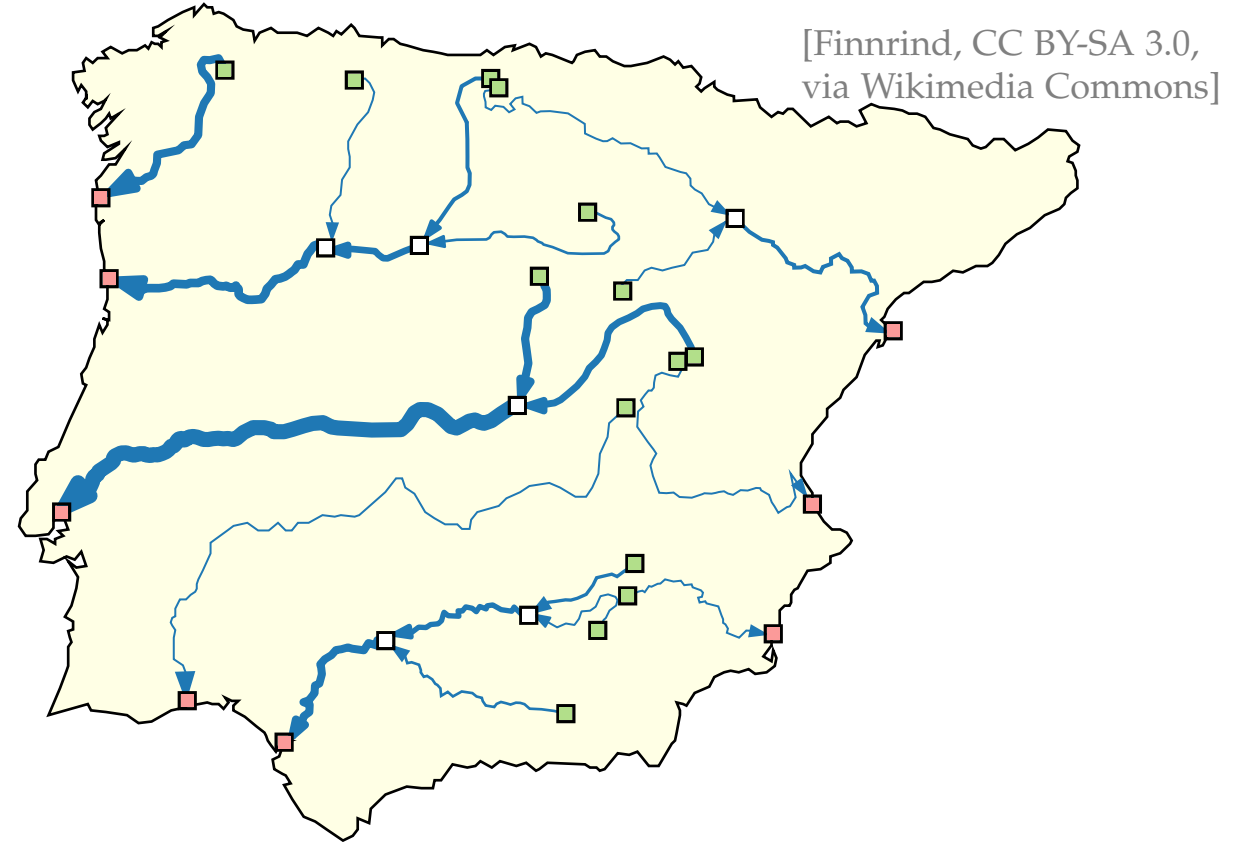
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# Flow Networks

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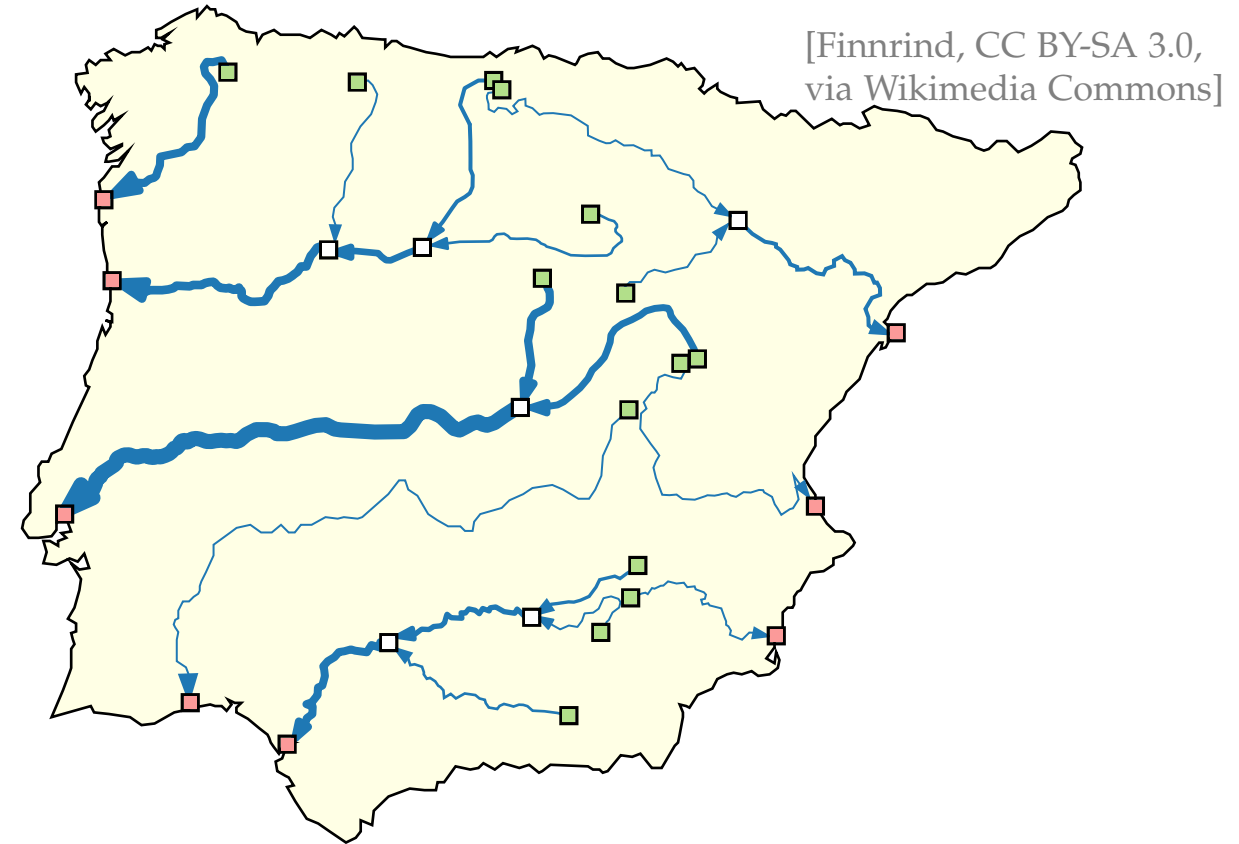
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# Flow Networks

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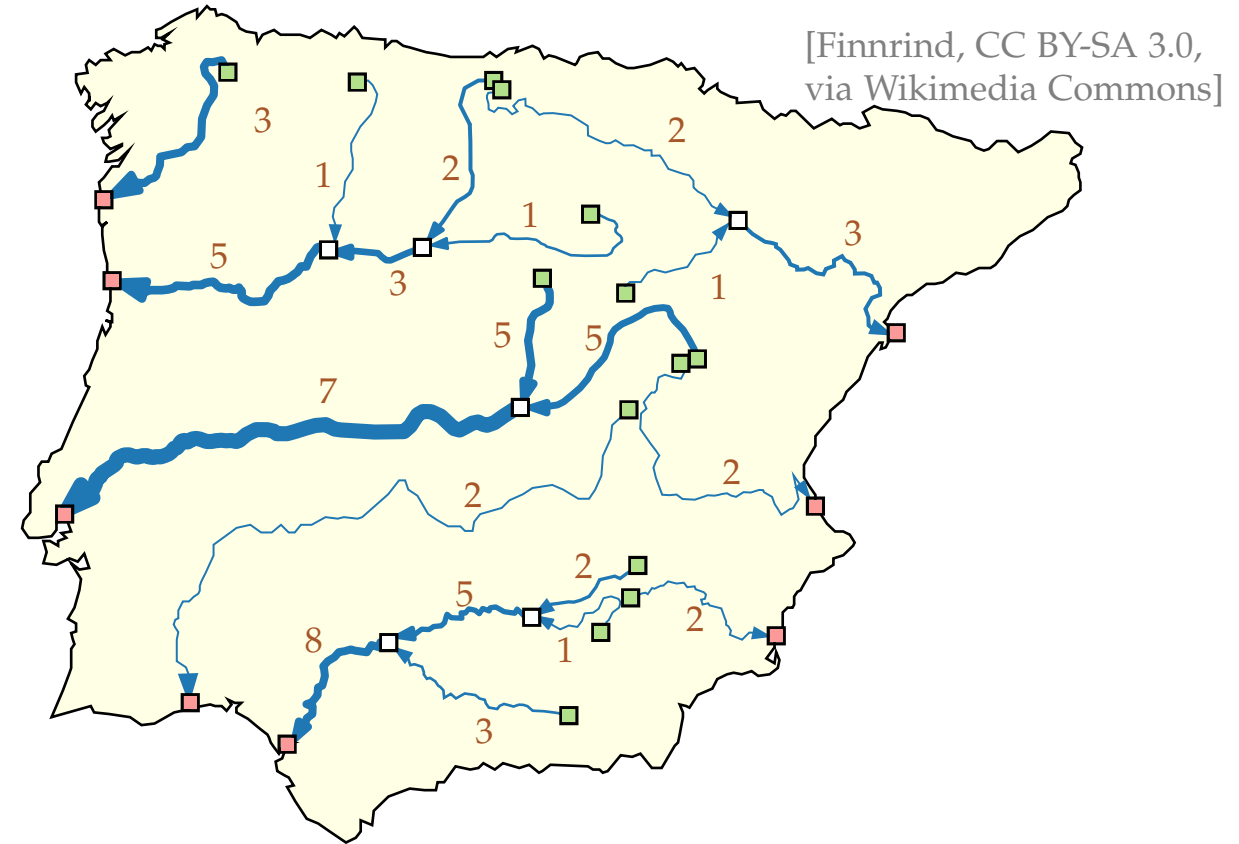
- directed graph  $G = (V, E)$
- *sources*  $S \subseteq V$ , *sinks*  $T \subseteq V$
- edge *capacity*  $u: E \rightarrow \mathbb{R}_0^+$



# Flow Networks

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- *sources*  $S \subseteq V$ , *sinks*  $T \subseteq V$
- edge *capacity*  $u: E \rightarrow \mathbb{R}_0^+$

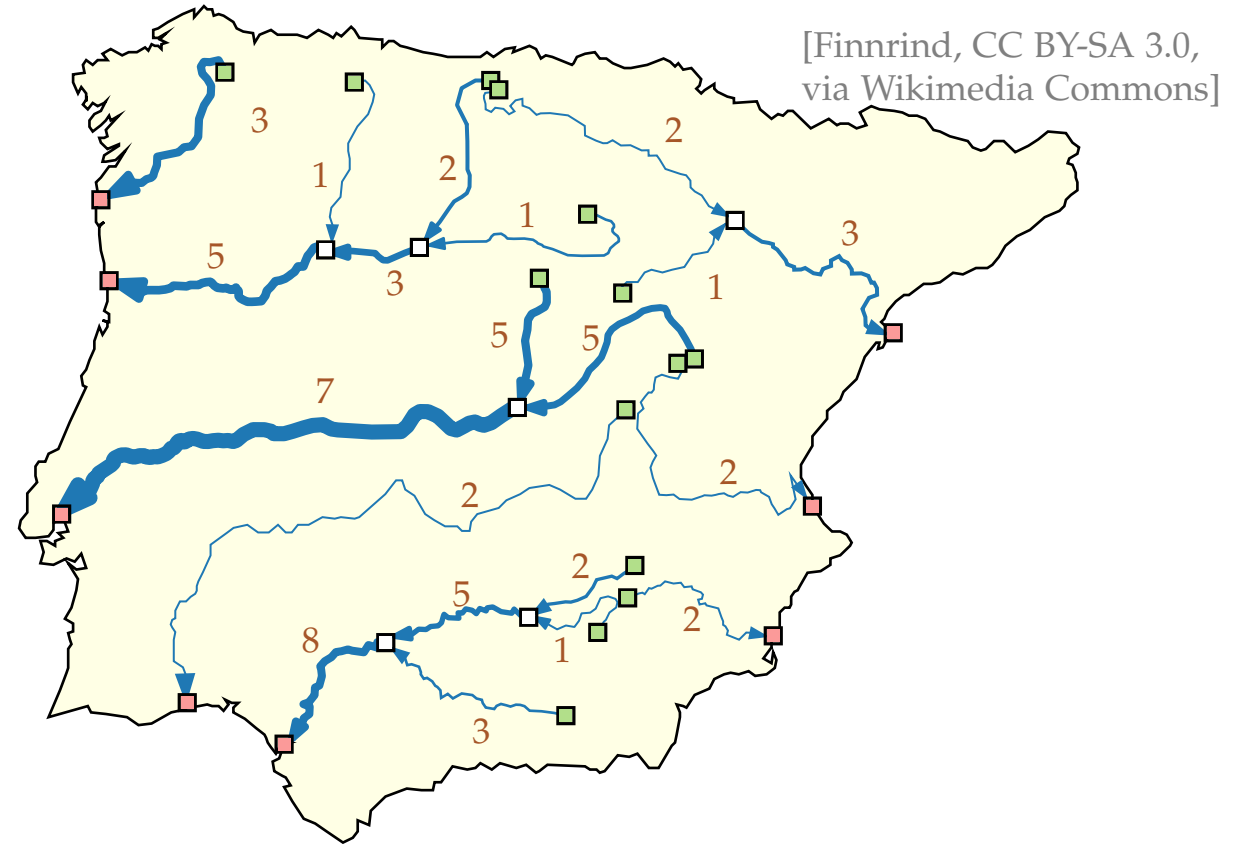


# Flow Networks

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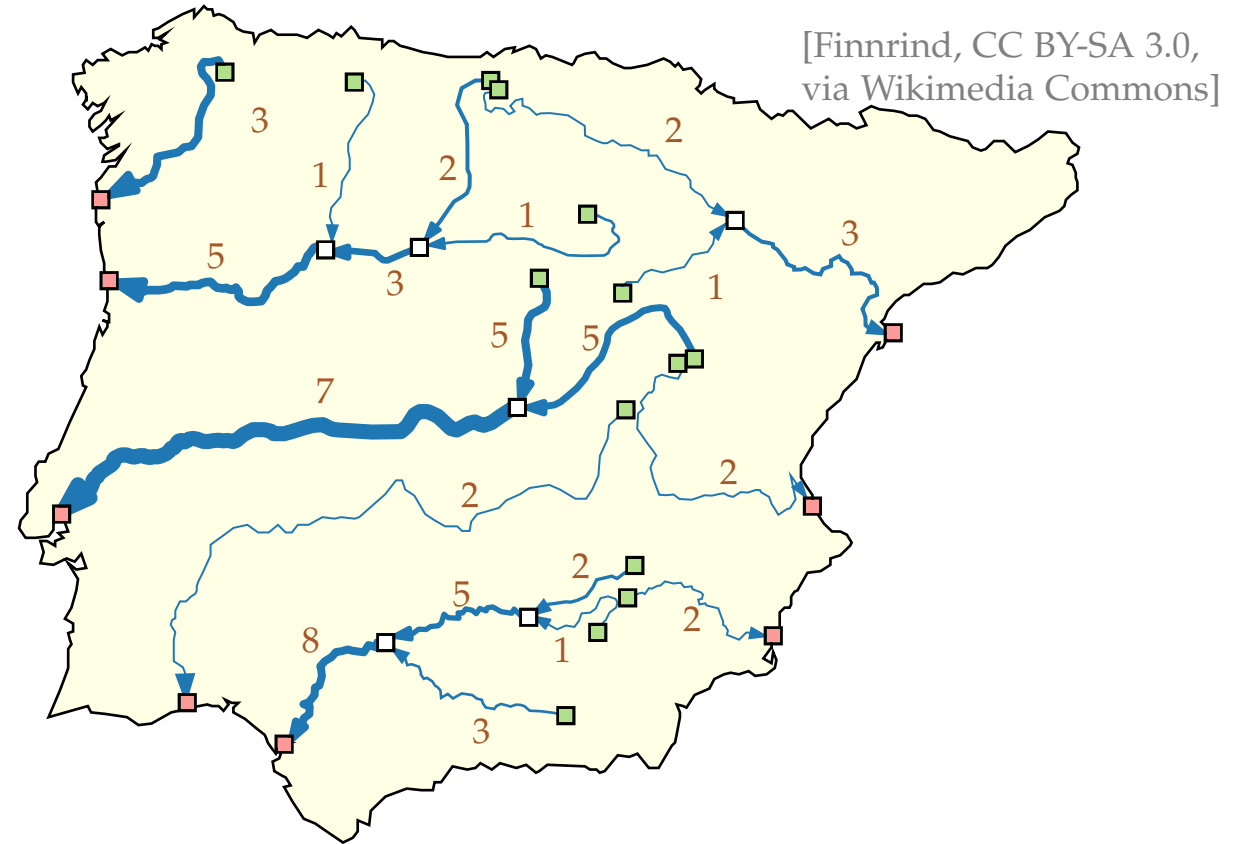
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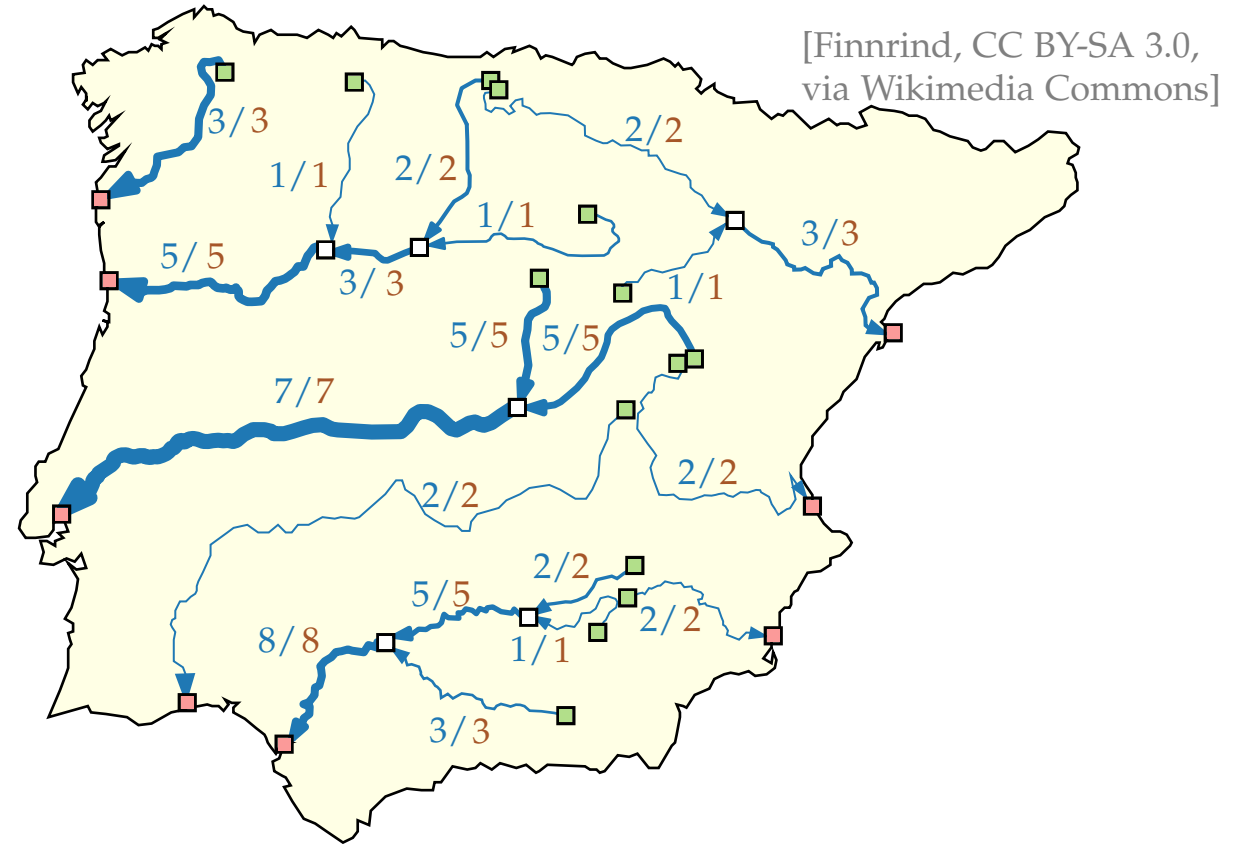
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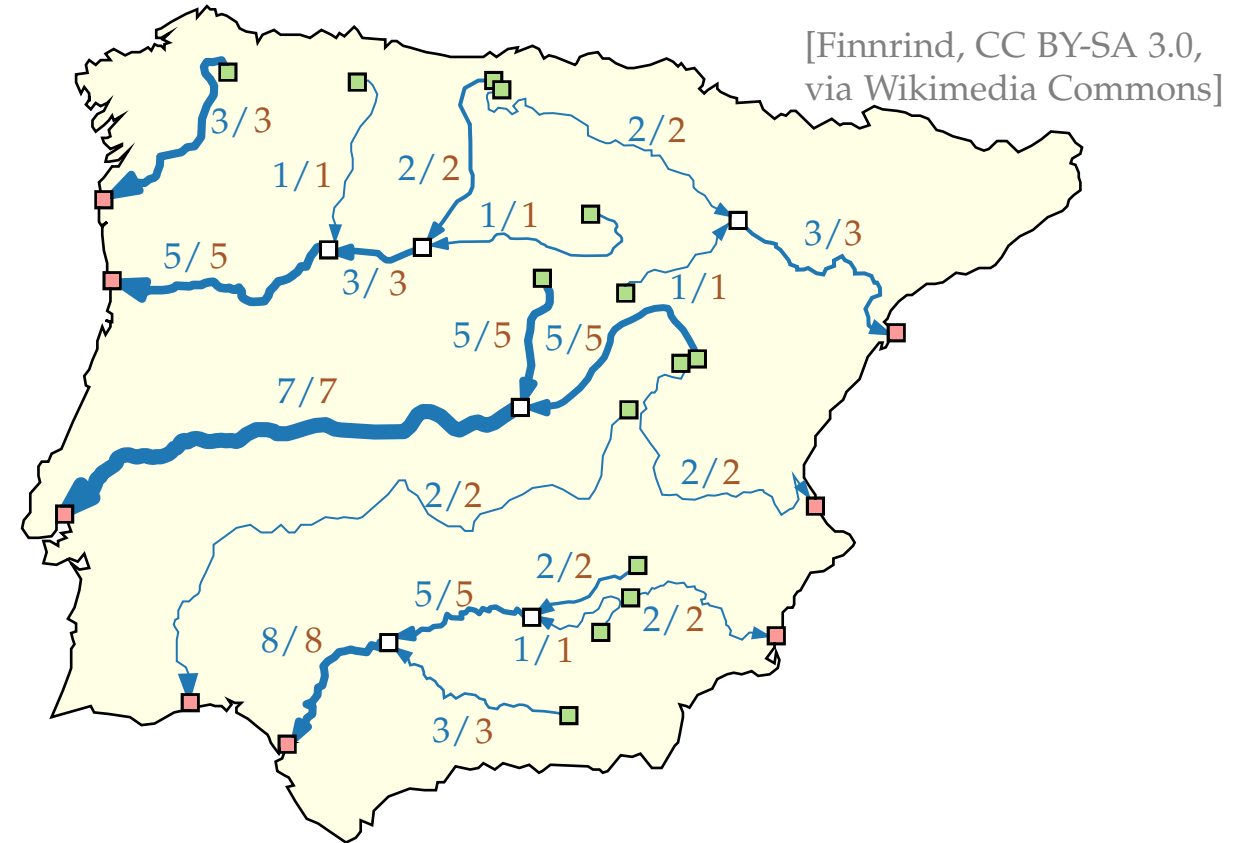
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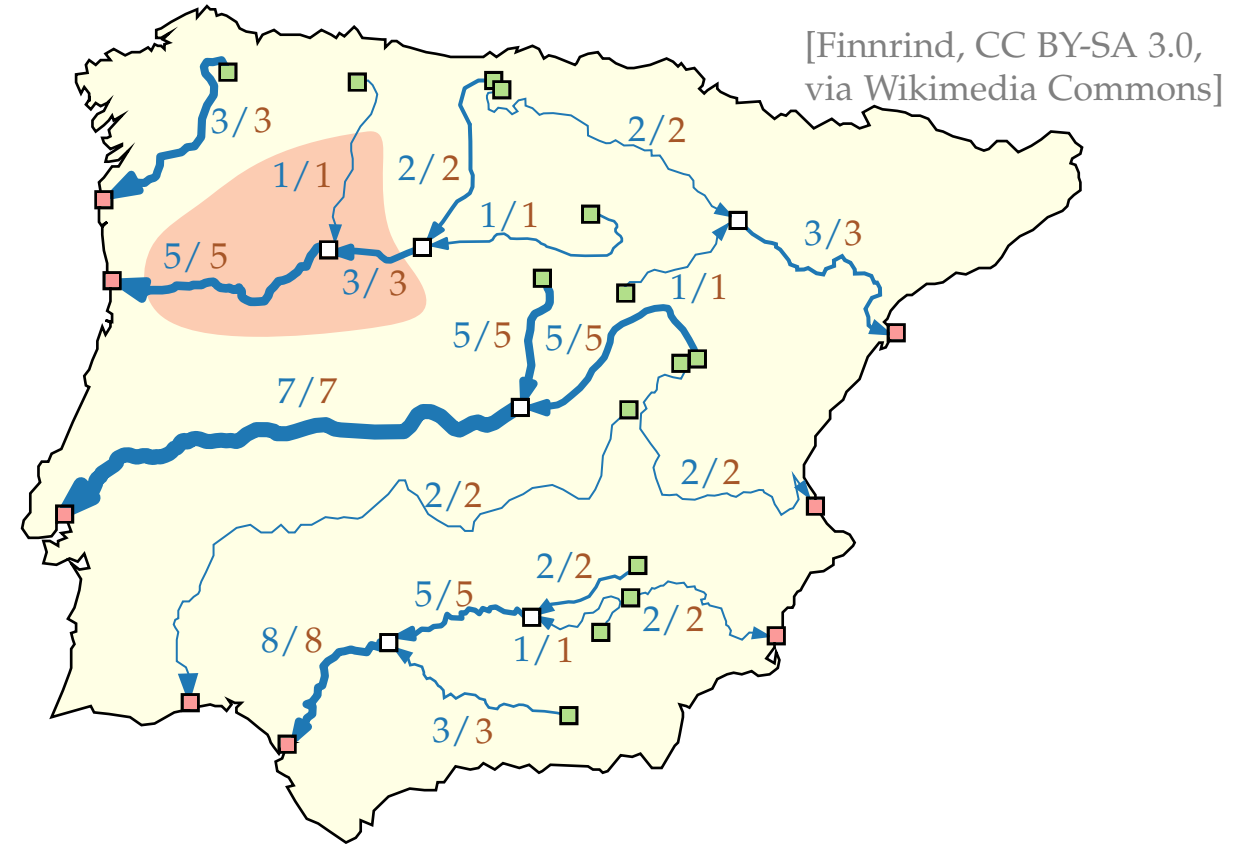
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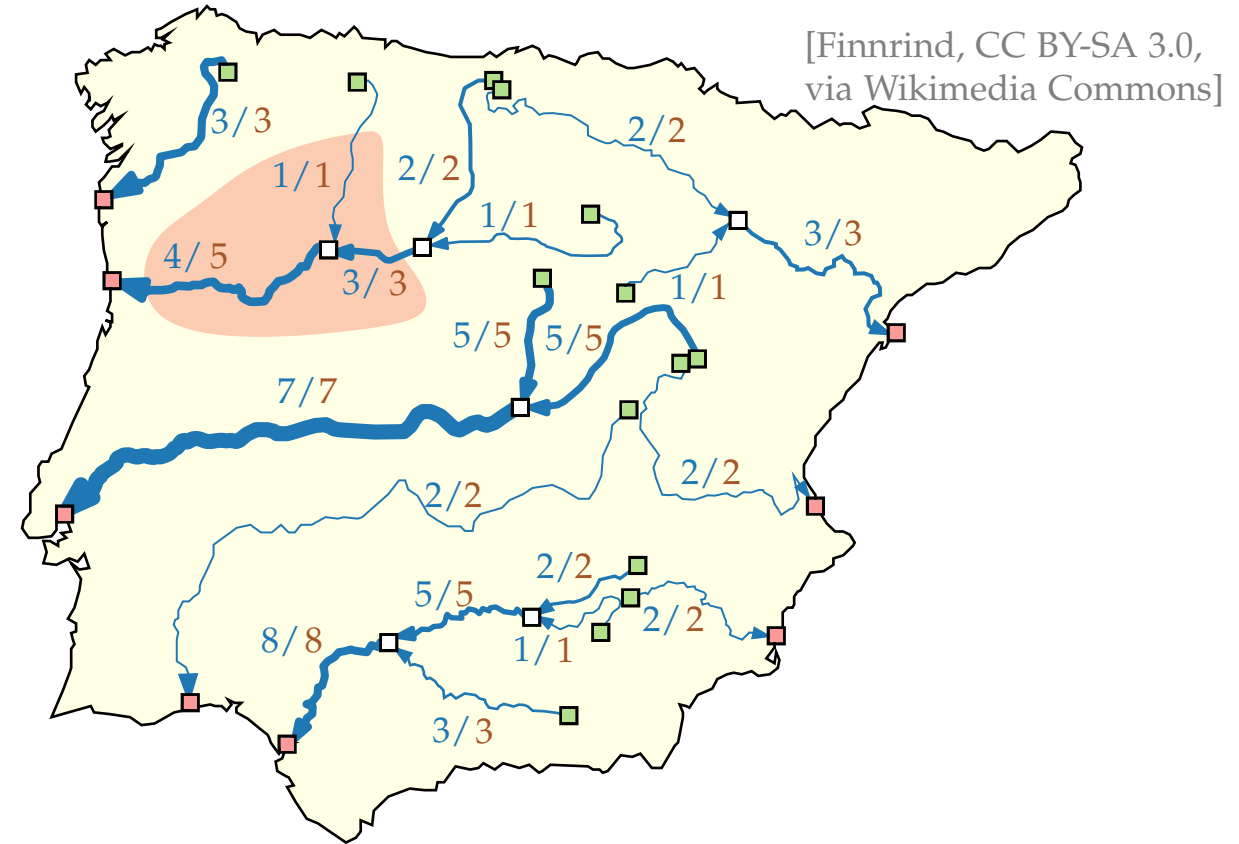
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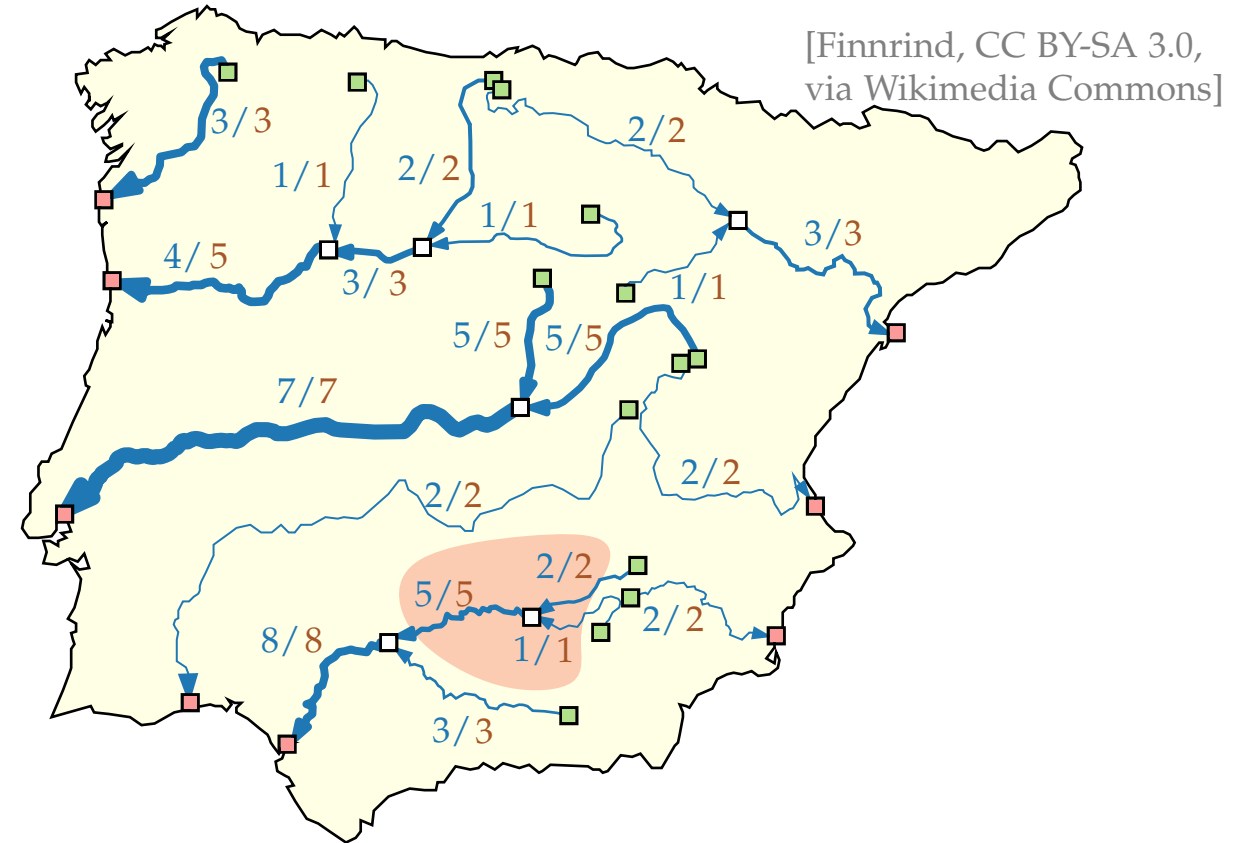
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# Flow Networks

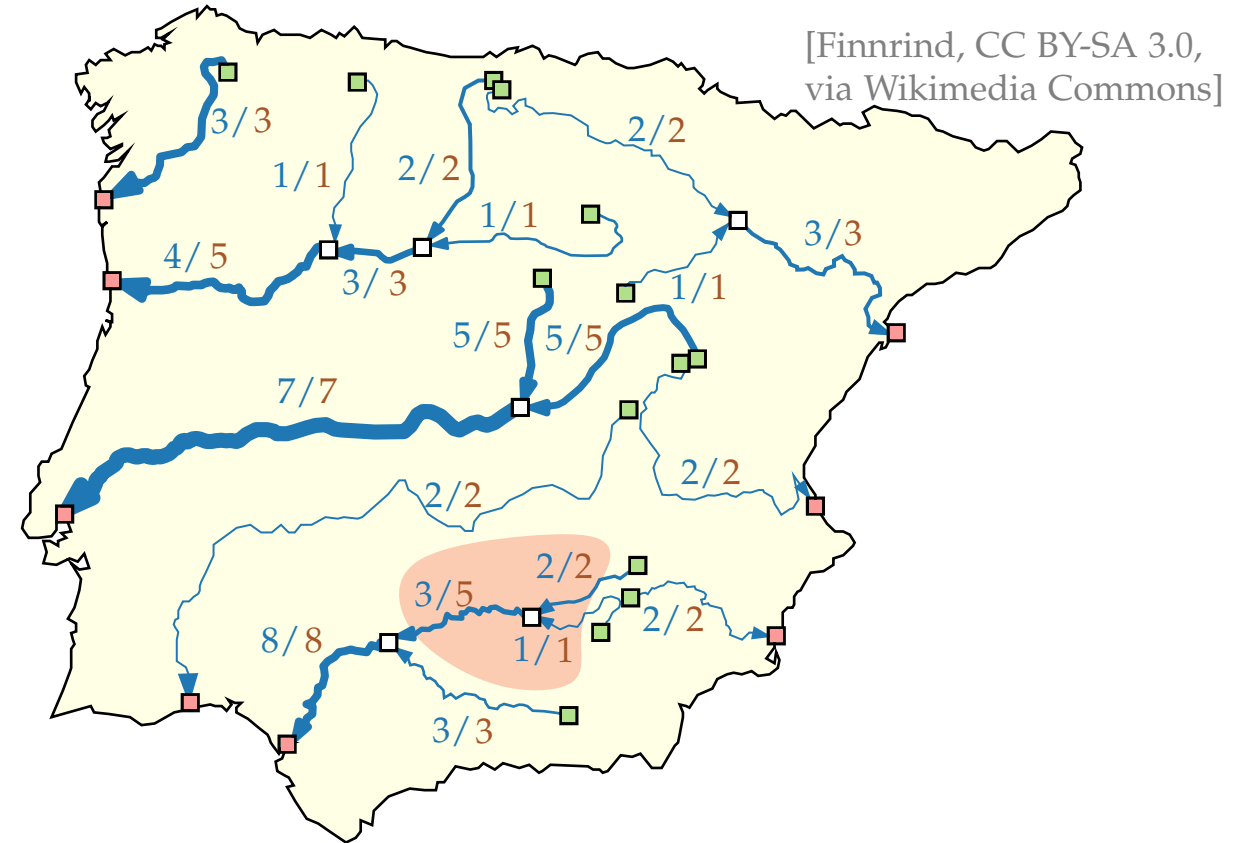
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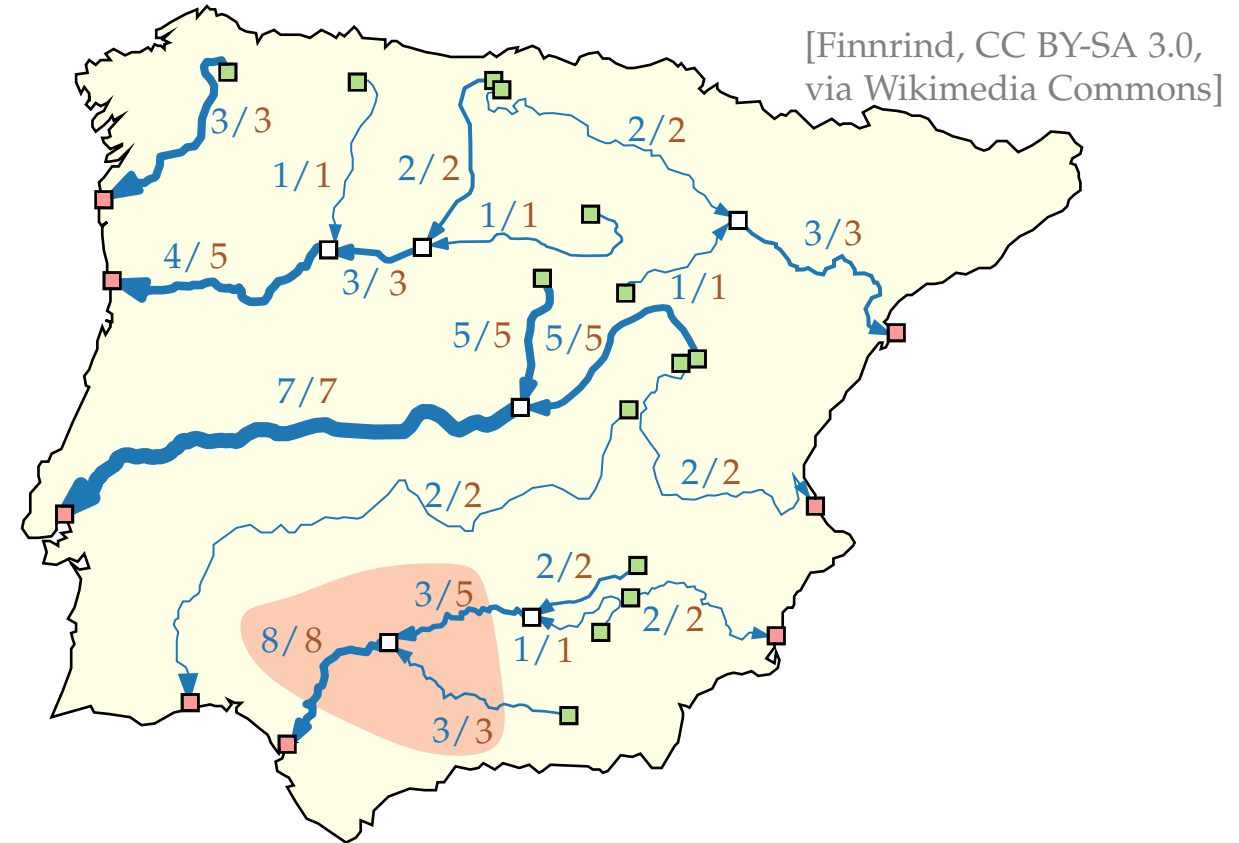
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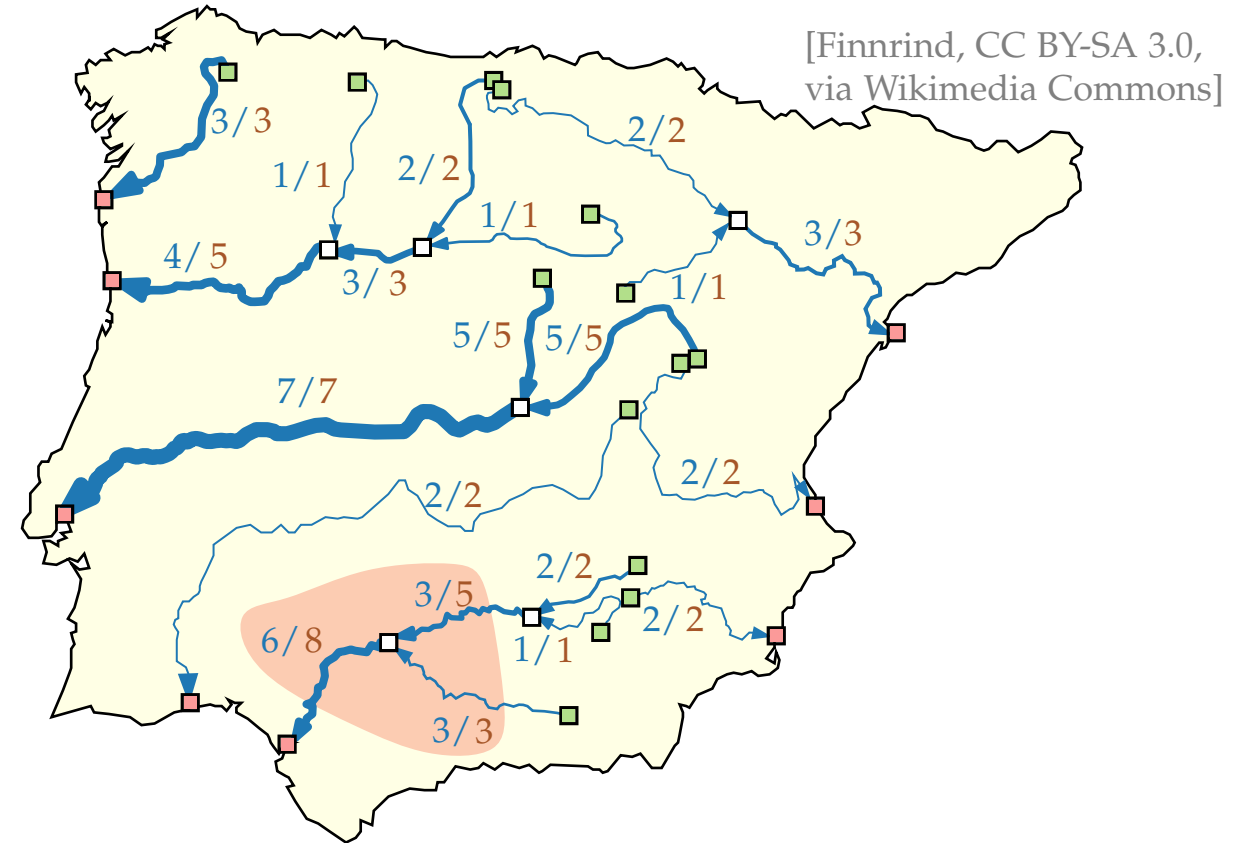
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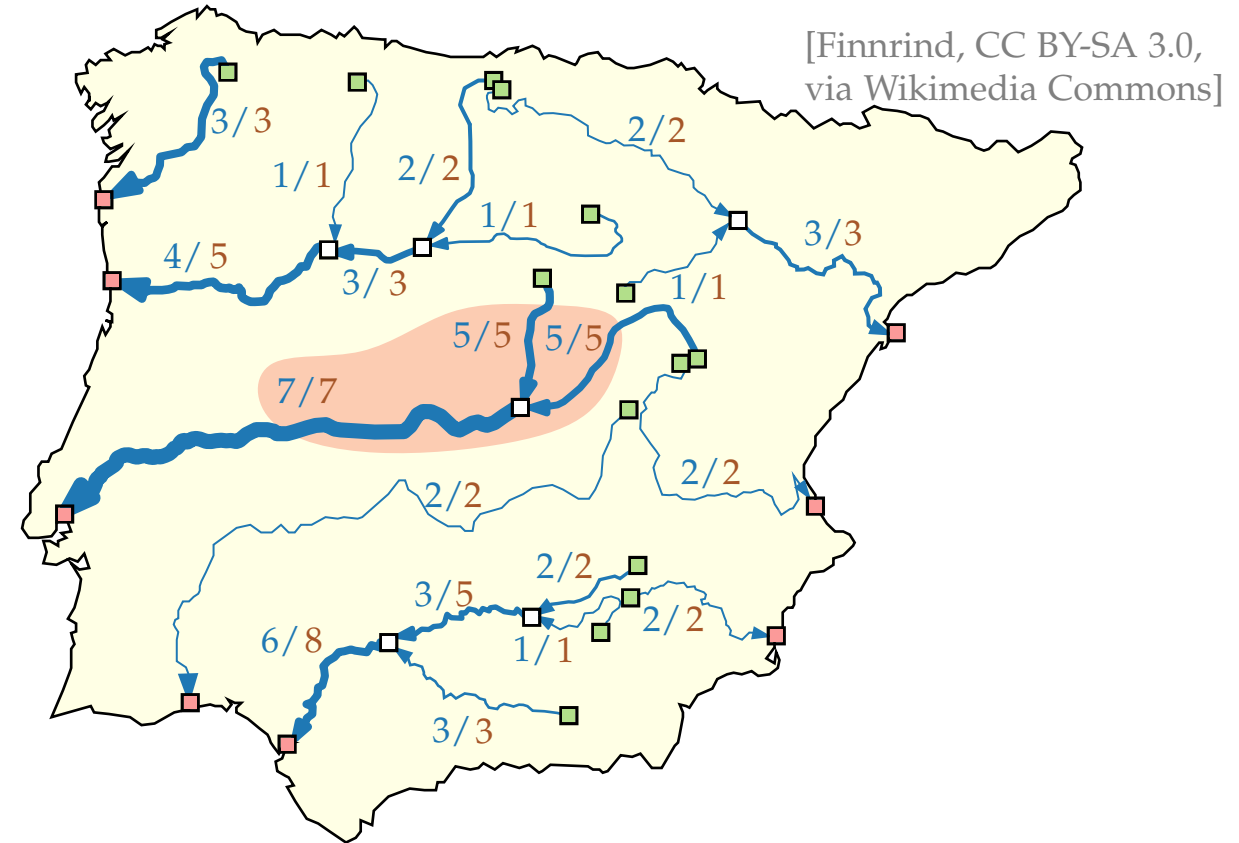
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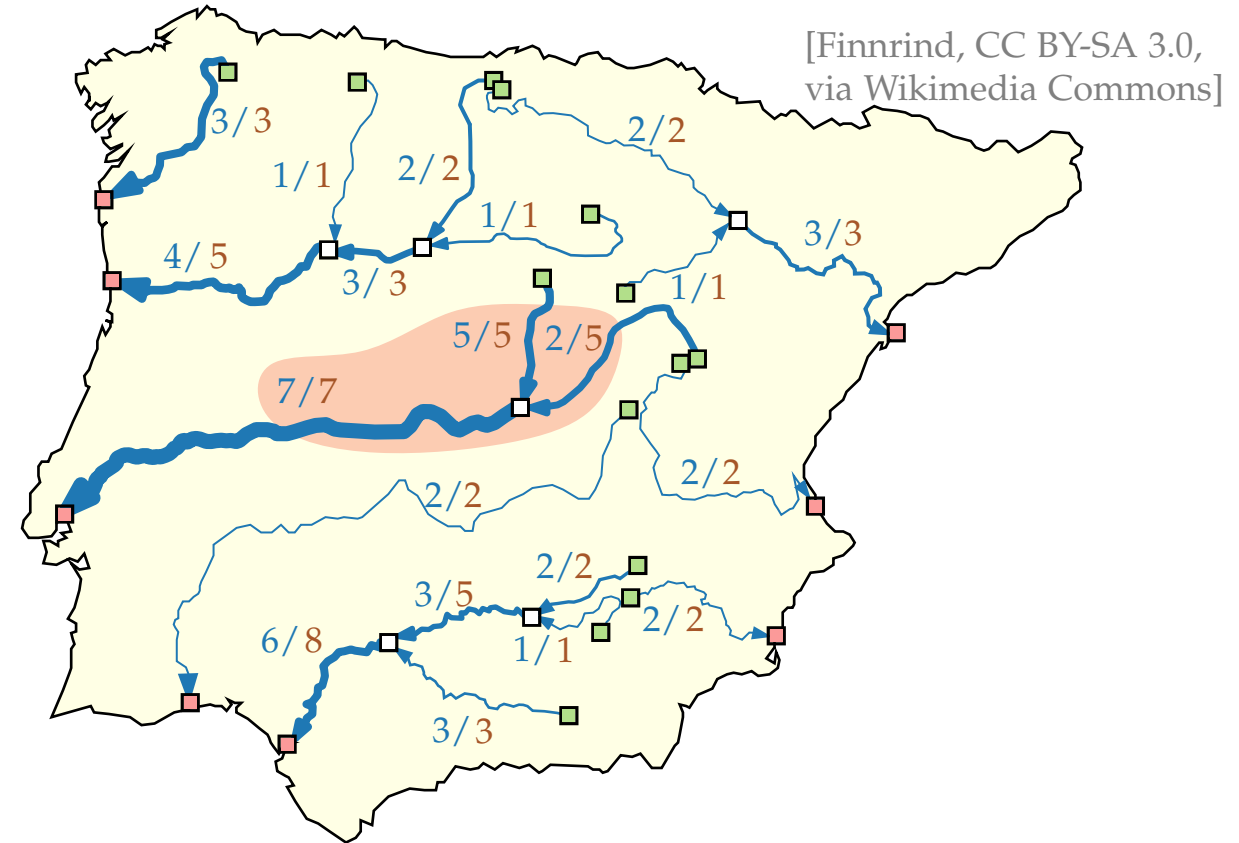
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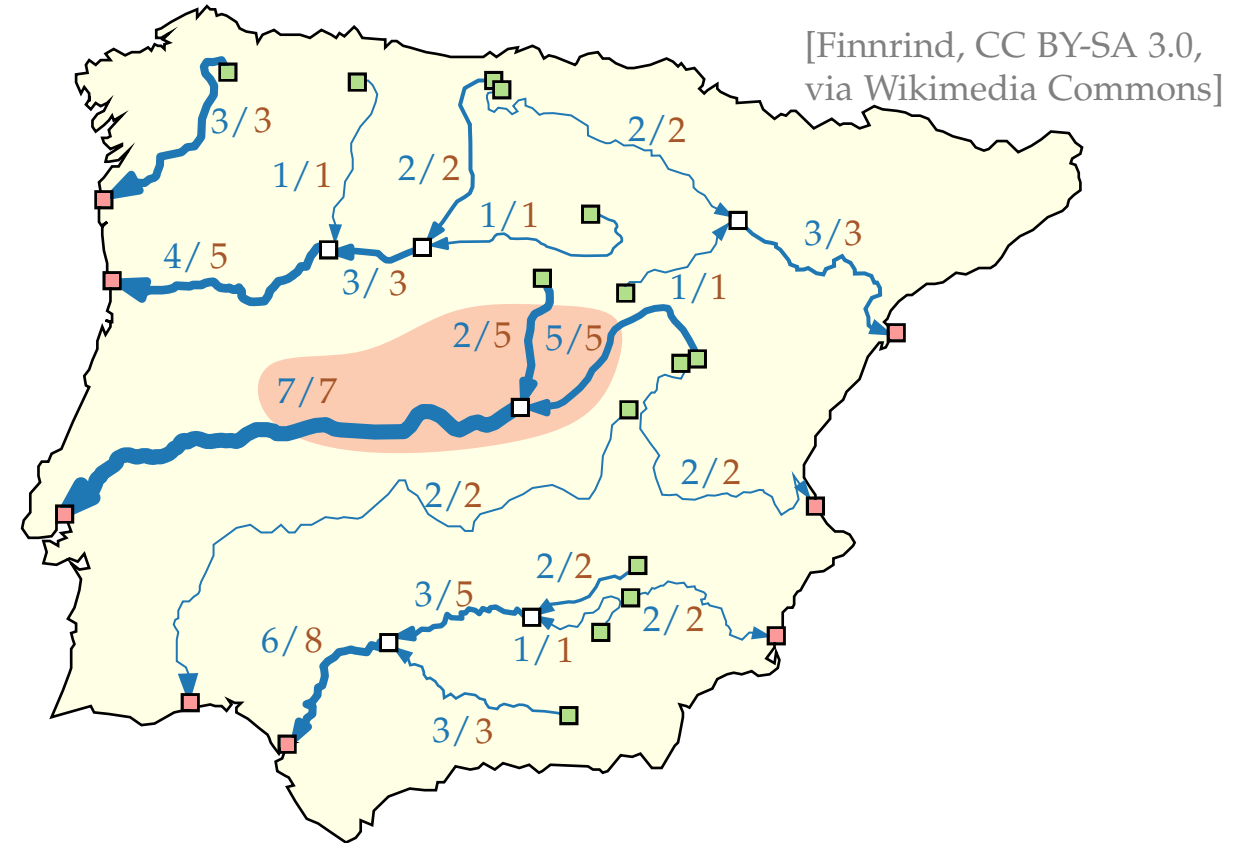
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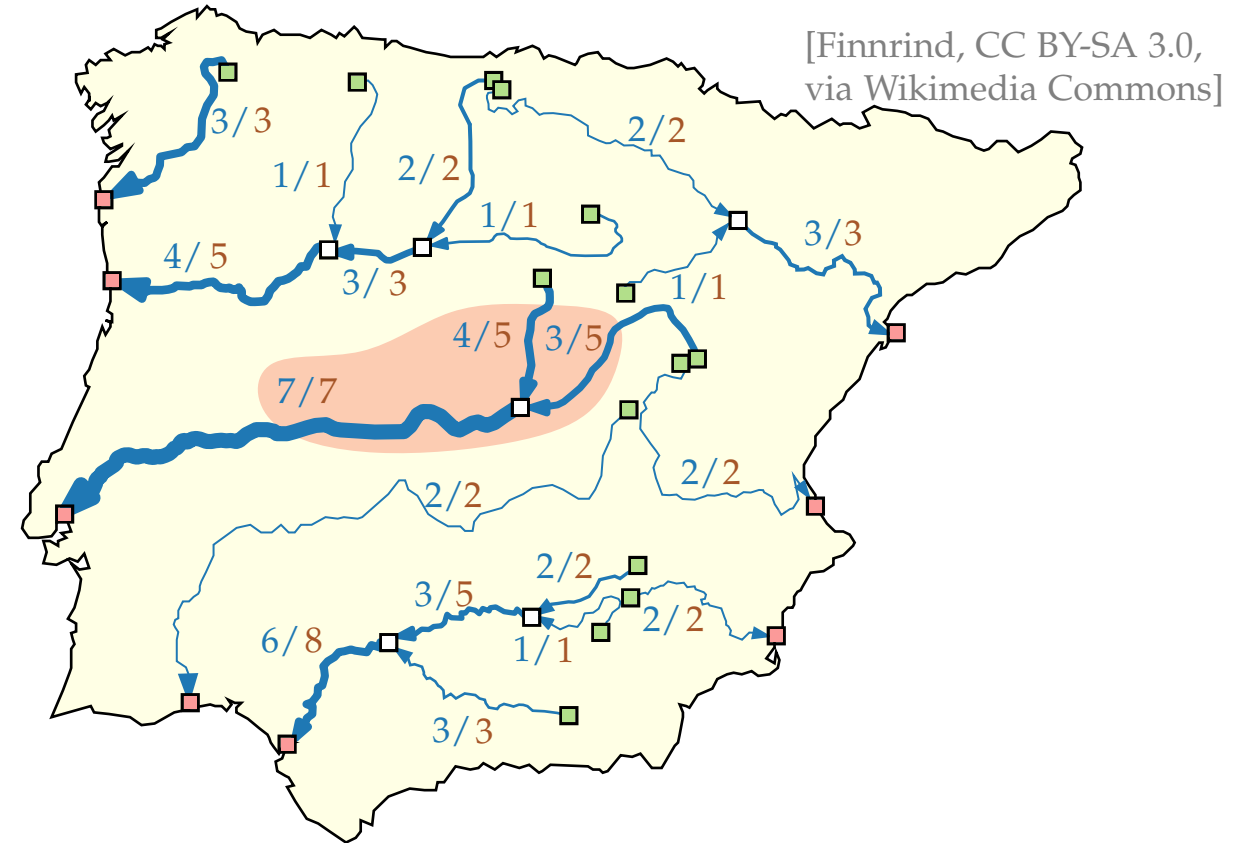
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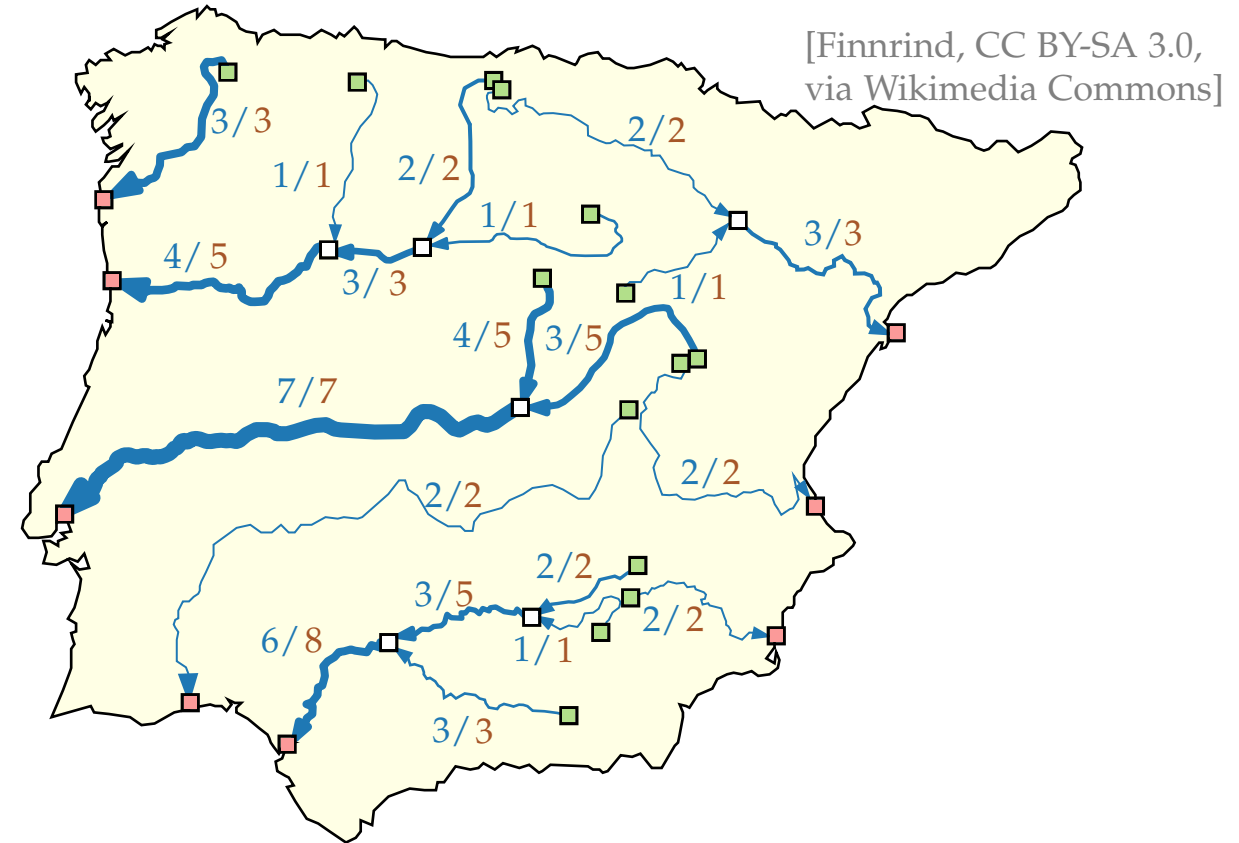
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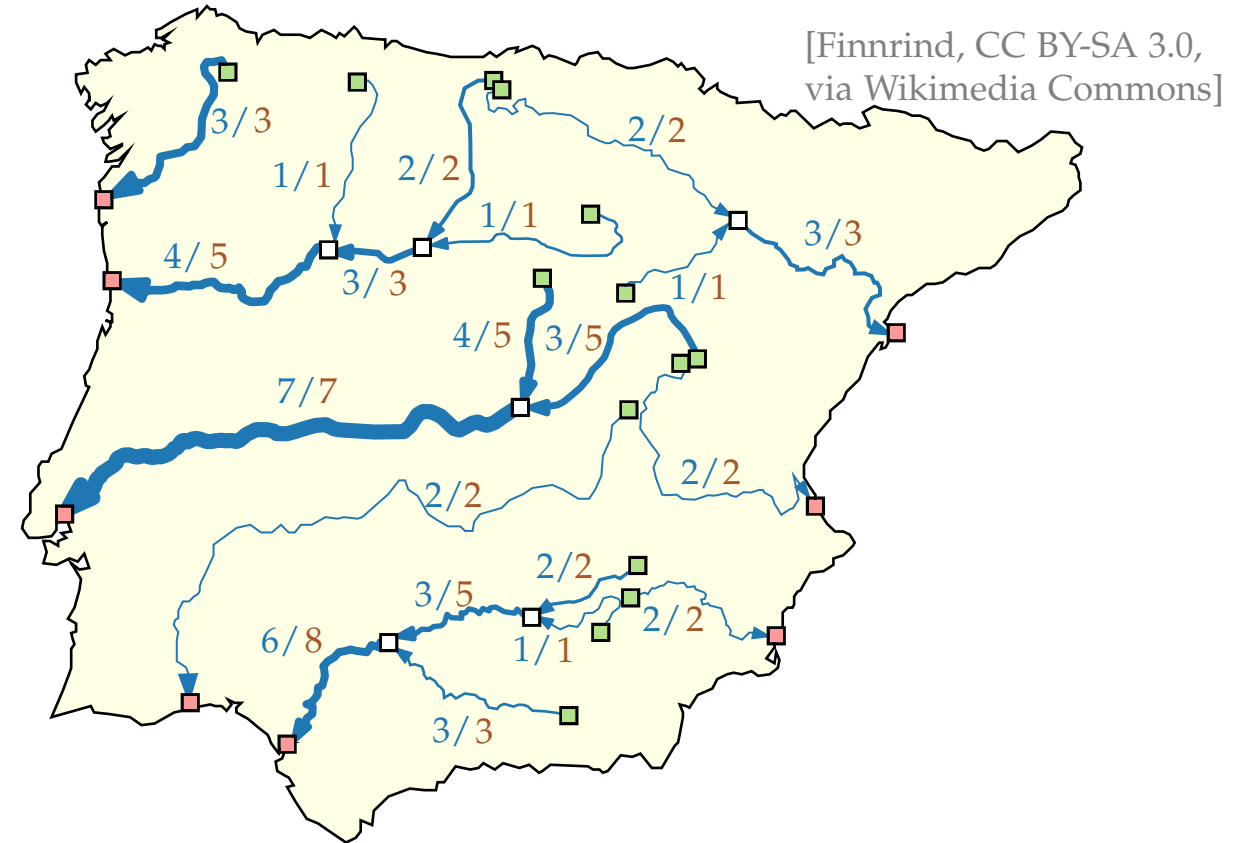
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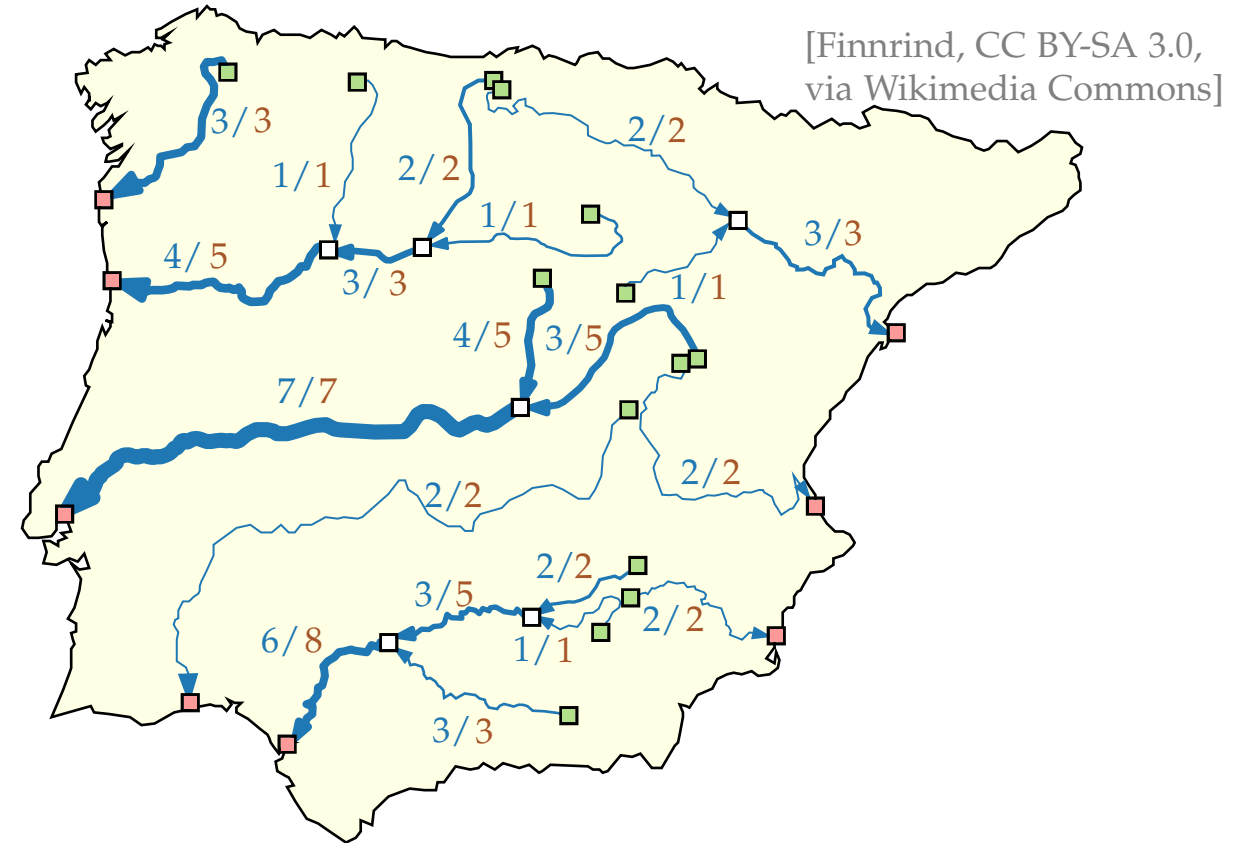
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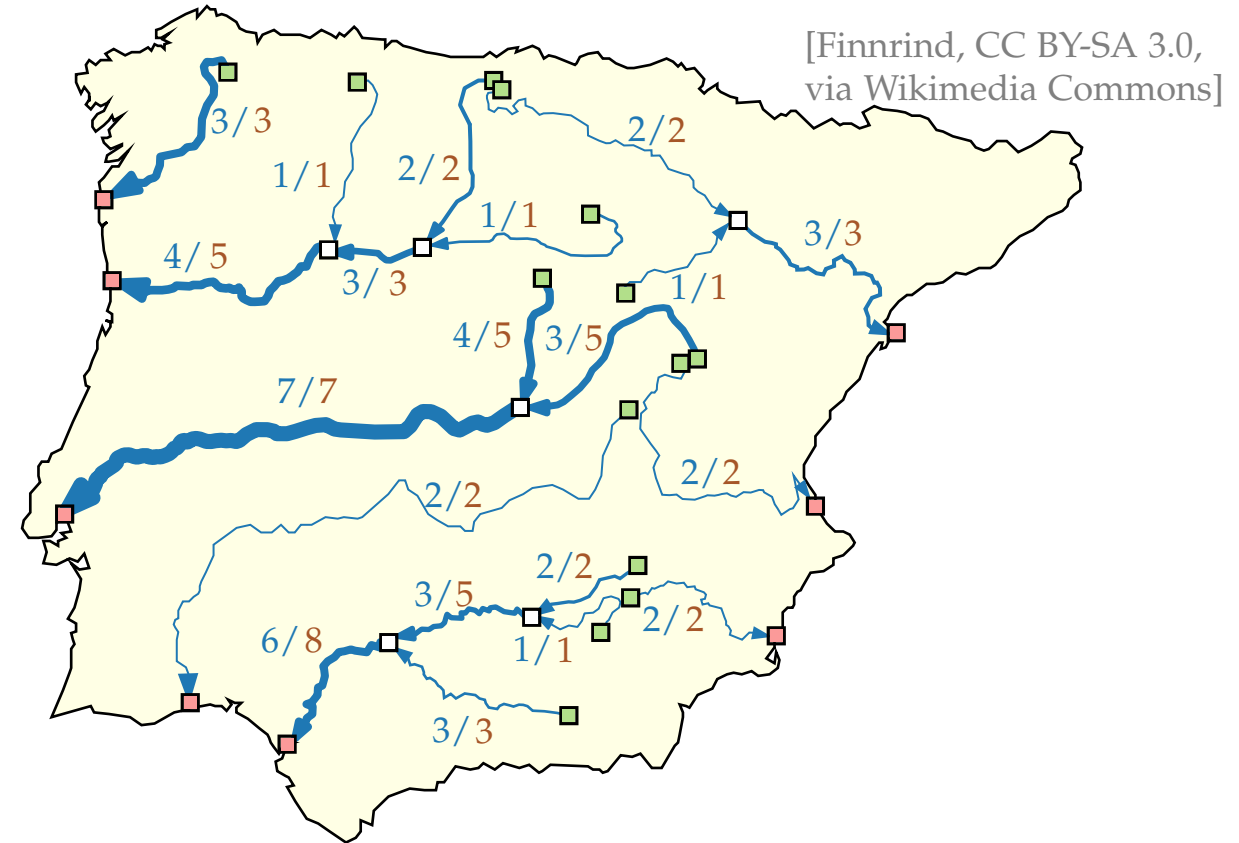
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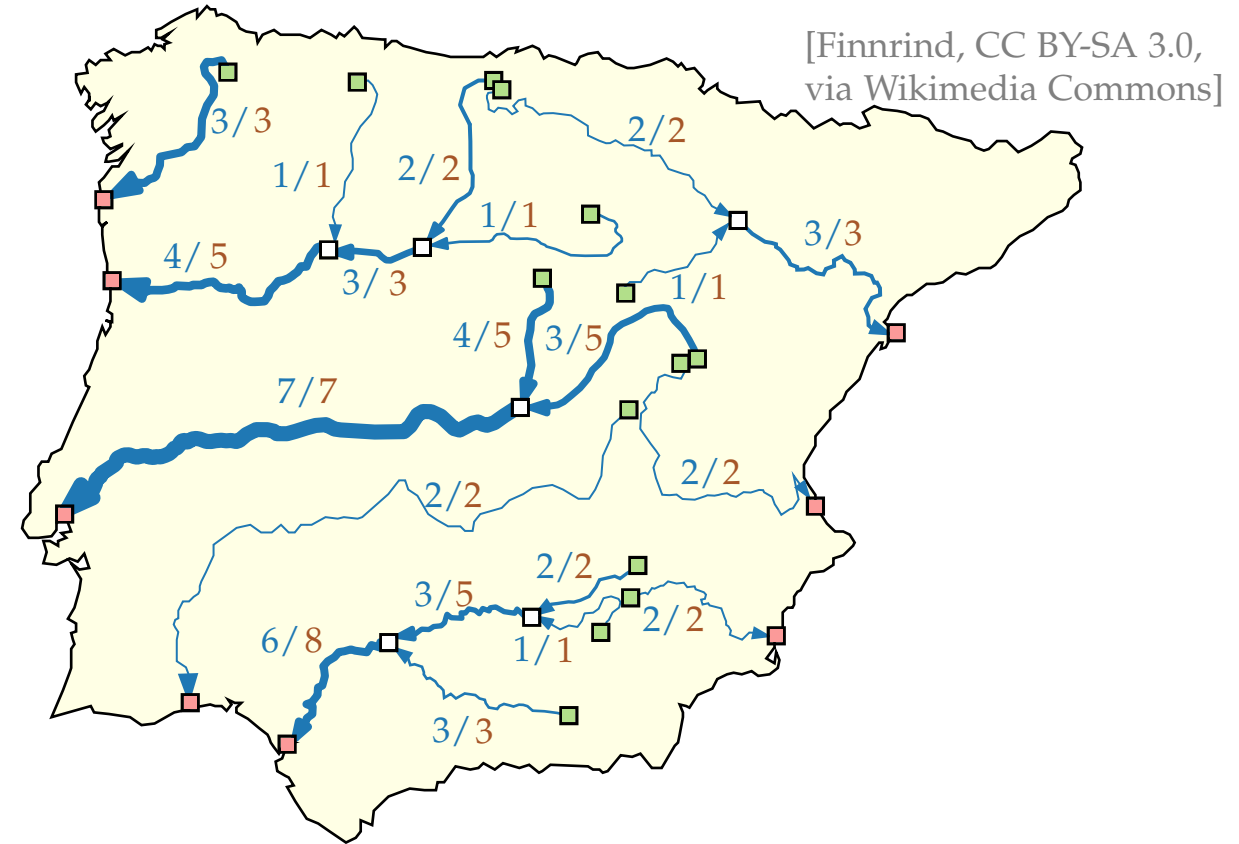
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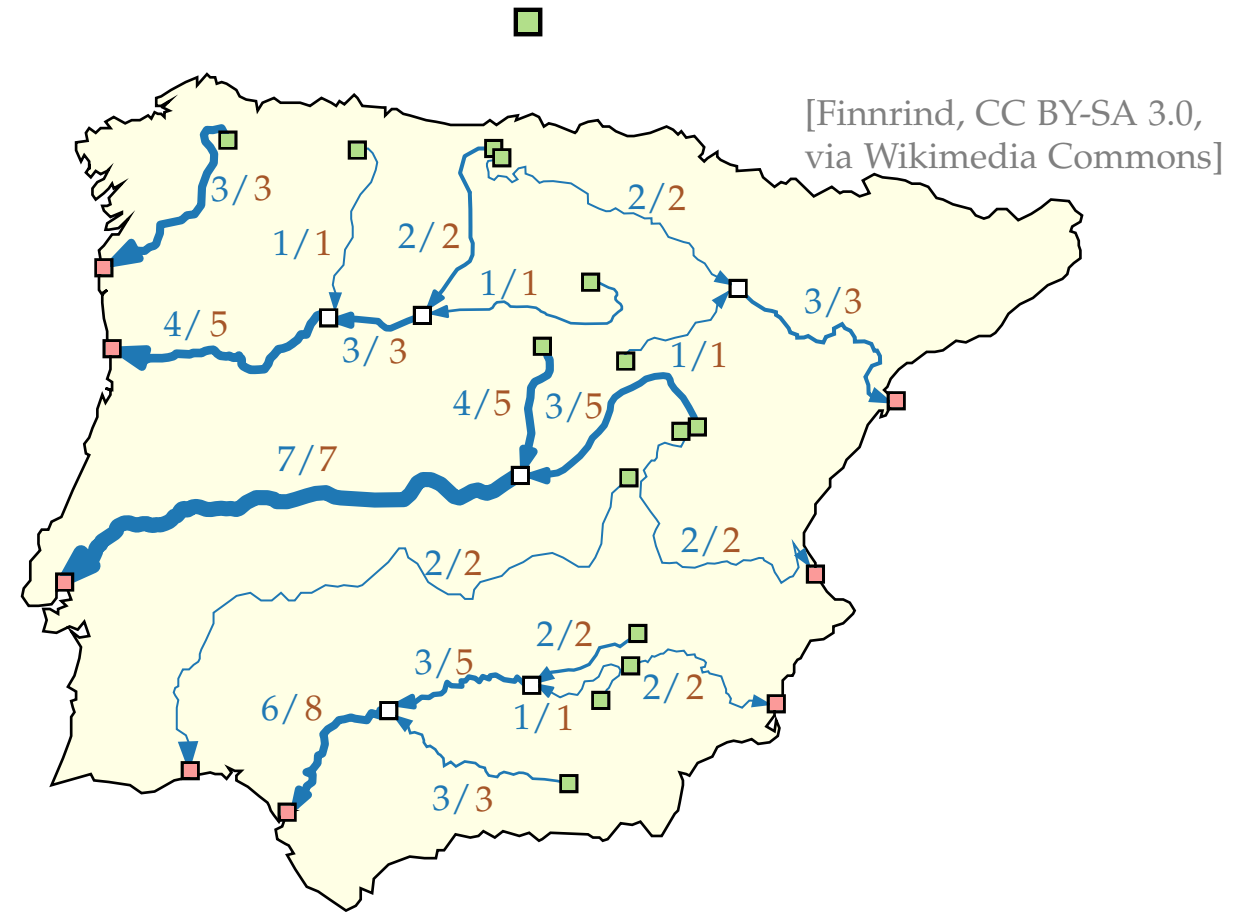
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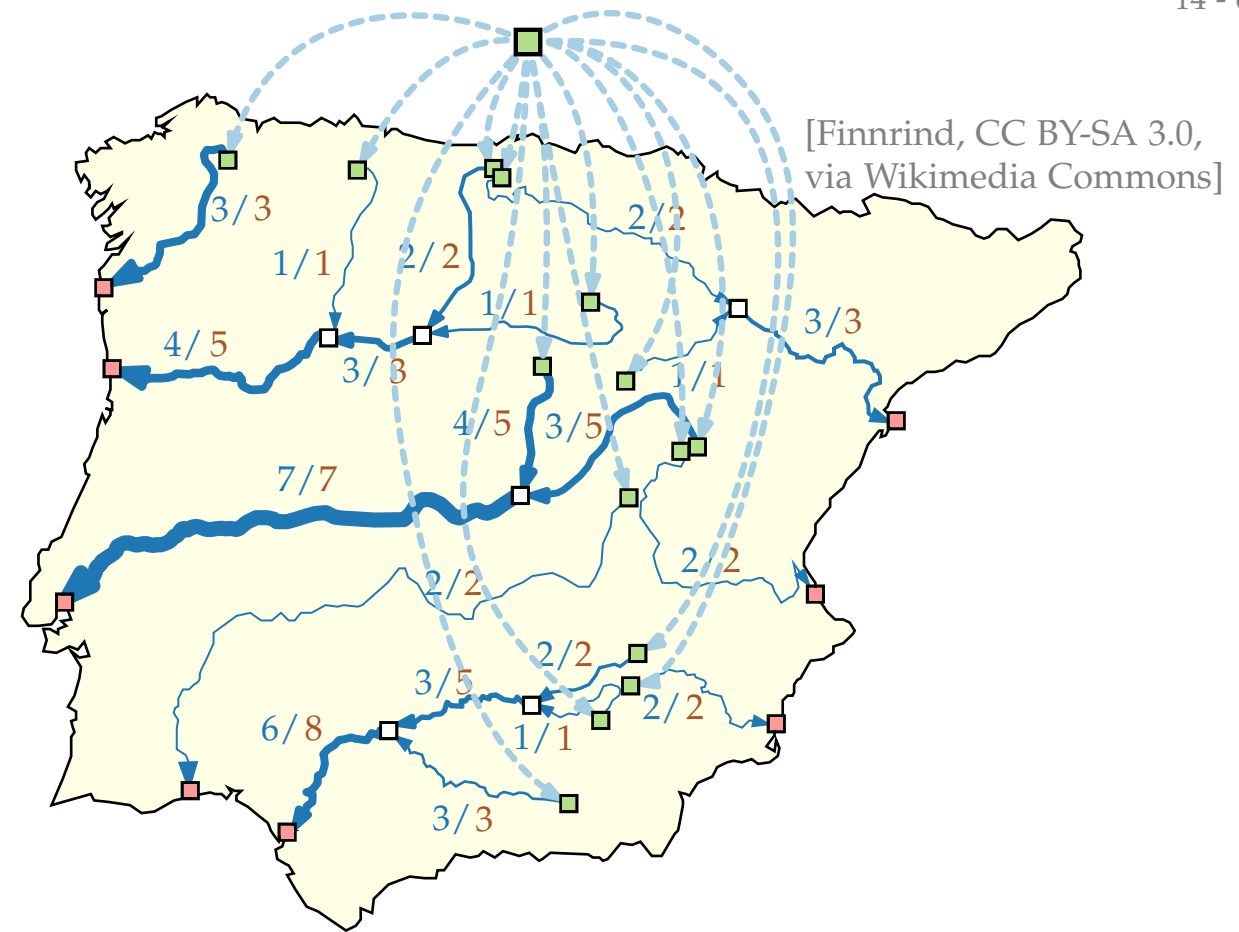
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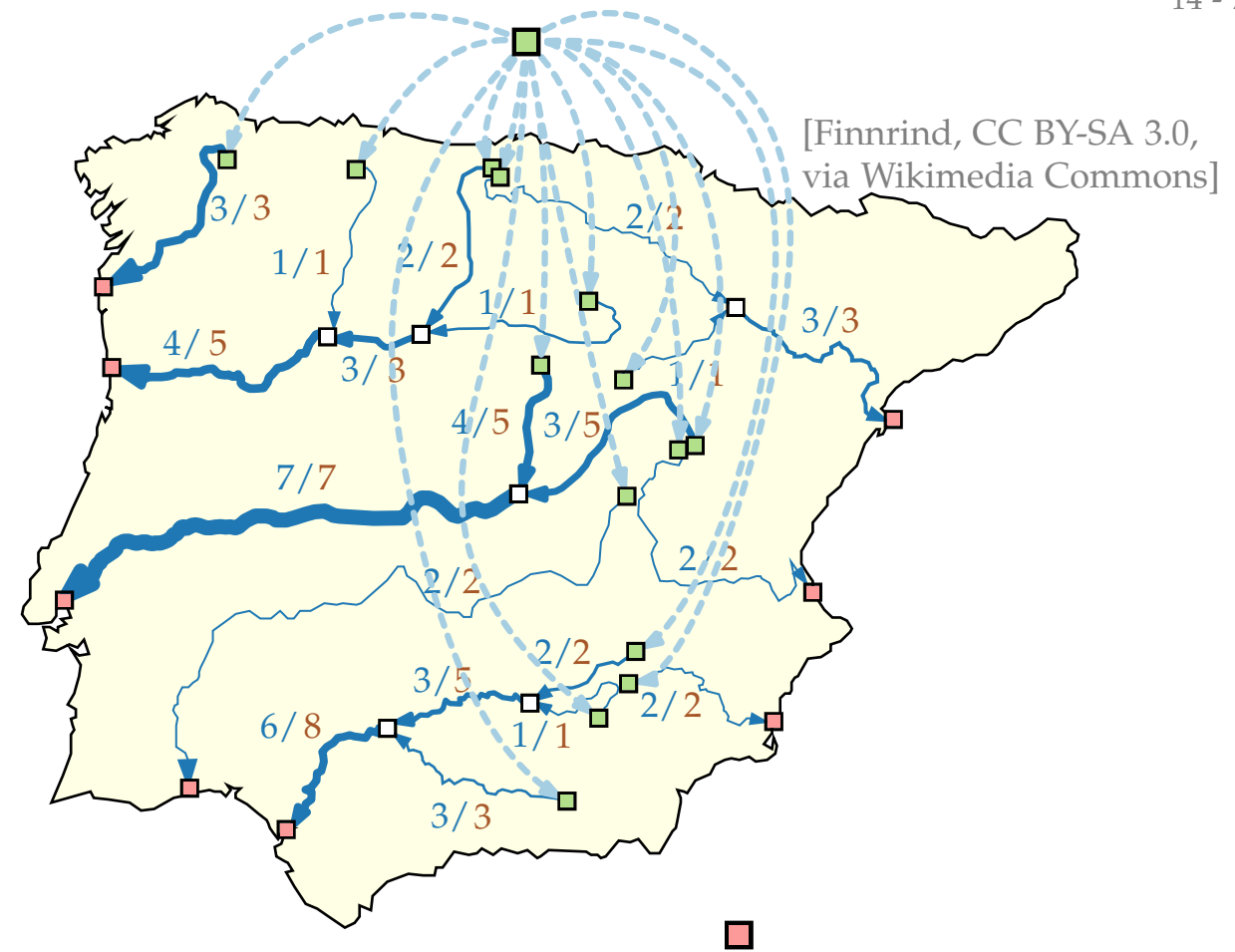
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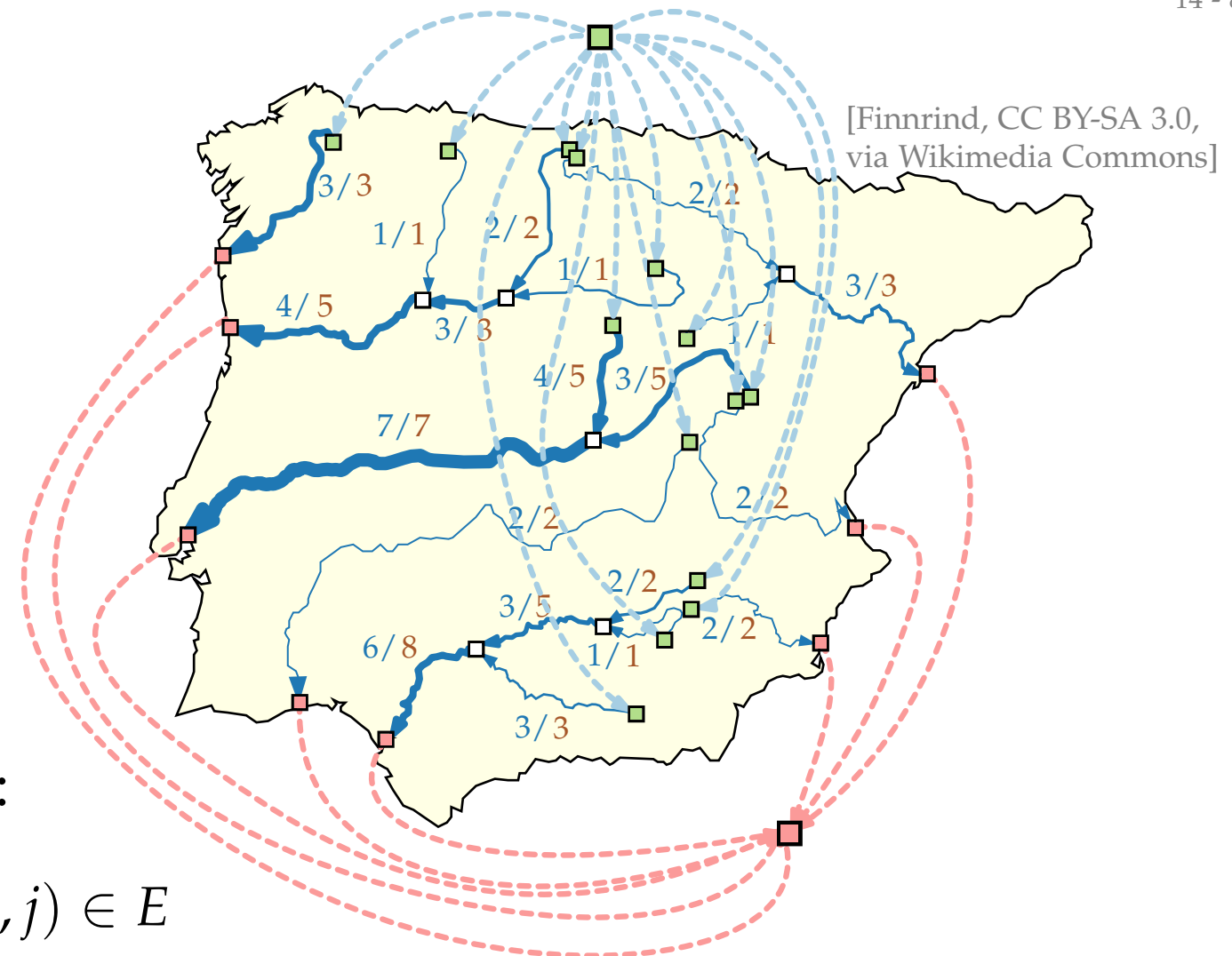
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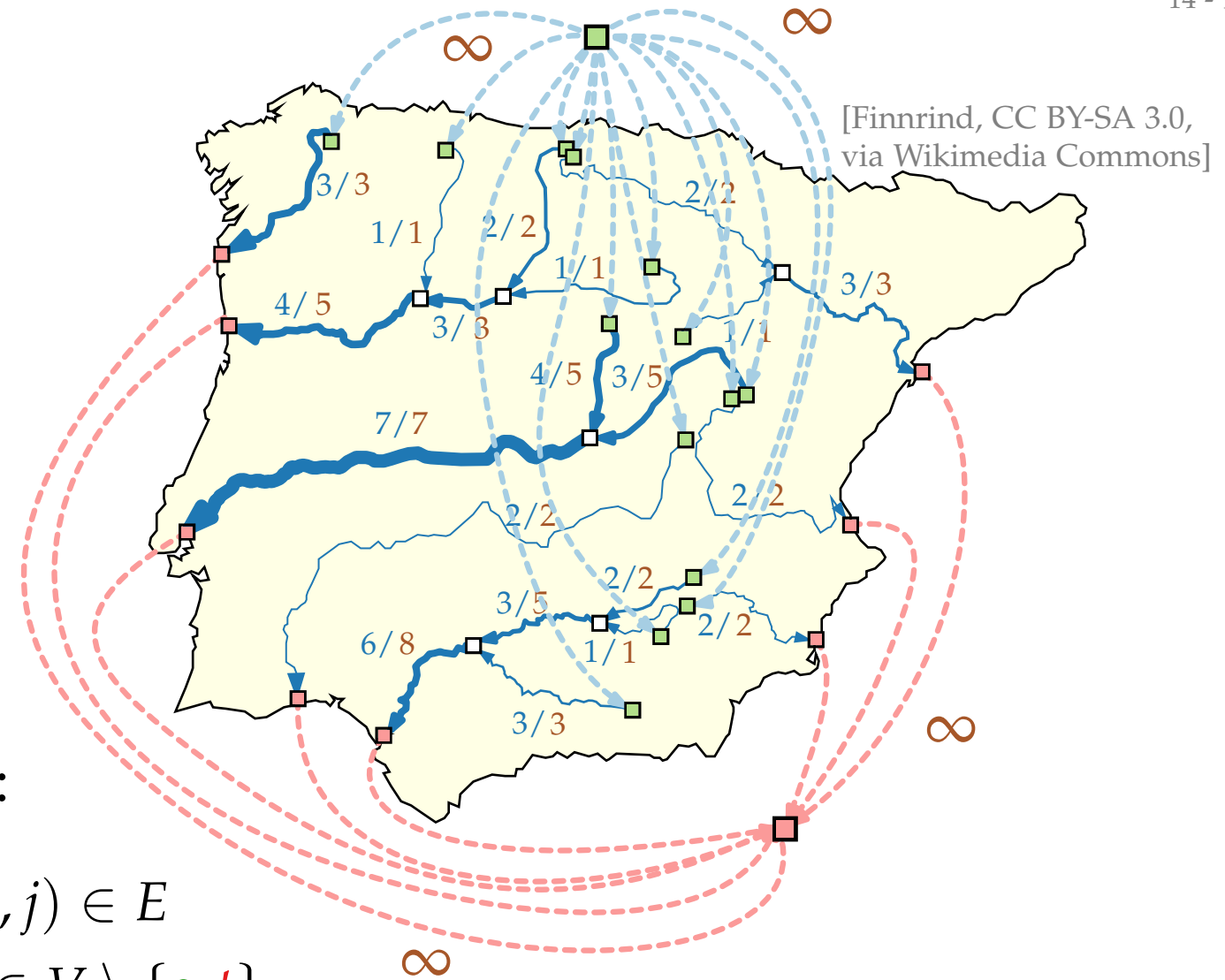
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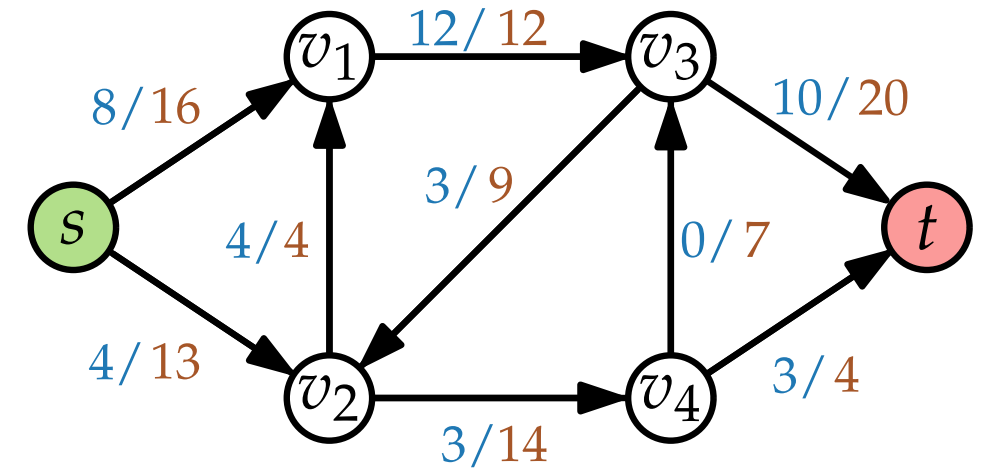
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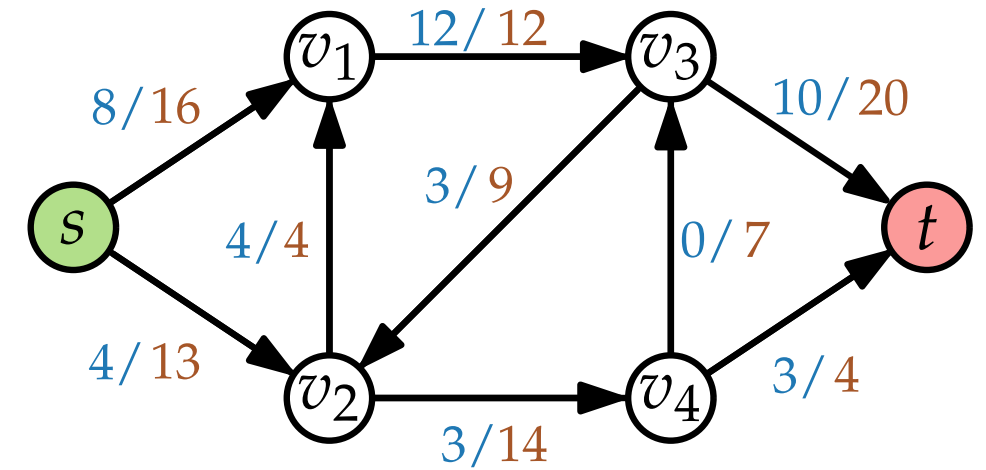
Flow network  $(G = (V, E); s, t; u)$



# Residual Network

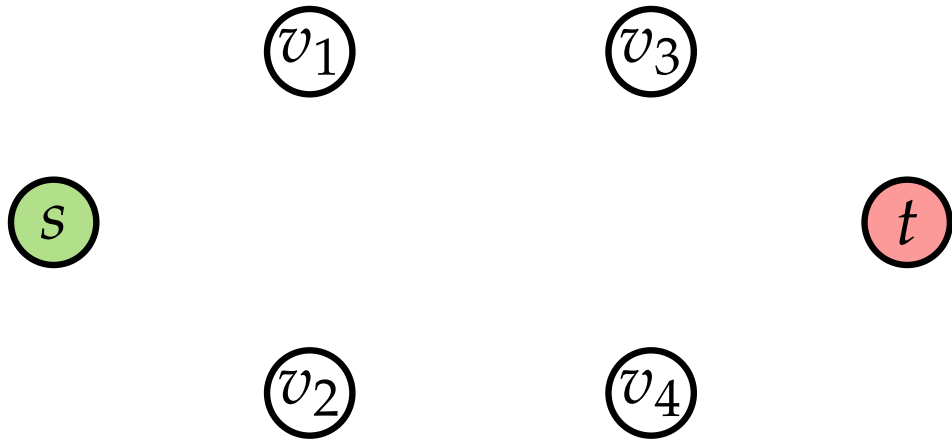
Residual network  $G_X = (V, E')$ :

Flow network  $(G = (V, E); s, t; u)$

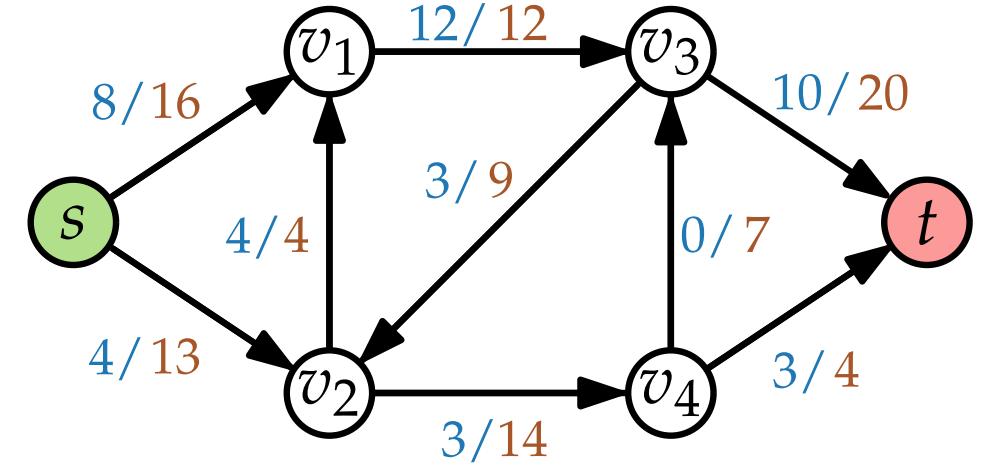


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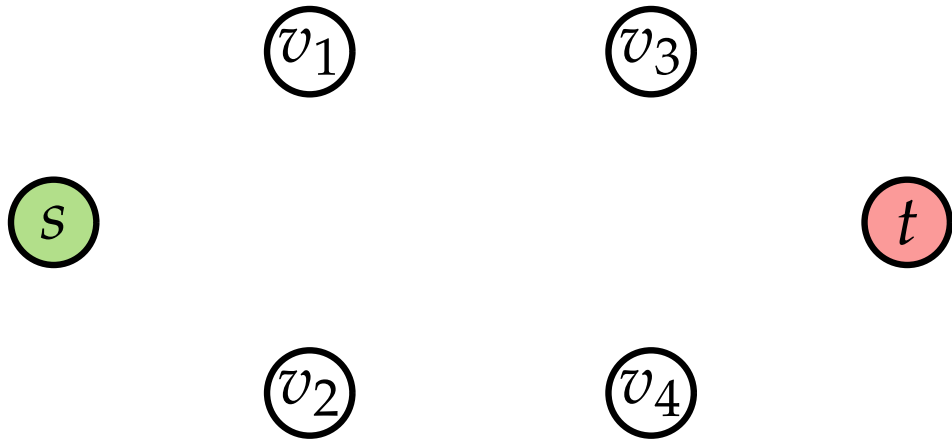




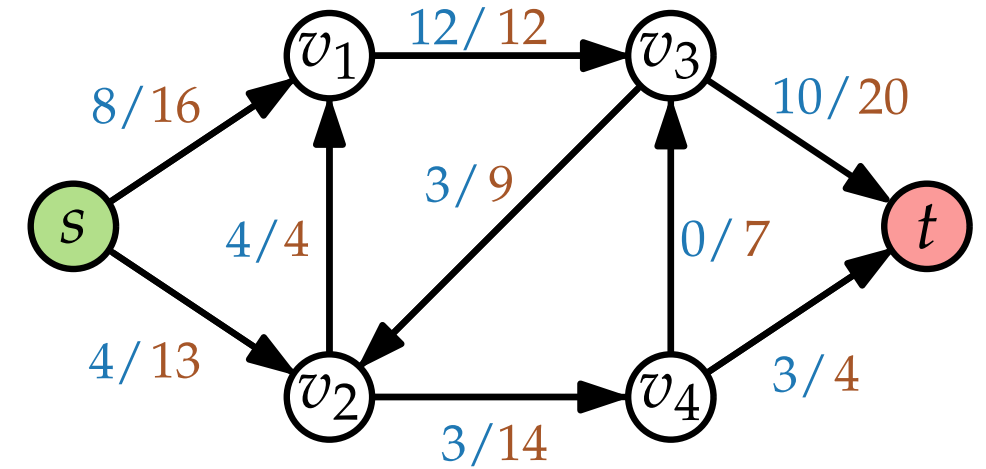
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Residual network  $G_X = (V, E')$ :

- $X(v, v') < u(v, v') \Rightarrow (v, v') \in E'$



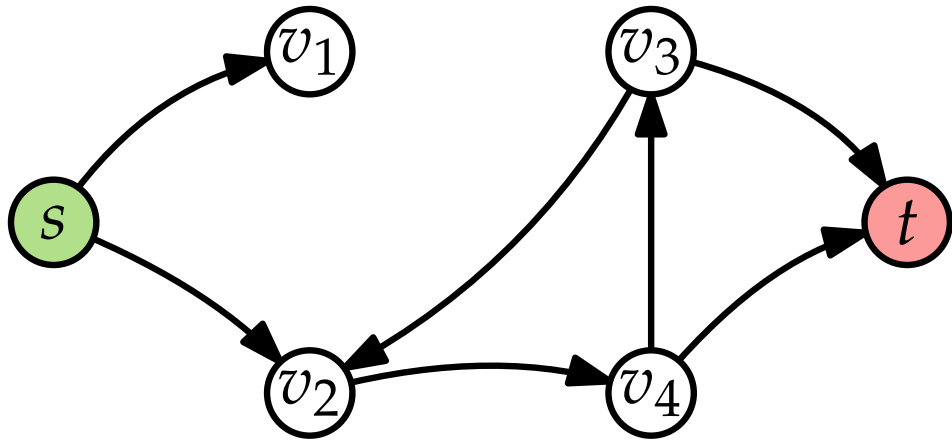
Flow network  $(G = (V, E); s, t; u)$



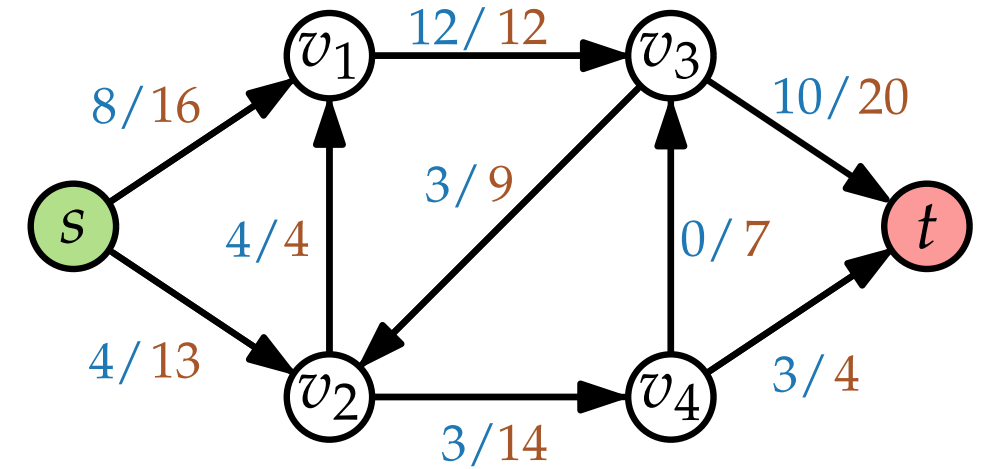
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Flow network  $(G = (V, E); s, t; u)$

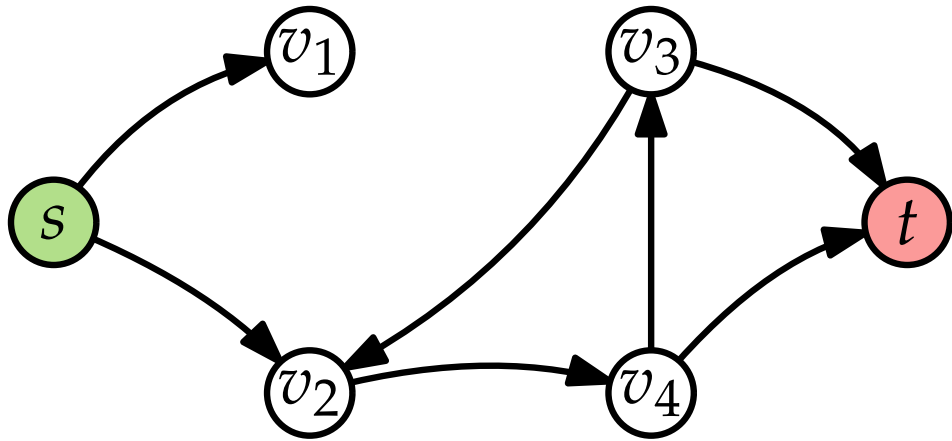


# Residual Network

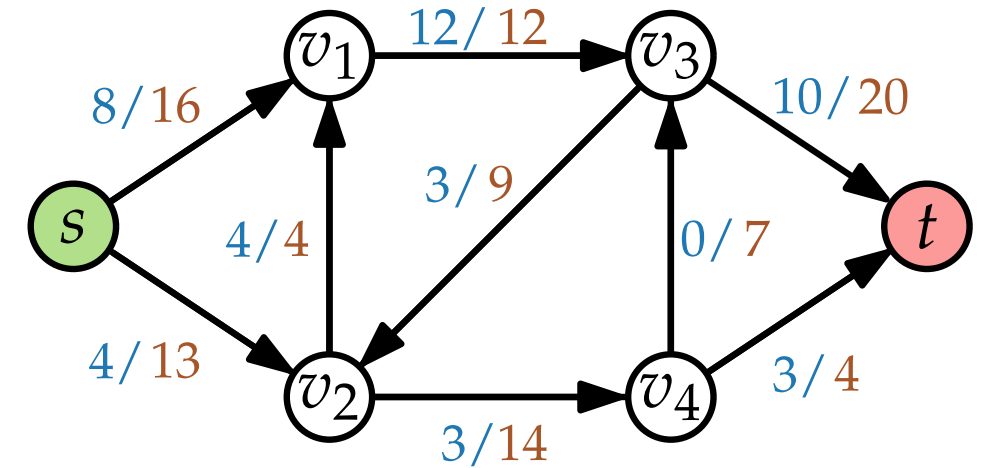
Residual network  $G_X = (V, E')$ :

■  $X(v, v') < u(v, v') \Rightarrow (v, v') \in E'$

■  $X(v, v') > 0 \Rightarrow (v', v) \in E'$



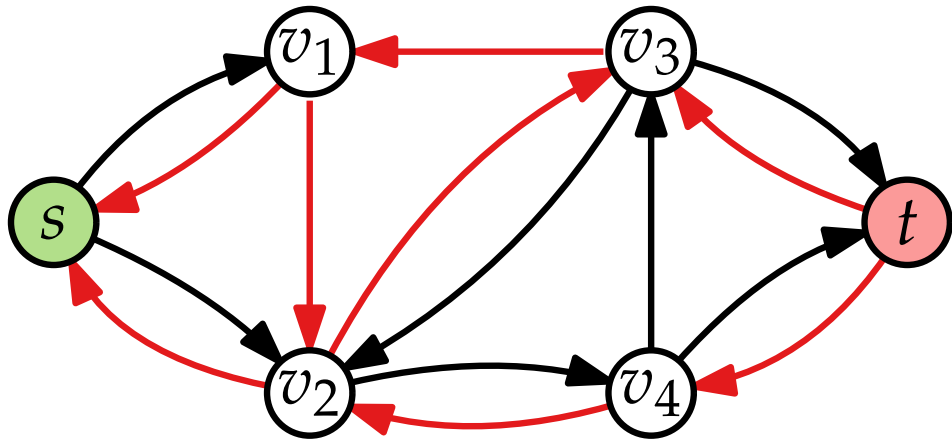
Flow network  $(G = (V, E); s, t; u)$



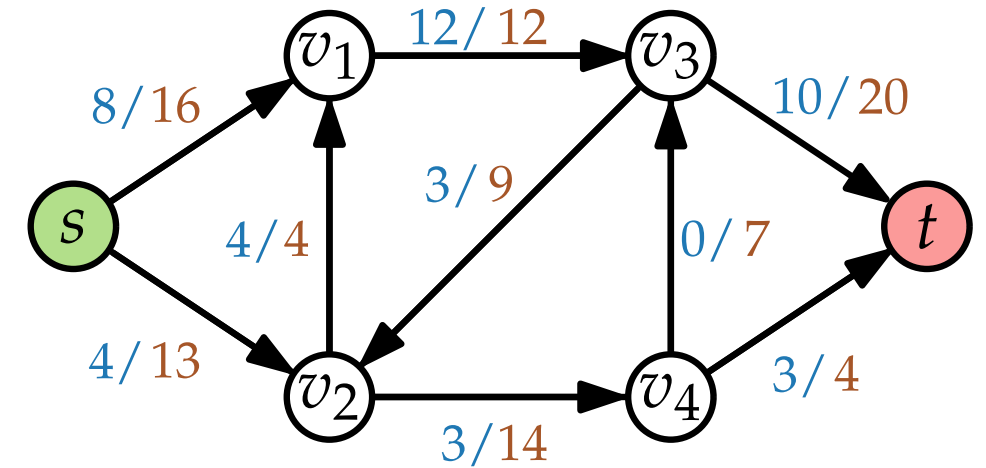
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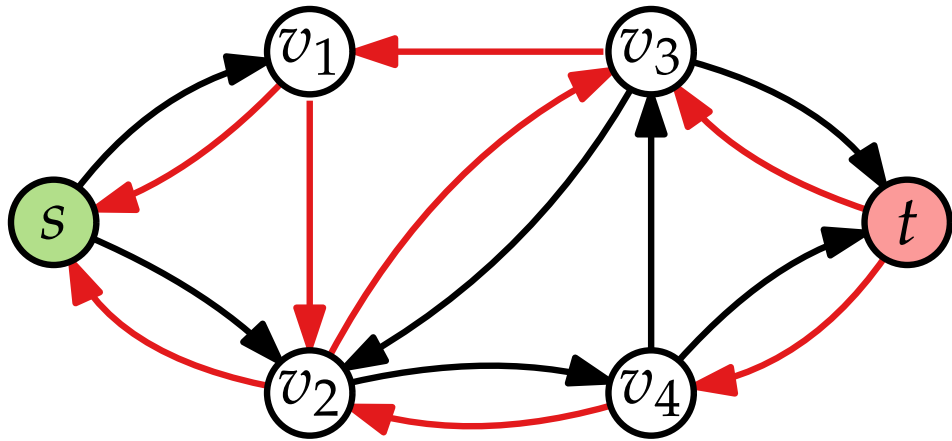
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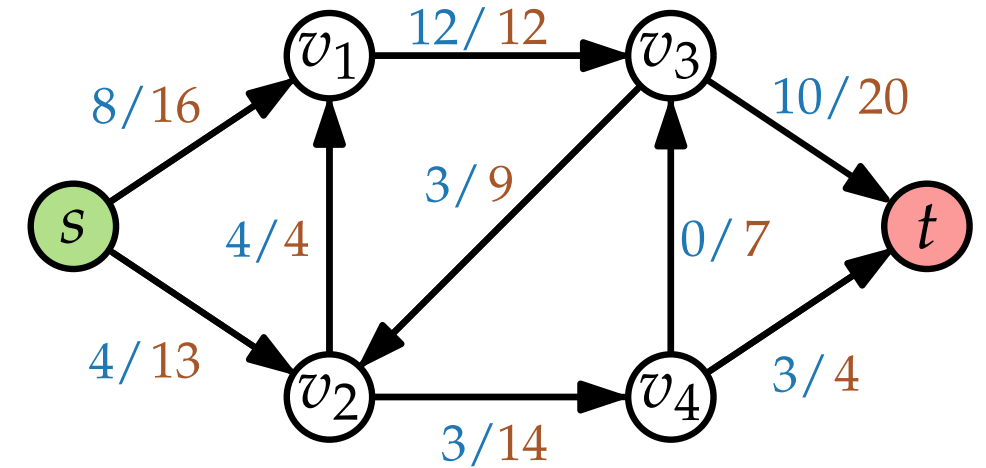
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Residual network  $G_X = (V, E')$ :

- $X(v, v') < u(v, v') \Rightarrow (v, v') \in E'$   
 $c(v, v') = u(v, v') - X(v, v')$
- $X(v, v') > 0 \Rightarrow (v', v) \in E'$



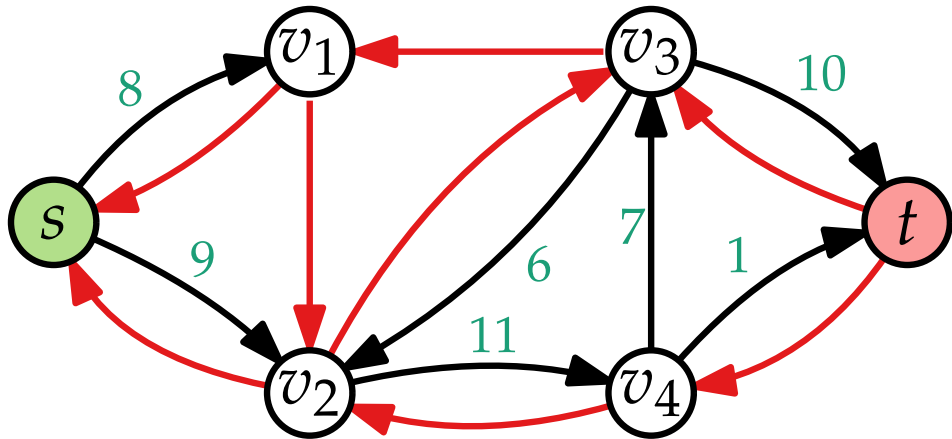
Flow network  $(G = (V, E); s, t; u)$



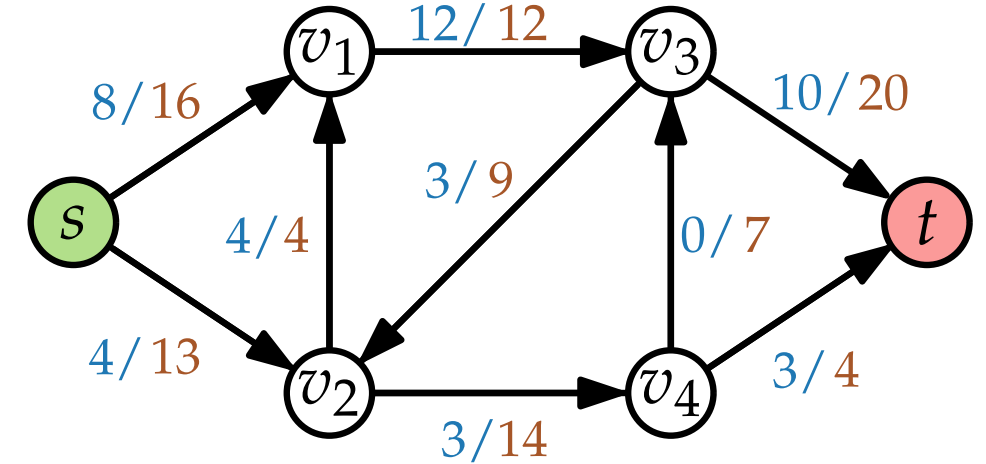
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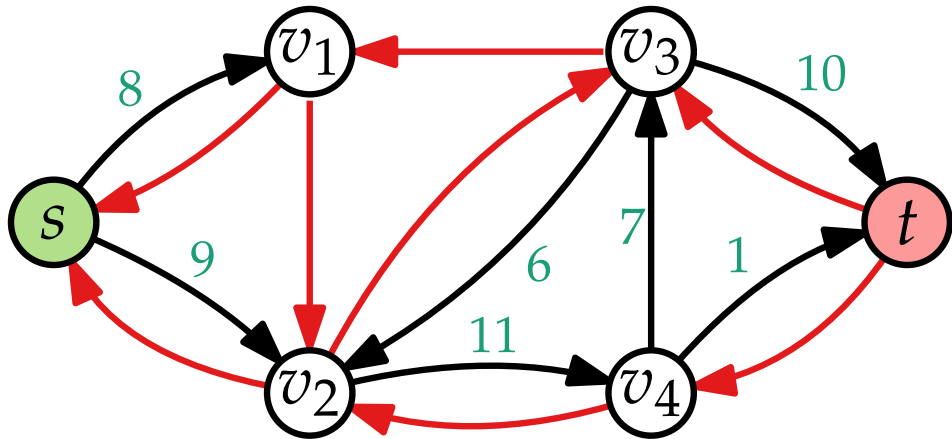
Flow network  $(G = (V, E); s, t; u)$



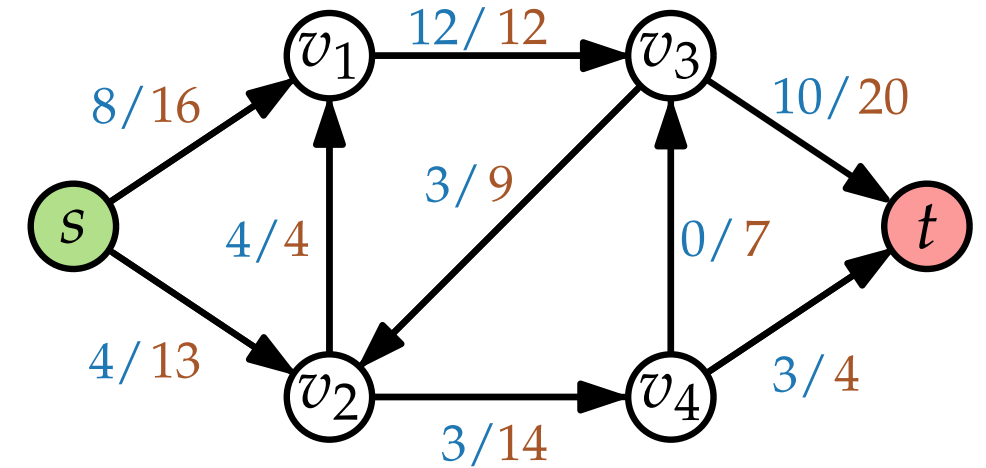
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- $X(v, v') > 0 \Rightarrow (v', v) \in E'$   
 $c(v, v') = X(v, v')$



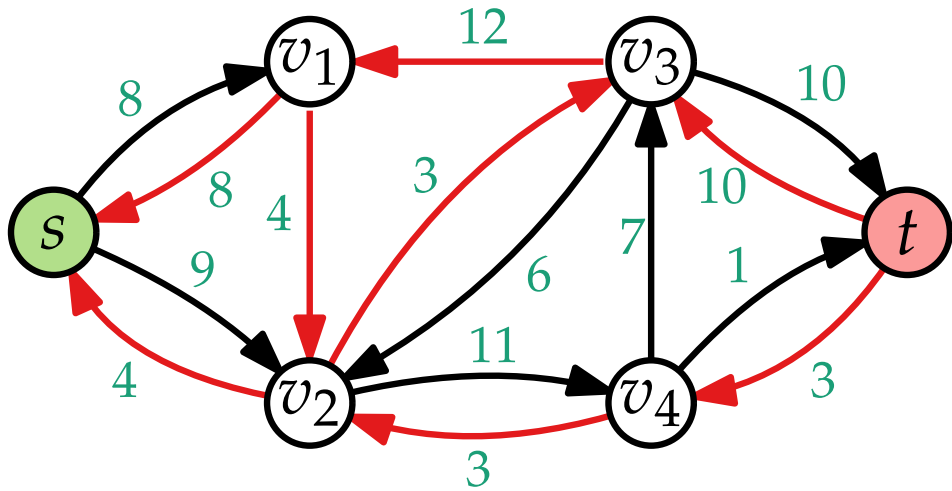
Flow network  $(G = (V, E); s, t; u)$



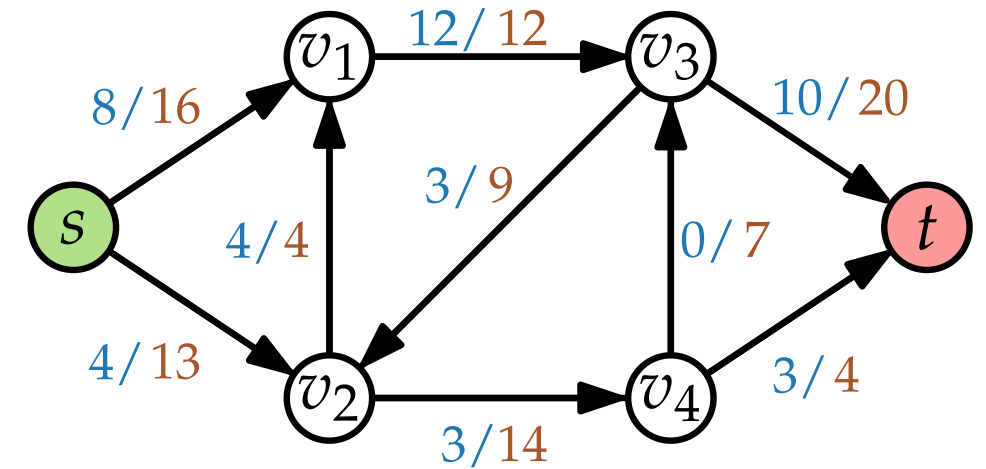
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Flow network  $(G = (V, E); s, t; u)$

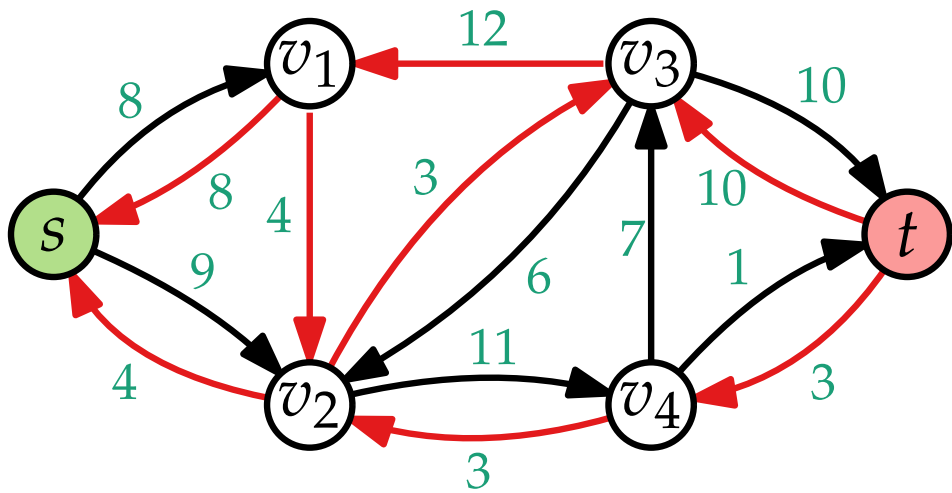




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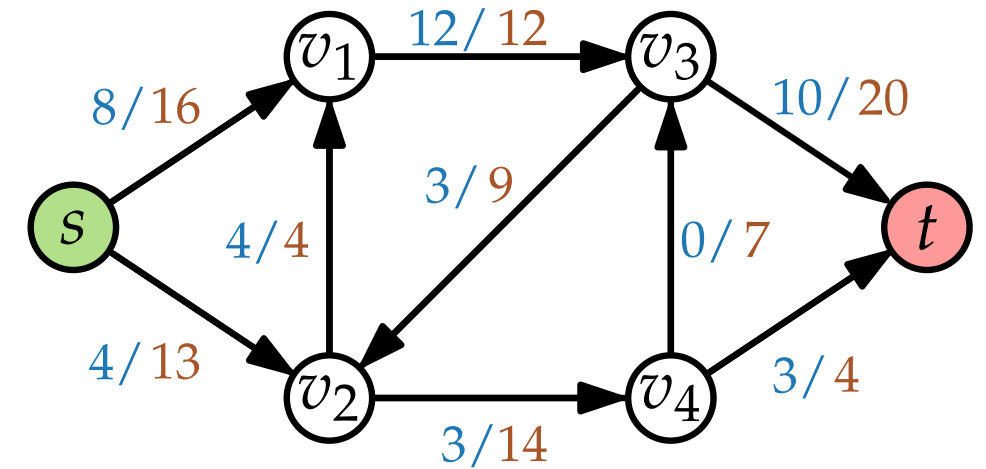
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- $X(v, v') > 0 \Rightarrow (v', v) \in E'$   
 $c(v, v') = u(v, v')$



Flow-increasing path  $W$

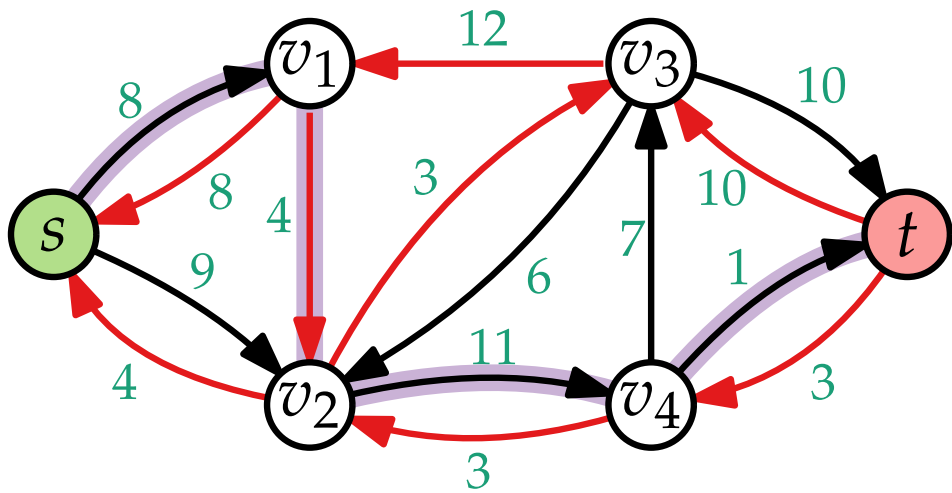
Flow network  $(G = (V, E); s, t; u)$



# Residual Network

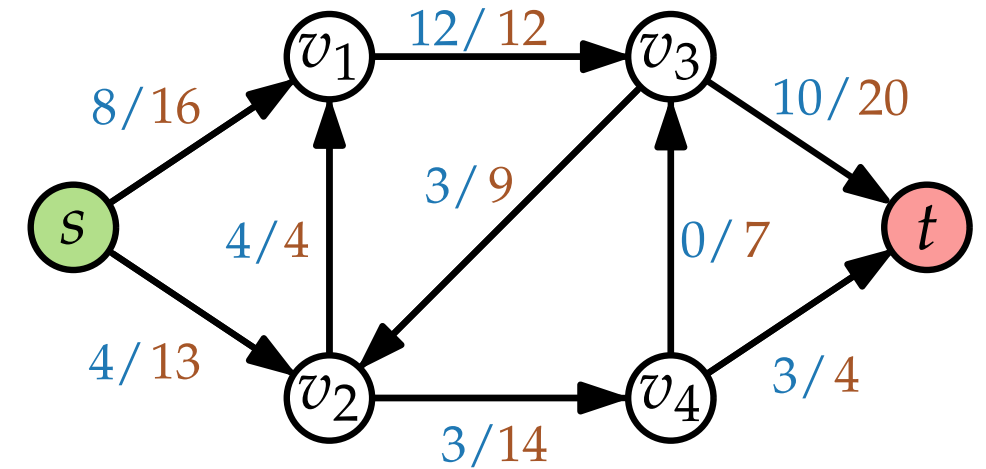
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Flow-increasing path  $W$

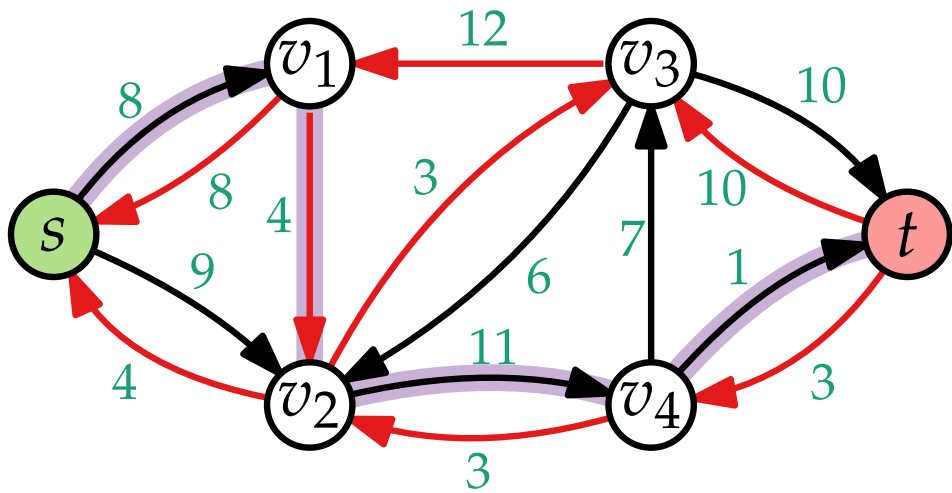
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# Residual Network

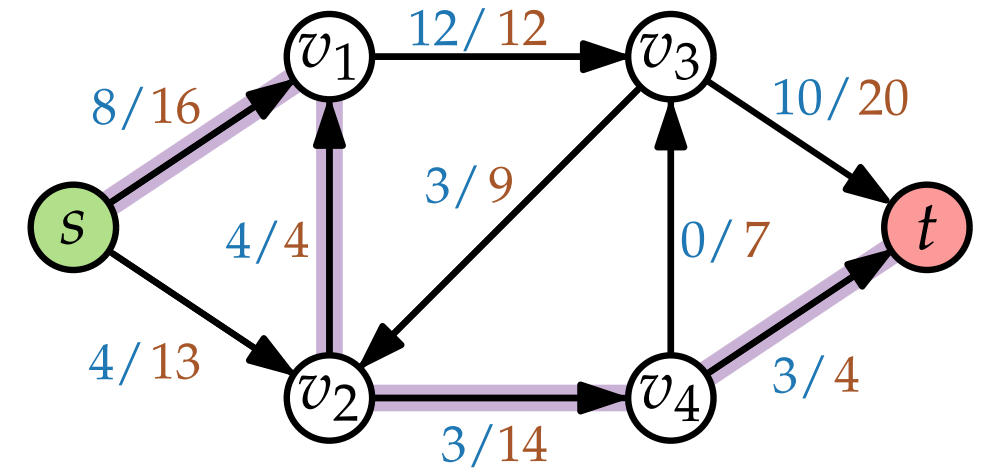
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Flow-increasing path  $W$

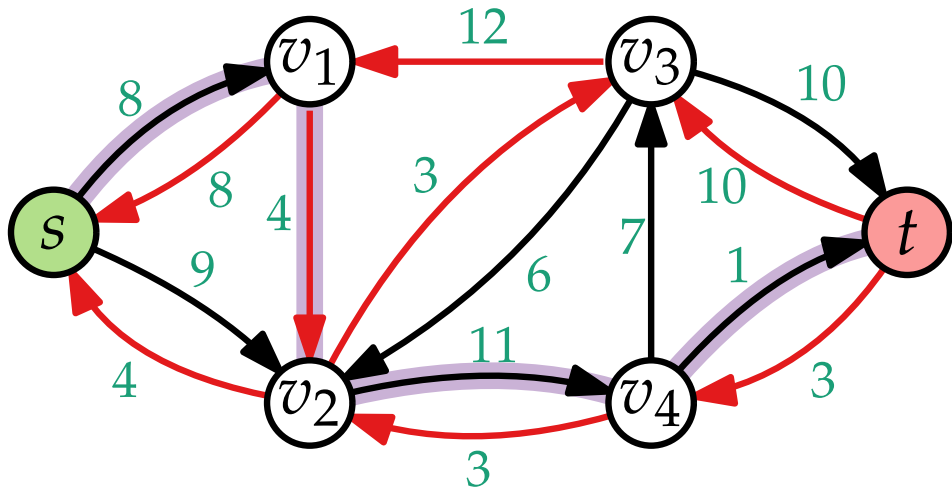
Flow network  $(G = (V, E); s, t; u)$



# Residual Network

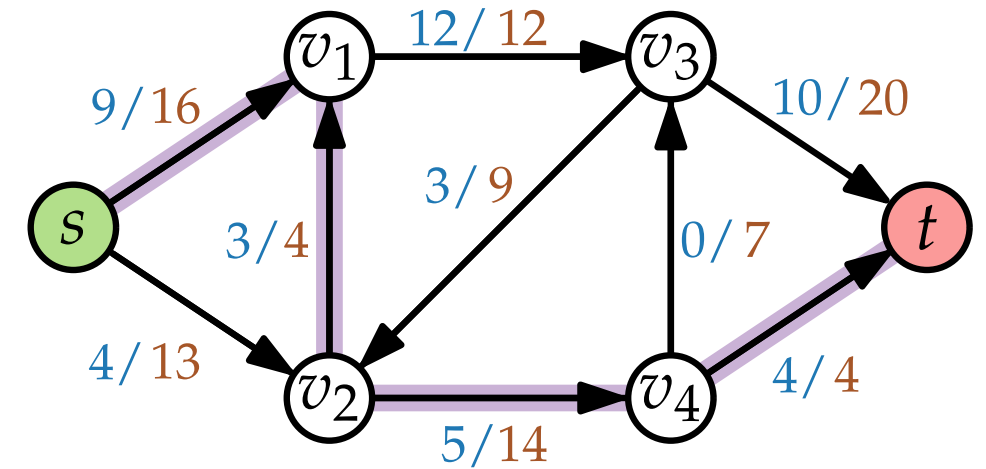
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Flow-increasing path  $W$

Flow network  $(G = (V, E); s, t; u)$



# FordFulkerson

FordFulkerson( $G = (V, E); s, t; u$ )

# FordFulkerson

```
FordFulkerson( $G = (V, E); s, t; u$ )
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  foreach  $(v, v') \in E$  do
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     $X(v, v') = 0$ 
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} Initialization with Zero-flow

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} Max Flow



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  while  $G_X$  contains  $s$ - $t$ -path  $W$  do
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```
     $\Delta_W = \min_{(v, v') \in W} c(v, v')$ 
```

```
  return  $X$ 
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} Initialization with Zero-flow

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} Initialization with Zero-flow

} Capacity of  $W$

} Max Flow

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```

```
    foreach  $(v, v') \in W$  do
```

```
      if  $(v, v') \in E$  then
```

```
         $X(v, v') = X(v, v') + \Delta_W$ 
```

```
  return  $X$ 
```

} Initialization with Zero-flow

} Capacity of  $W$

} Max Flow

# FordFulkerson

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FordFulkerson( $G = (V, E); s, t; u$ )
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```

```
      else
```

```
         $X(v, v') = X(v, v') - \Delta_W$ 
```

```
  return  $X$ 
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} Initialization with Zero-flow

} Capacity of  $W$

} Max Flow

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      else
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```

```
  return  $X$ 
```

} Initialization with Zero-flow

} Capacity of  $W$

} Increasing flow along  $W$

} Max Flow

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      else
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         $X(v, v') = X(v, v') - \Delta_W$ 
```

```
  return  $X$ 
```

} Initialization with Zero-flow

} Capacity of  $W$

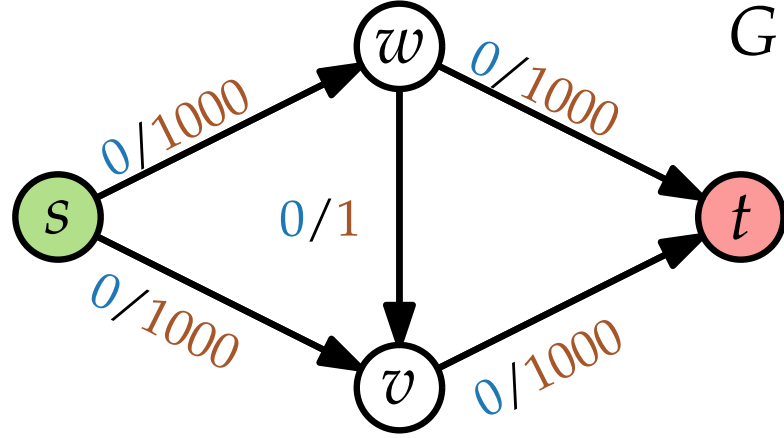
} Increasing flow along  $W$

} Max Flow

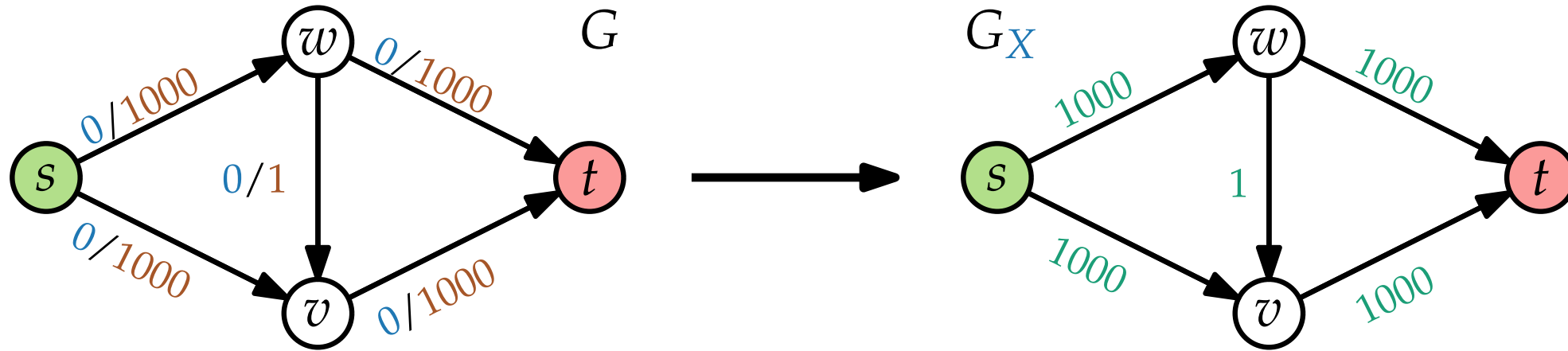
FordFulkerson finds a maximum  $s$ - $t$ -flow in  $O(|X^*| \cdot n)$  time.



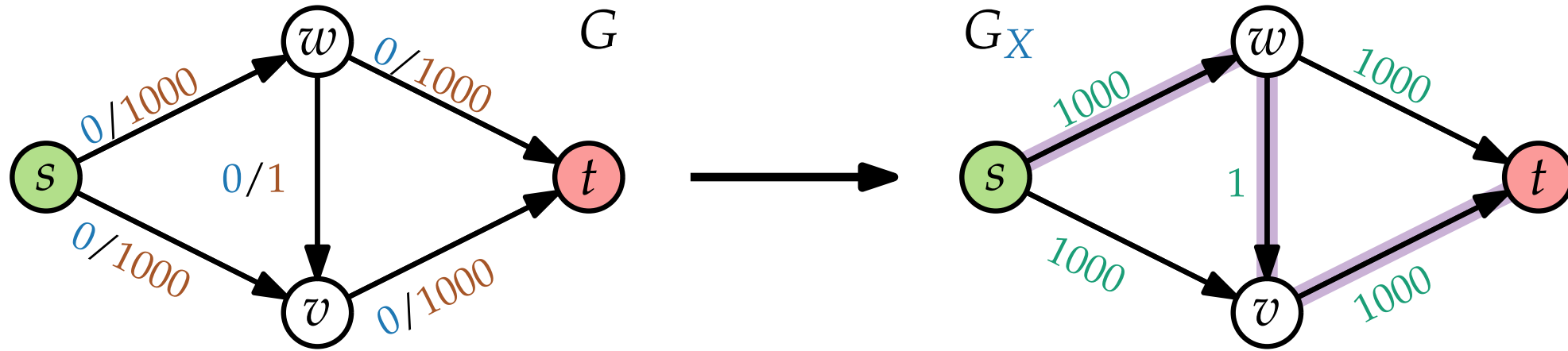
# FordFulkerson – Example



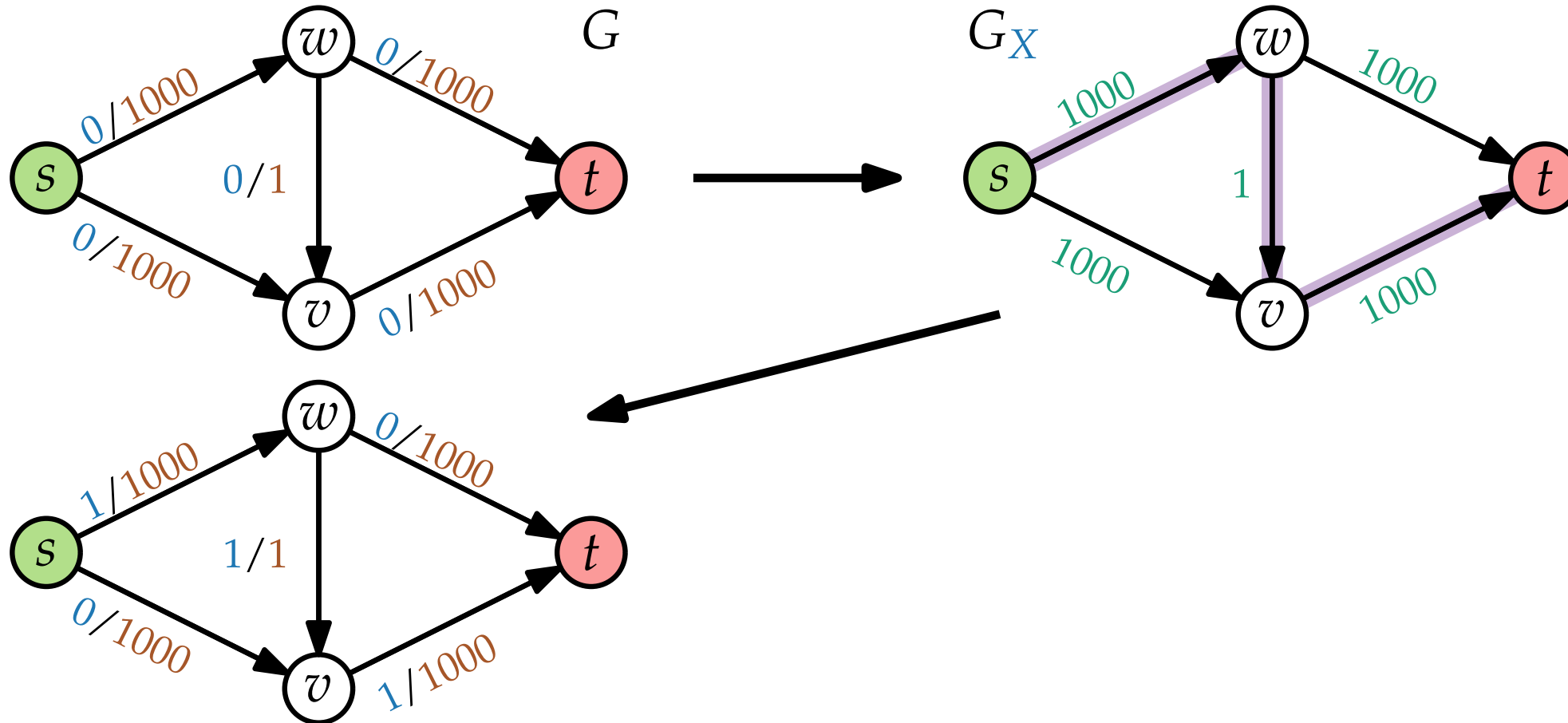
# FordFulkerson – Example



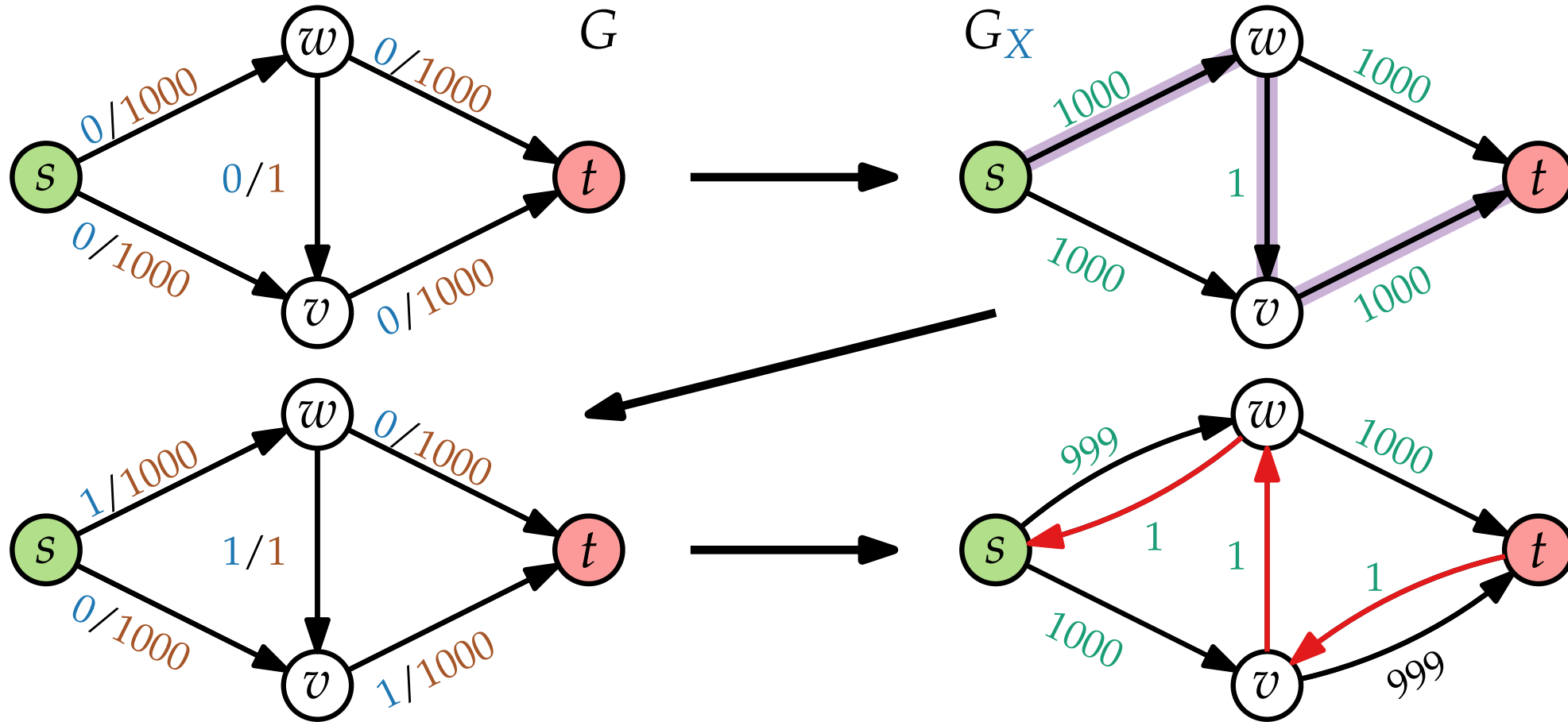
# FordFulkerson – Example



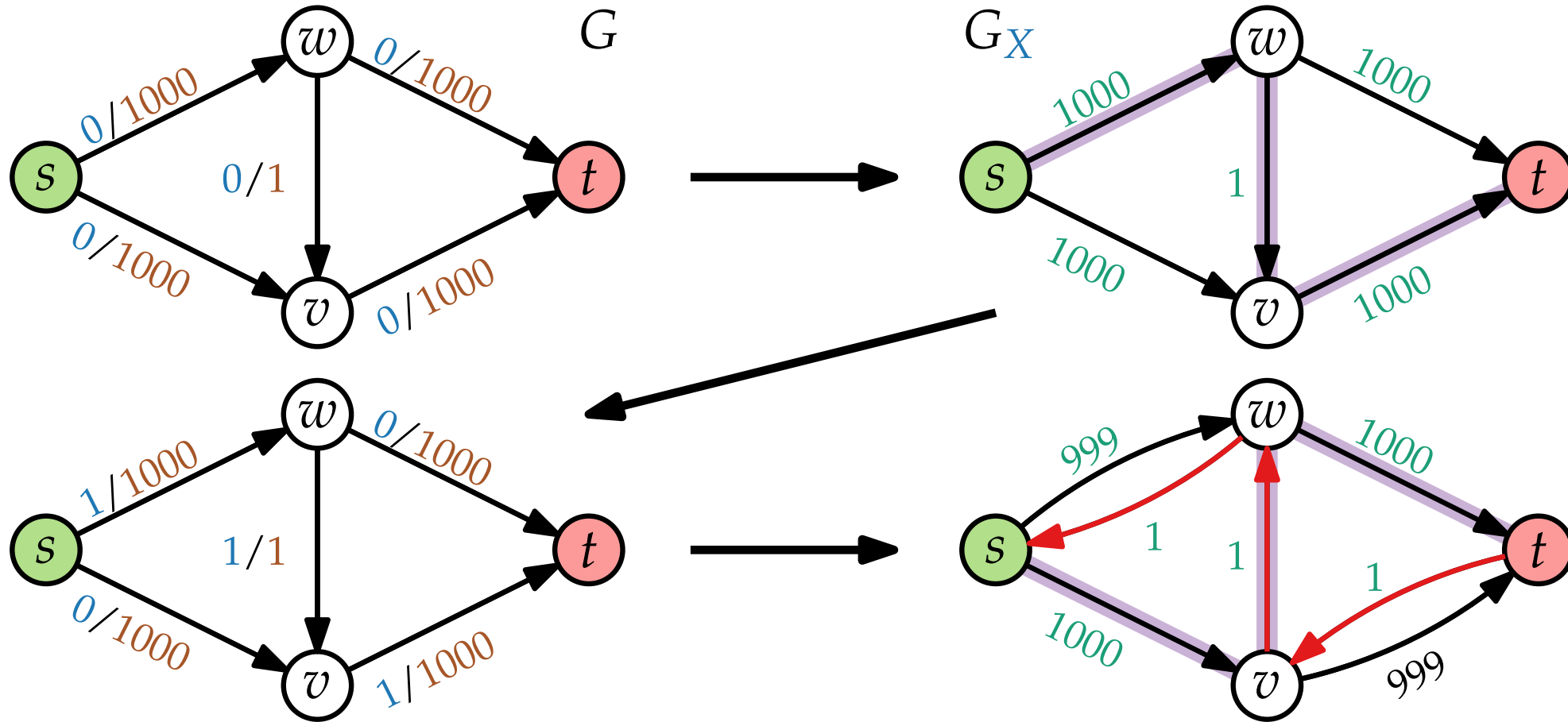
# FordFulkerson – Example



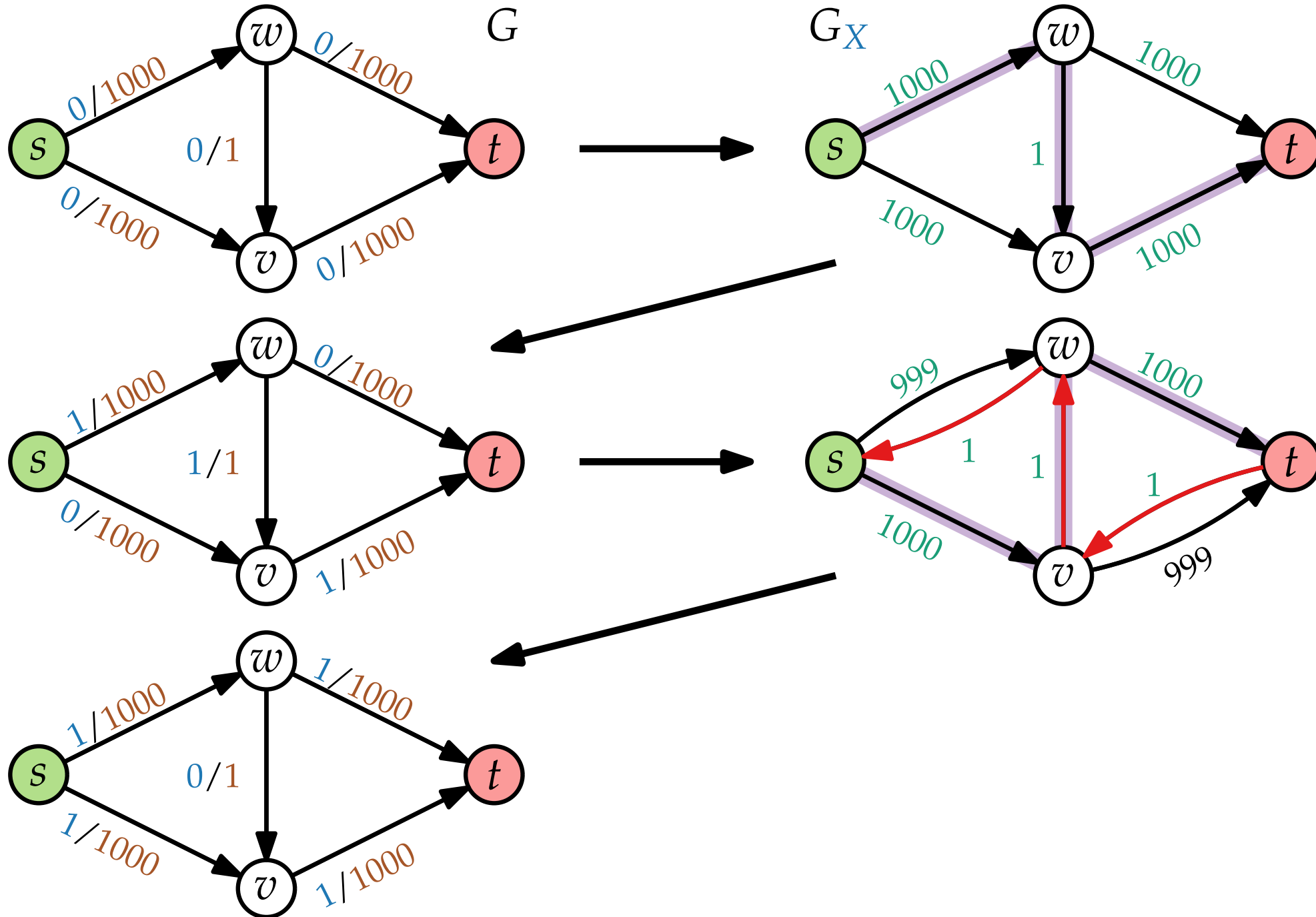
# FordFulkerson – Example



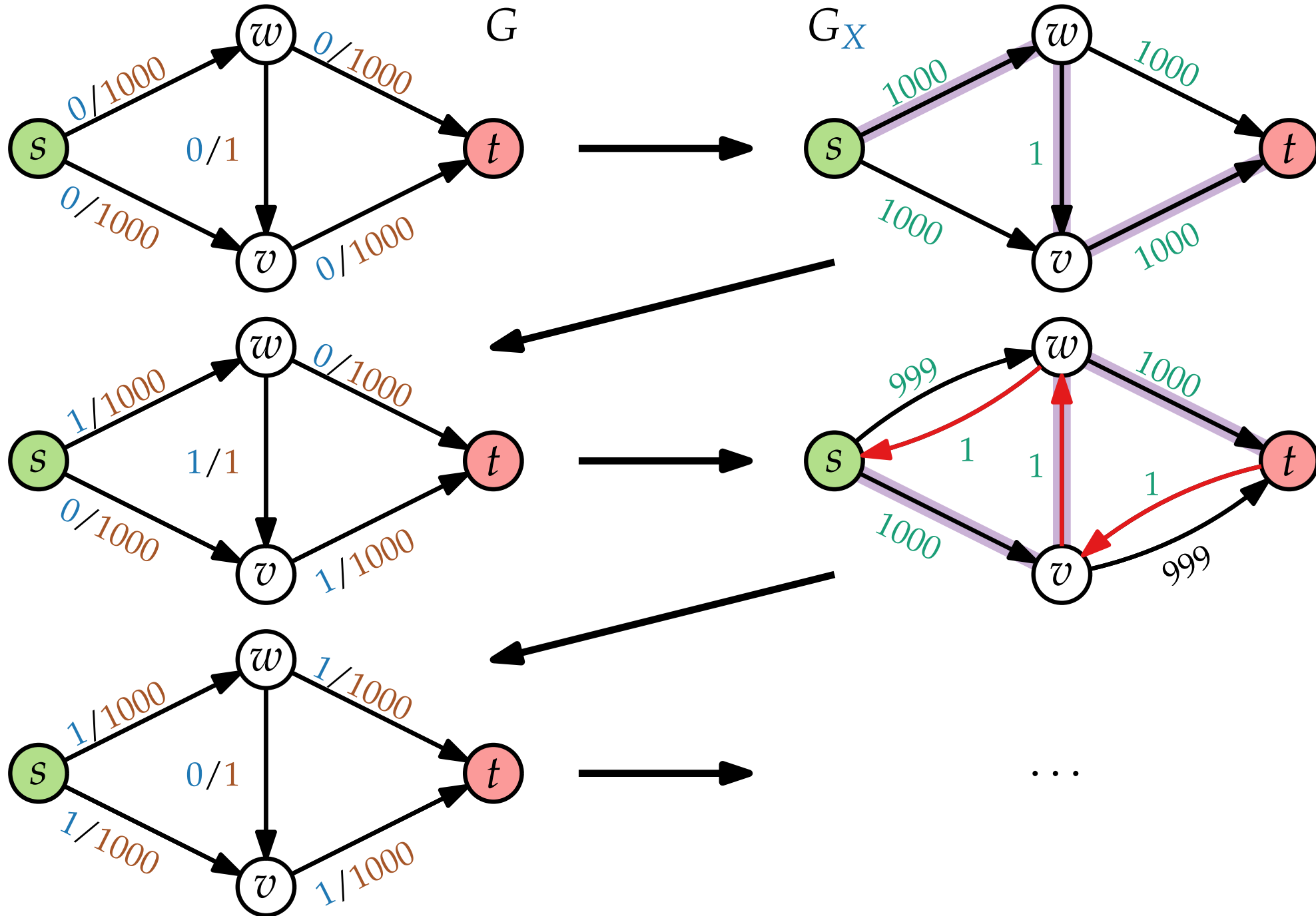
# FordFulkerson – Example



# FordFulkerson – Example



# FordFulkerson – Example





# EdmondsKarp

FordFulkerson( $G = (V, E); s, t; u$ )

**foreach**  $(v, v') \in E$  **do**

└  $X(v, v') = 0$

**while**  $G_X$  contains  $s$ - $t$ -path  $W$  **do**

└  $W =$   $s$ - $t$ -path in  $G_X$

└  $\Delta_W = \min_{(v, v') \in c(v, v')}$

**foreach**  $(v, v') \in W$  **do**

└ **if**  $(v, v') \in E$  **then**

└└  $X(v, v') = X(v, v') + \Delta_W$

**else**

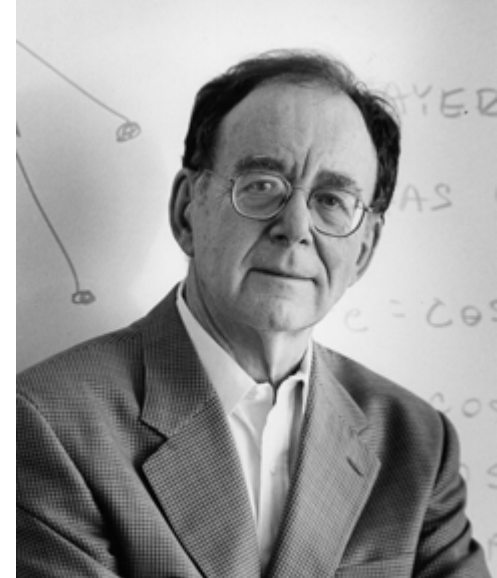
└└  $X(v, v') = X(v, v') - \Delta_W$

**return**  $X$

Jack R. Edmonds  
\*1934



Richard M. Karp  
\*1935 Boston, MA



# EdmondsKarp

EdmondsKarp

~~FordFulkerson~~( $G = (V, E); s, t; u$ )

**foreach**  $(v, v') \in E$  **do**

└  $X(v, v') = 0$

**while**  $G_X$  contains  $s$ - $t$ -path  $W$  **do**

└  $W =$   $s$ - $t$ -path in  $G_X$

└  $\Delta_W = \min_{(v, v') \in c(v, v')}$

└ **foreach**  $(v, v') \in W$  **do**

└└ **if**  $(v, v') \in E$  **then**

└└└  $X(v, v') = X(v, v') + \Delta_W$

└└ **else**

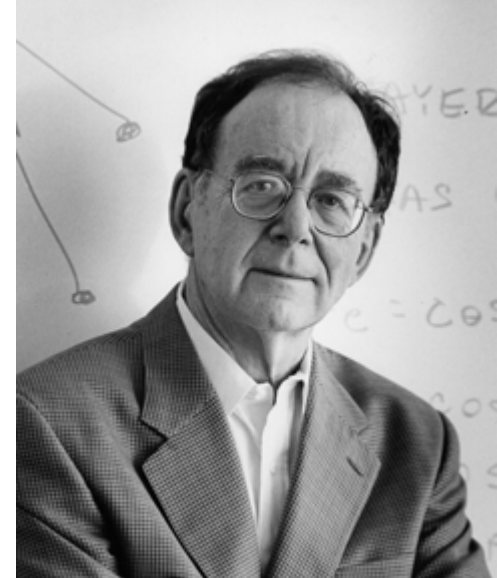
└└└  $X(v, v') = X(v, v') - \Delta_W$

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EdmondsKarp

~~FordFulkerson~~( $G = (V, E); s, t; u$ )

**foreach**  $(v, v') \in E$  **do**

└  $X(v, v') = 0$

**while**  $G_X$  contains  $s$ - $t$ -path  $W$  **do**

└  $W =$  shortest  $s$ - $t$ -path in  $G_X$

└  $\Delta_W = \min_{(v, v') \in c(v, v')}$

└ **foreach**  $(v, v') \in W$  **do**

└└ **if**  $(v, v') \in E$  **then**

└└└  $X(v, v') = X(v, v') + \Delta_W$

└└ **else**

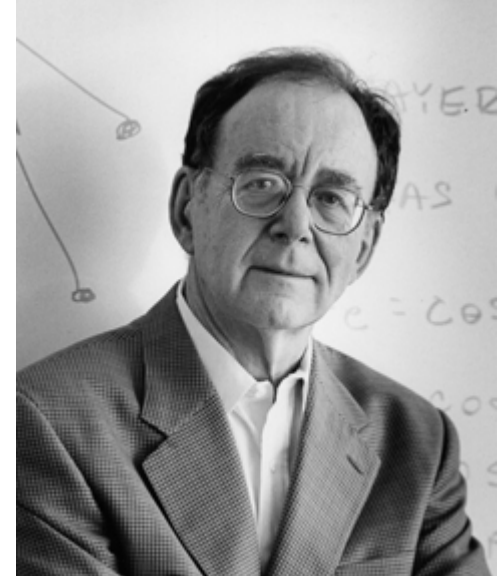
└└└  $X(v, v') = X(v, v') - \Delta_W$

**return**  $X$

Jack R. Edmonds  
\*1934



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# EdmondsKarp

EdmondsKarp

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while  $G_X$  contains  $s$ - $t$ -path  $W$  do

└  $W =$  shortest  $s$ - $t$ -path in  $G_X$

└  $\Delta_W = \min_{(v, v') \in c(v, v')}$

└ foreach  $(v, v') \in W$  do

└ if  $(v, v') \in E$  then

└ |  $X(v, v') = X(v, v') + \Delta_W$

└ else

└ |  $X(v, v') = X(v, v') - \Delta_W$

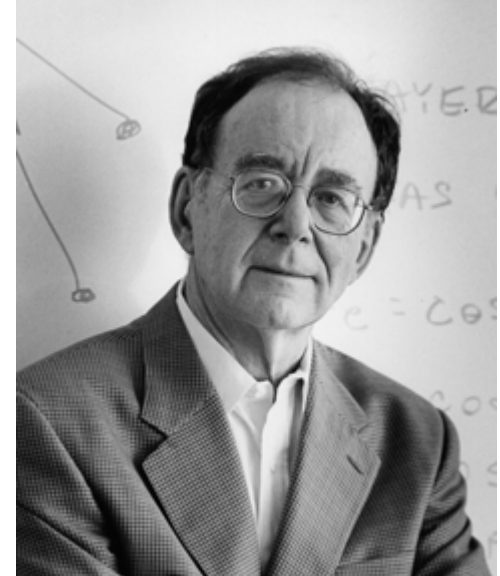
return  $X$

EdmondsKarp finds a maximum  $s$ - $t$ -flow in  $O(nm^2)$  time.

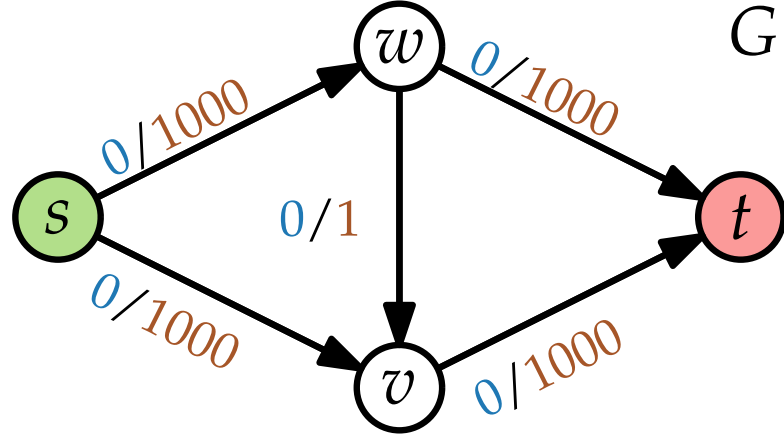
Jack R. Edmonds  
\*1934



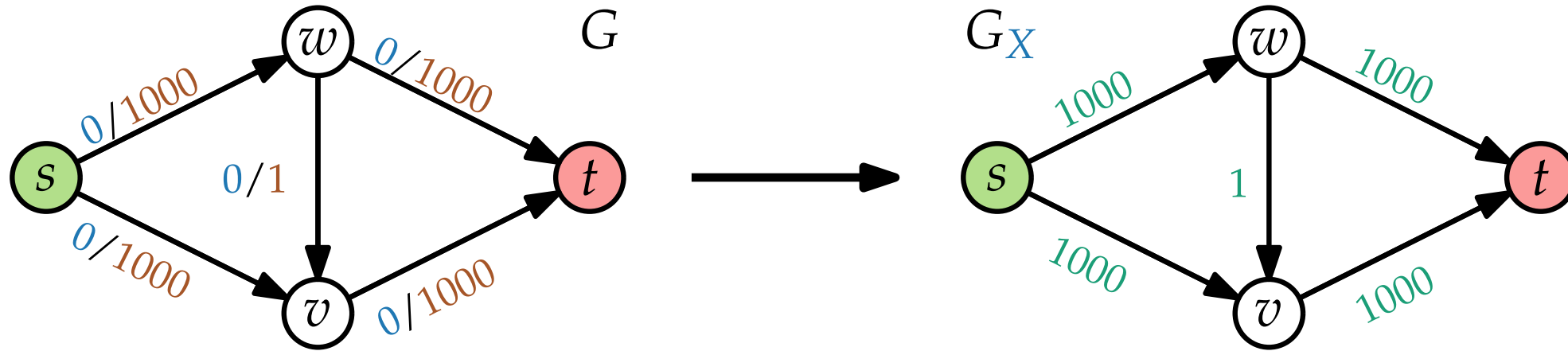
Richard M. Karp  
\*1935 Boston, MA



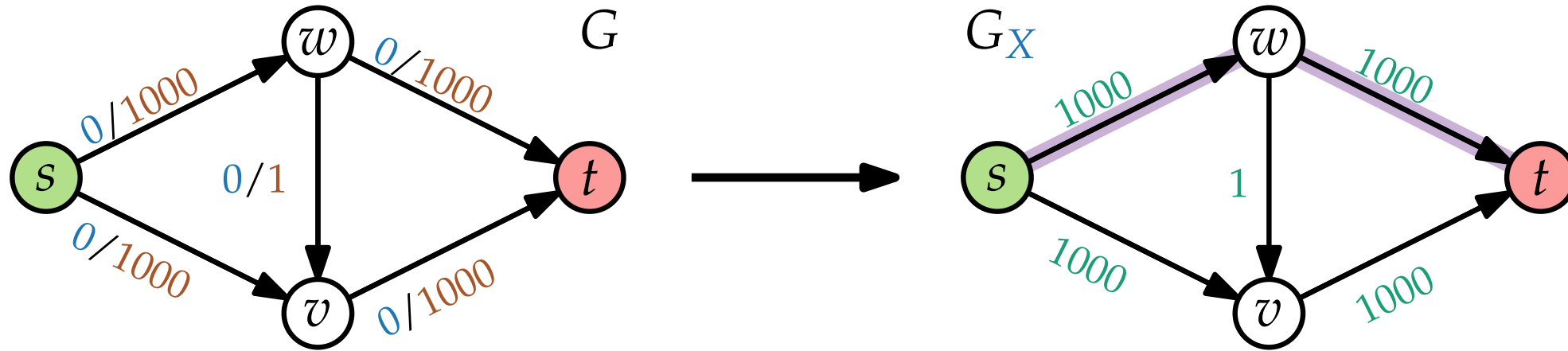
# EdmondsKarp – Example



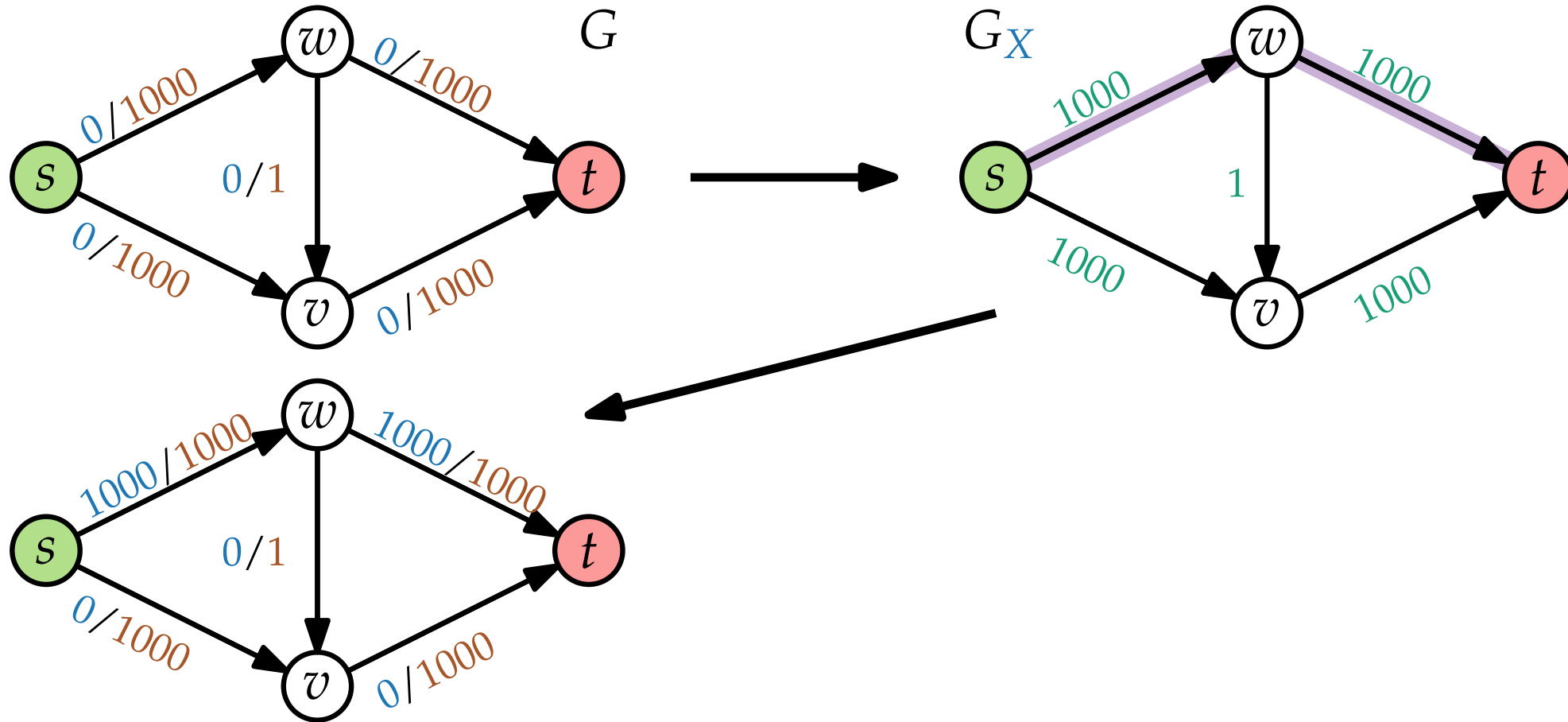
# EdmondsKarp – Example



# EdmondsKarp – Example

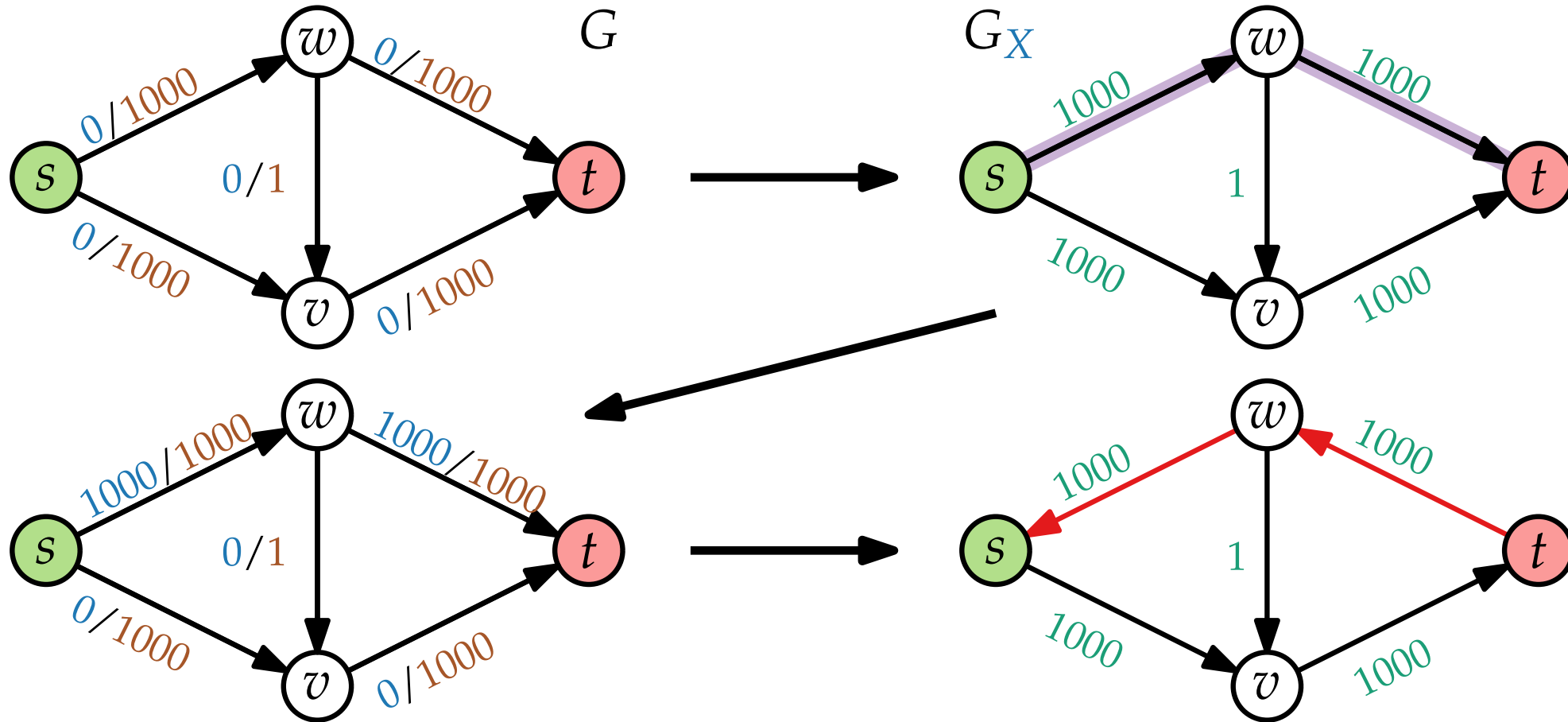


# EdmondsKarp – Example

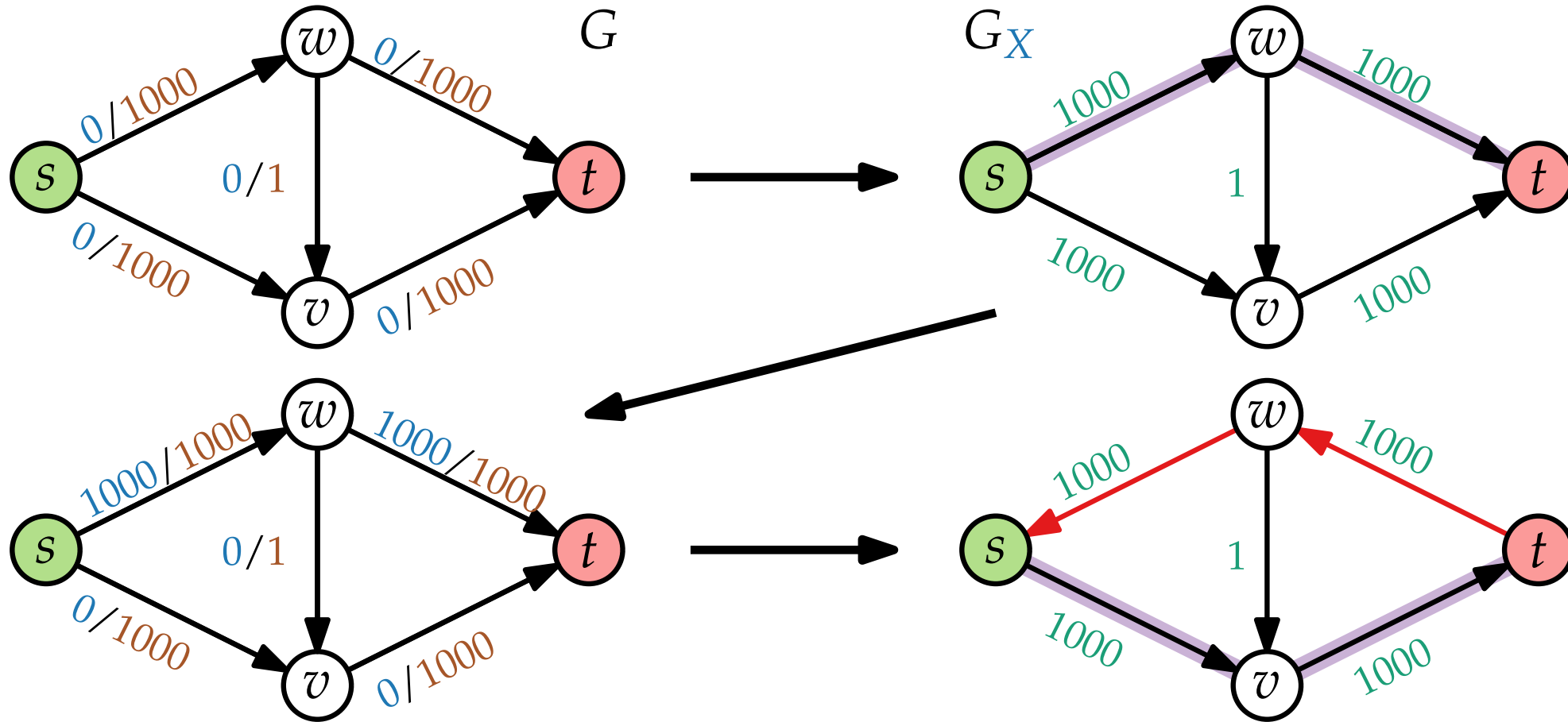




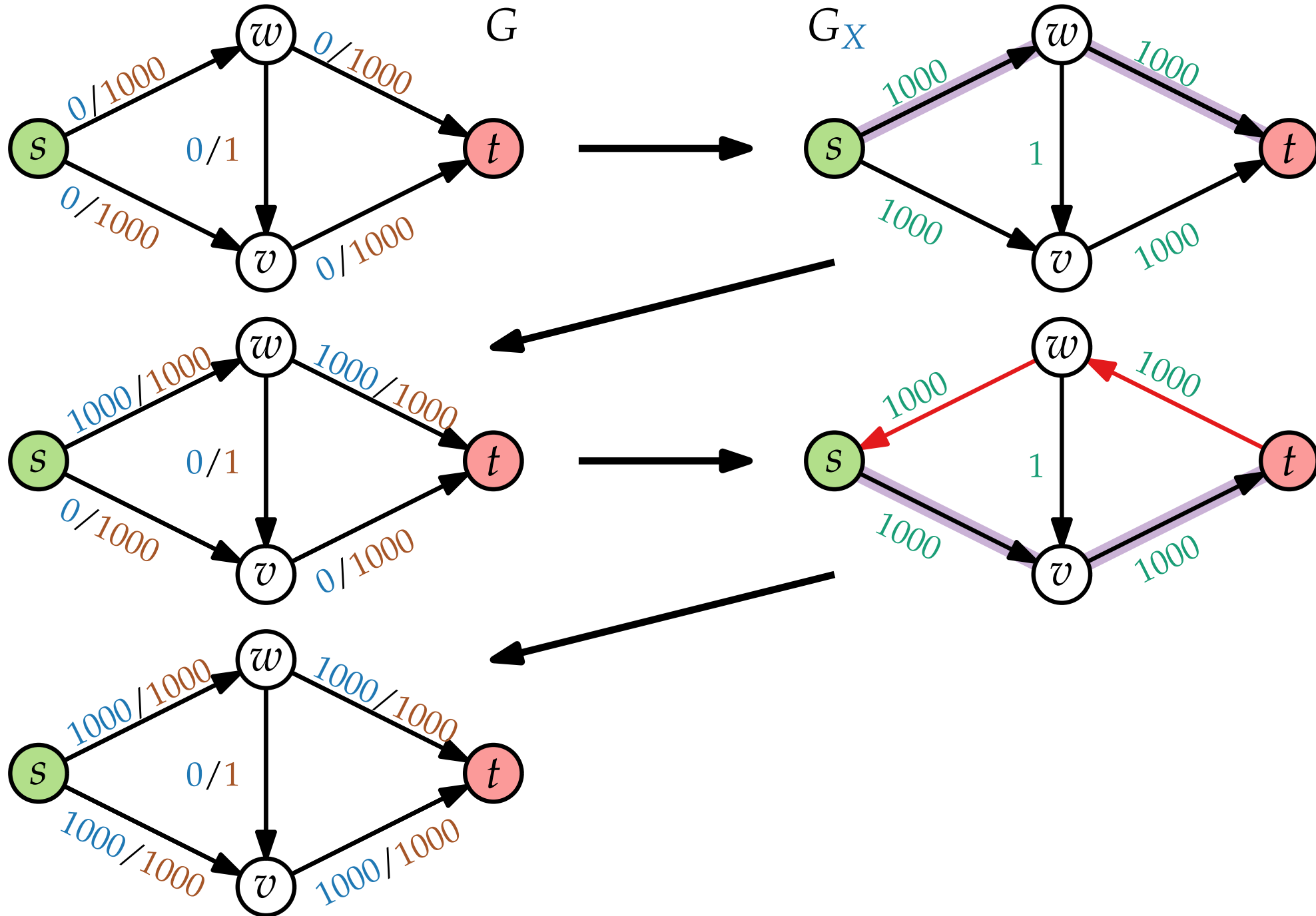
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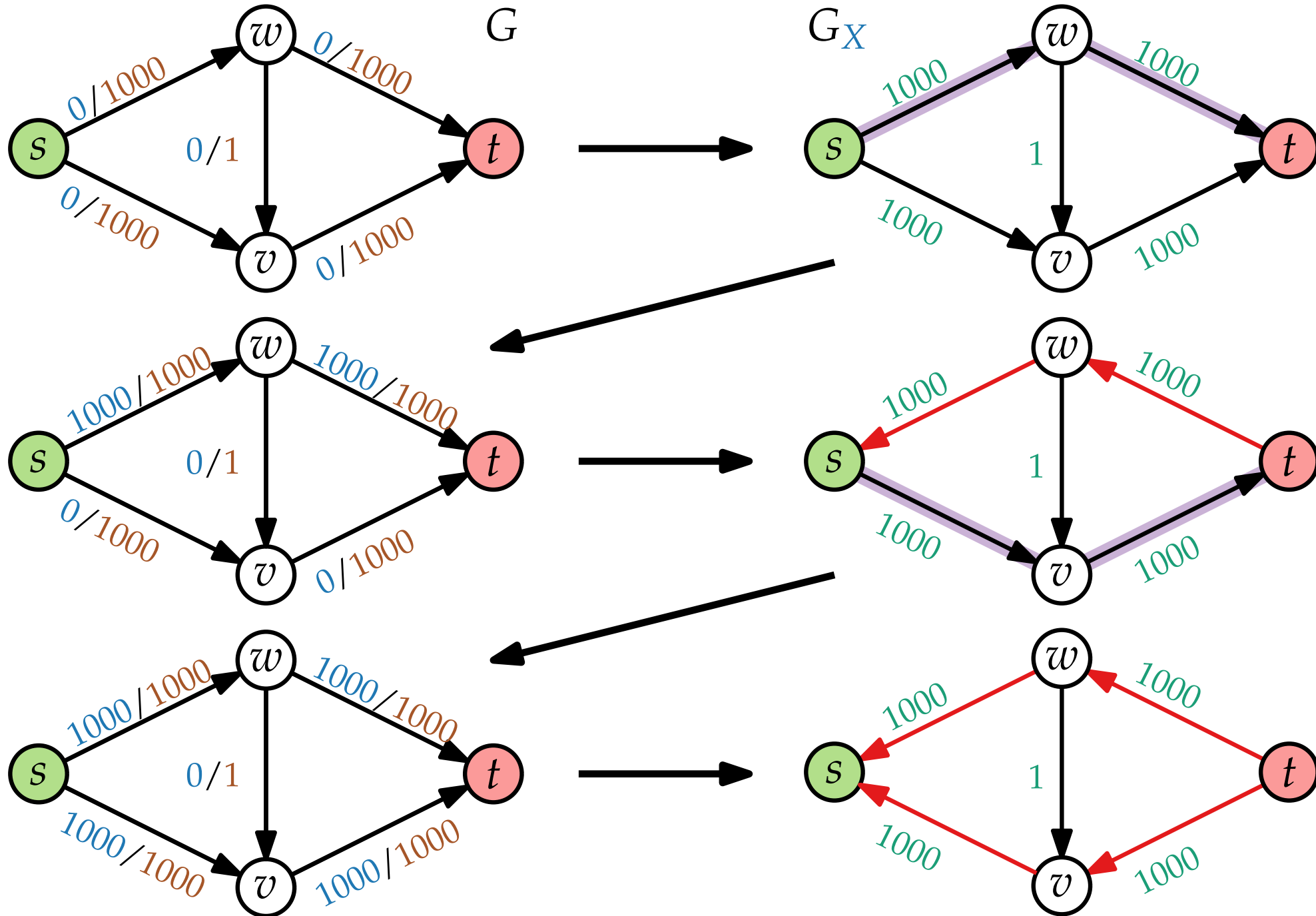
# EdmondsKarp – Example



# EdmondsKarp – Example



# EdmondsKarp – Example



# General Flow Network

**Flow network**  $(G = (V, E); S, T; u)$  with

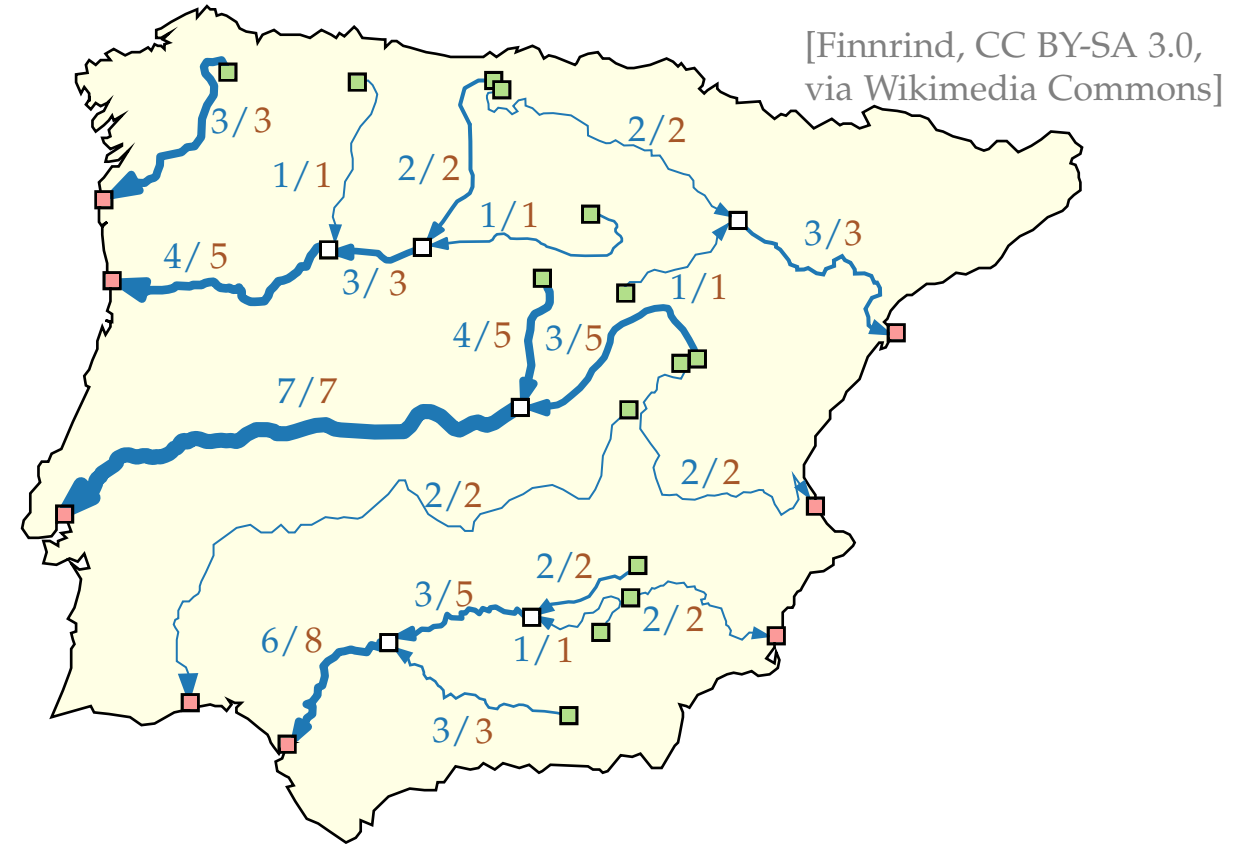
- directed graph  $G = (V, E)$
- *sources*  $S \subseteq V$ , *sinks*  $T \subseteq V$
- edge *capacity*  $u: E \rightarrow \mathbb{R}_0^+$

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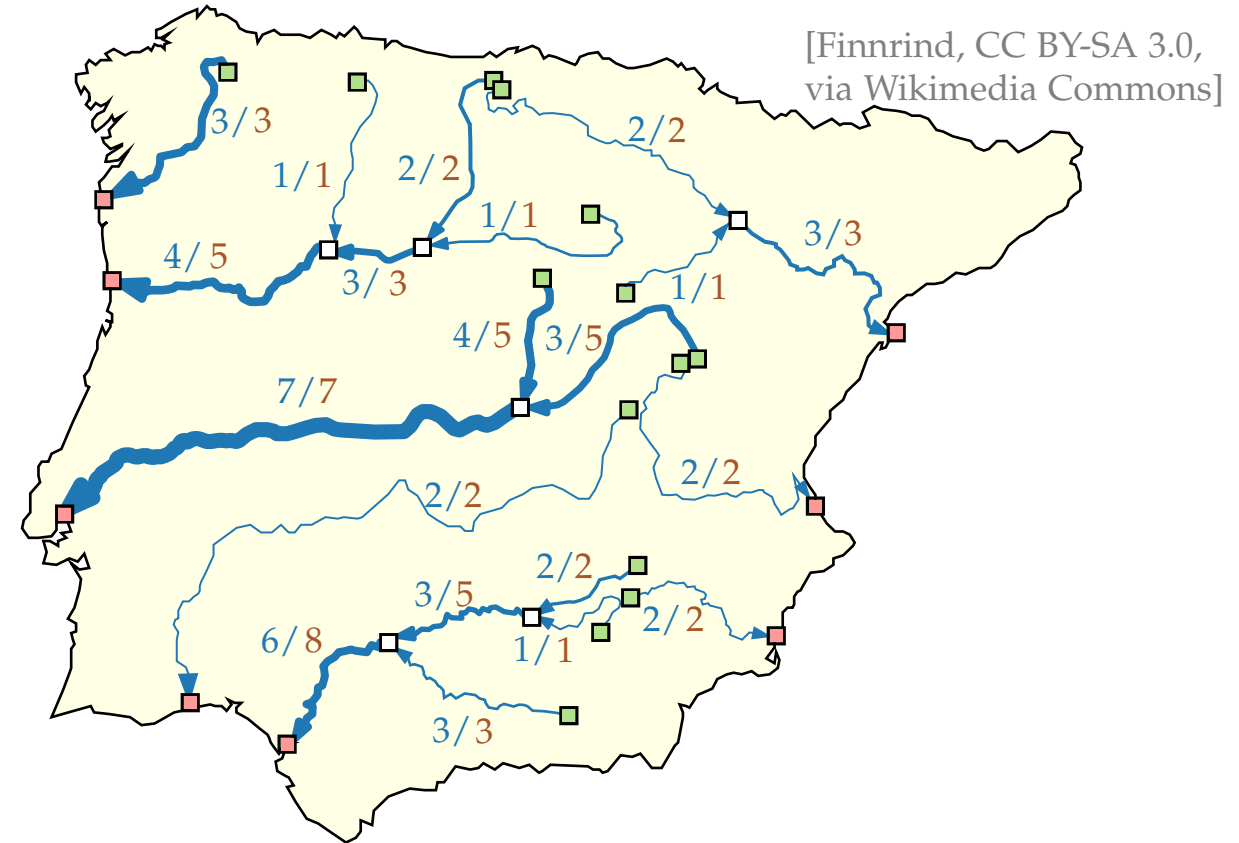
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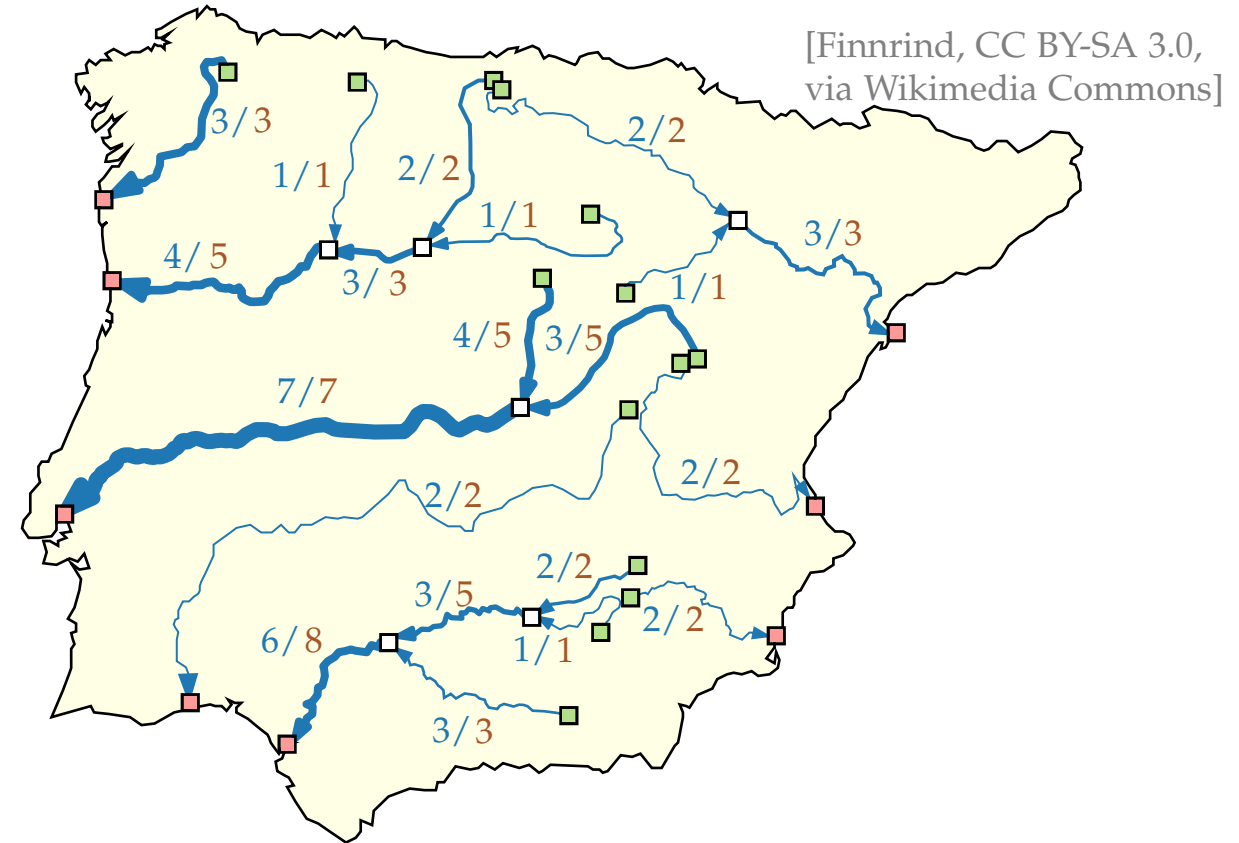
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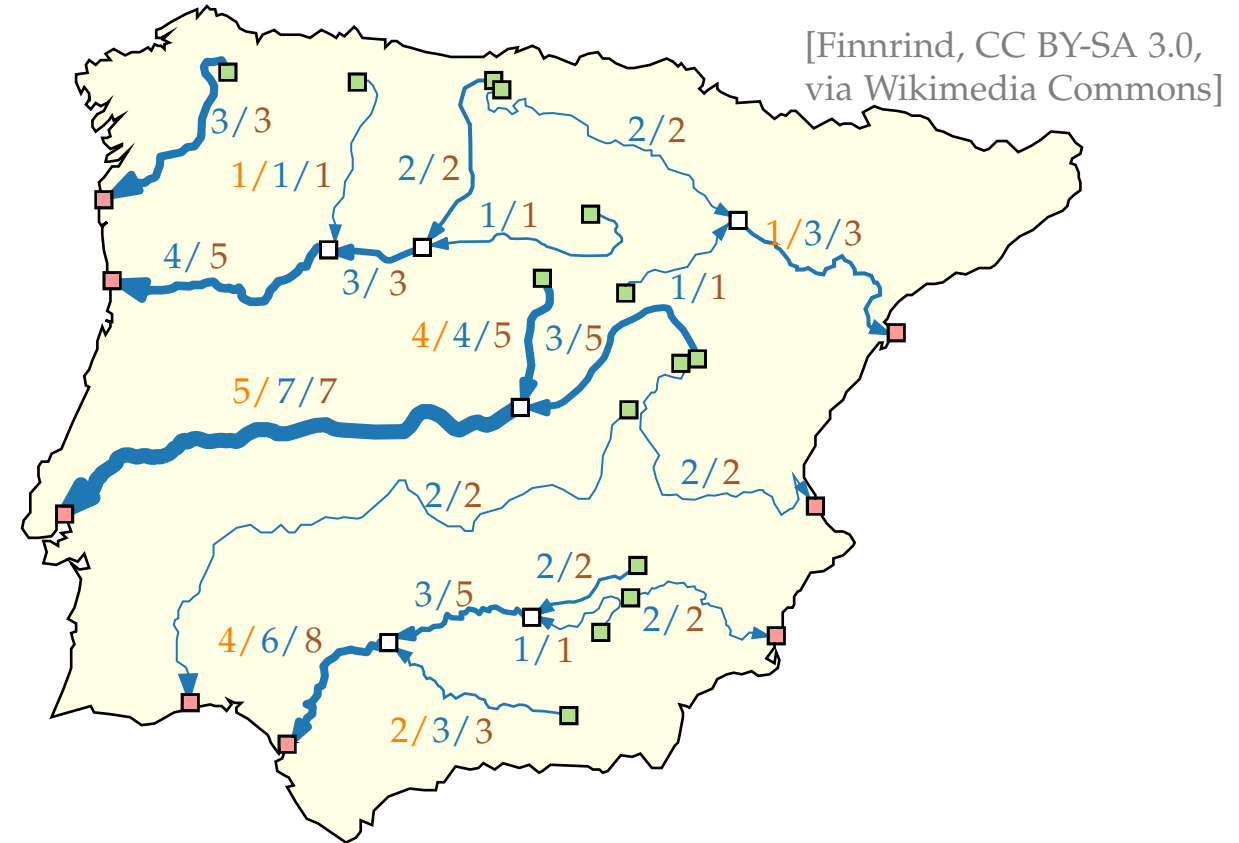
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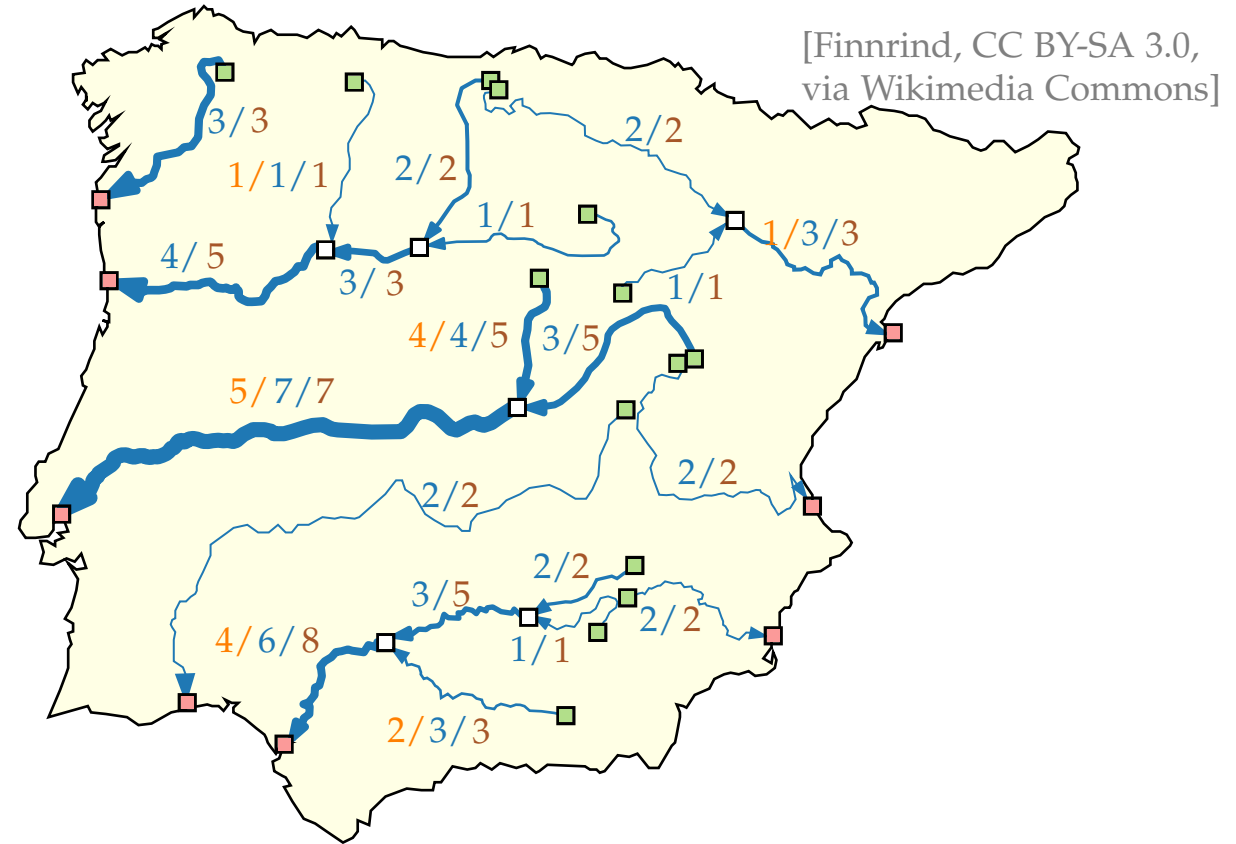
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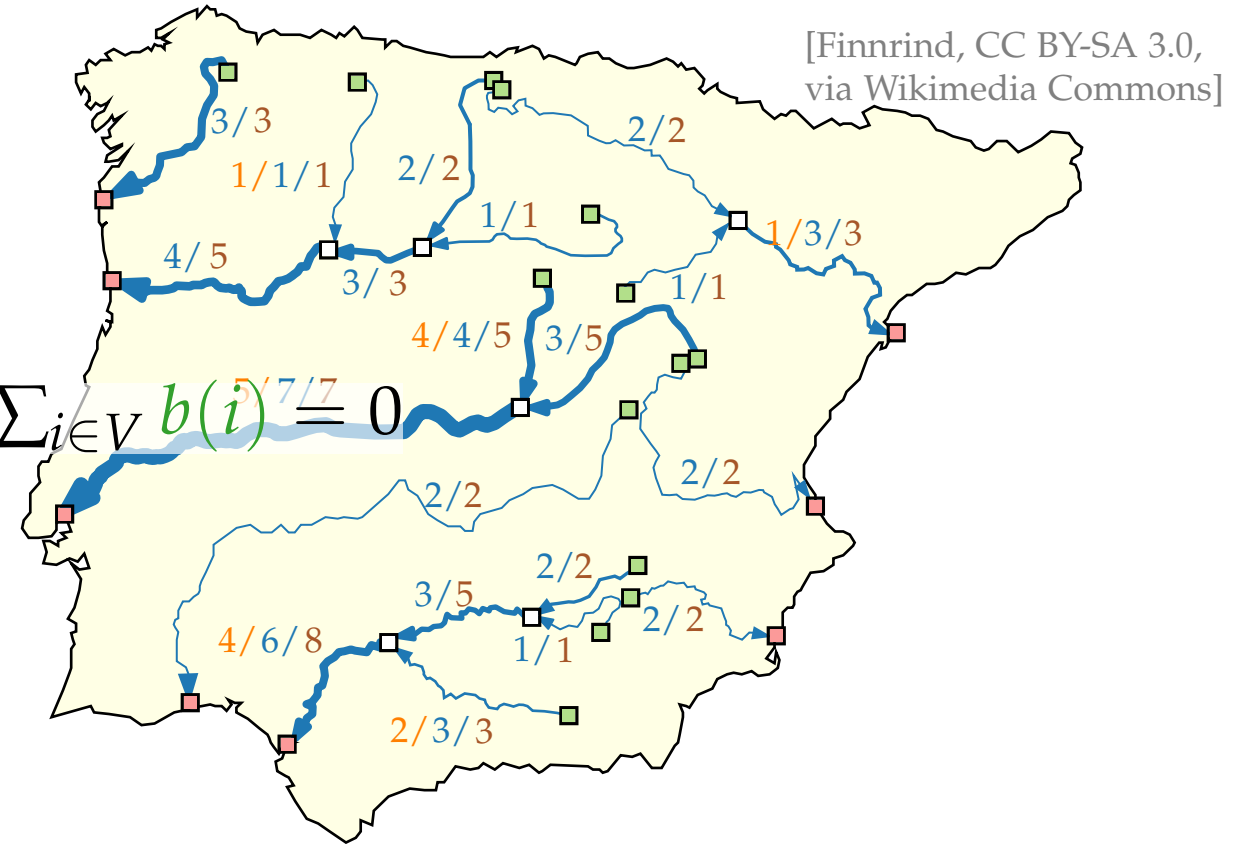
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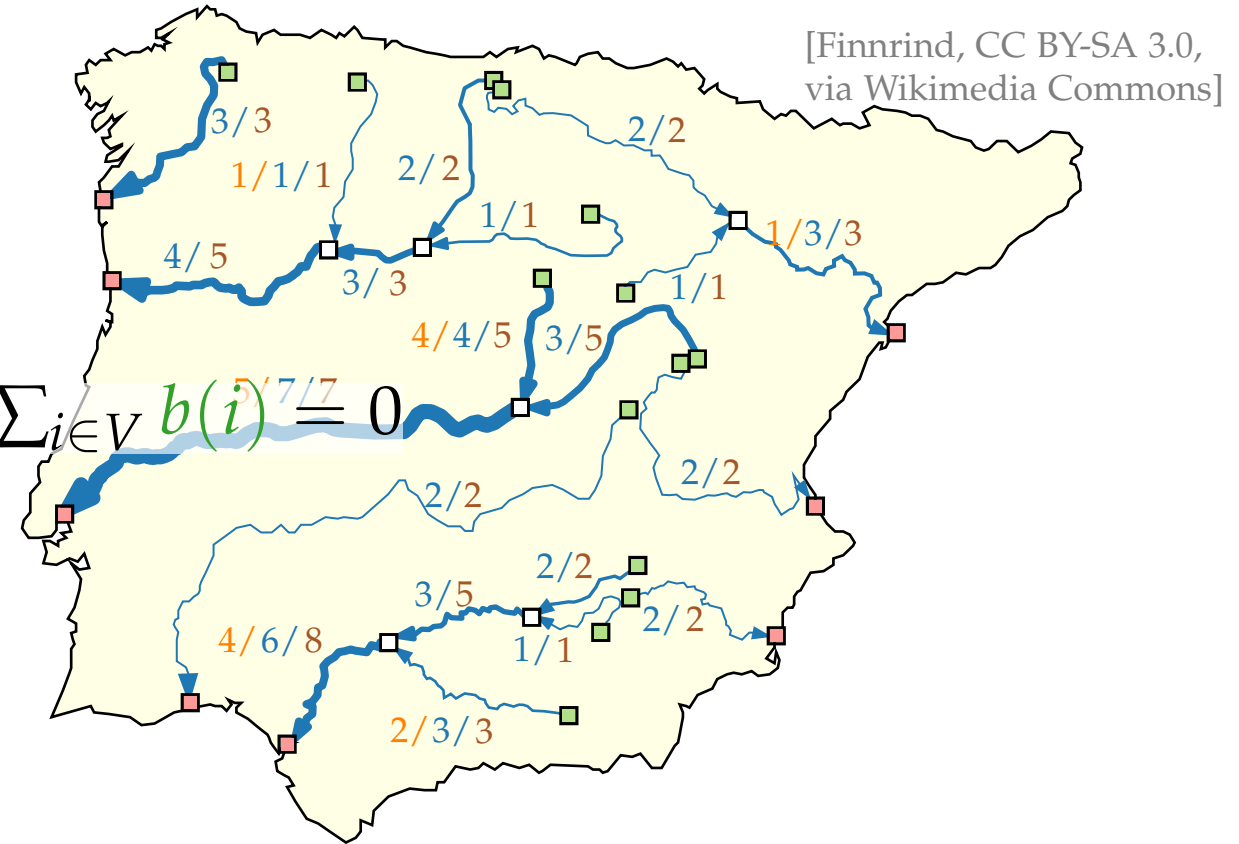
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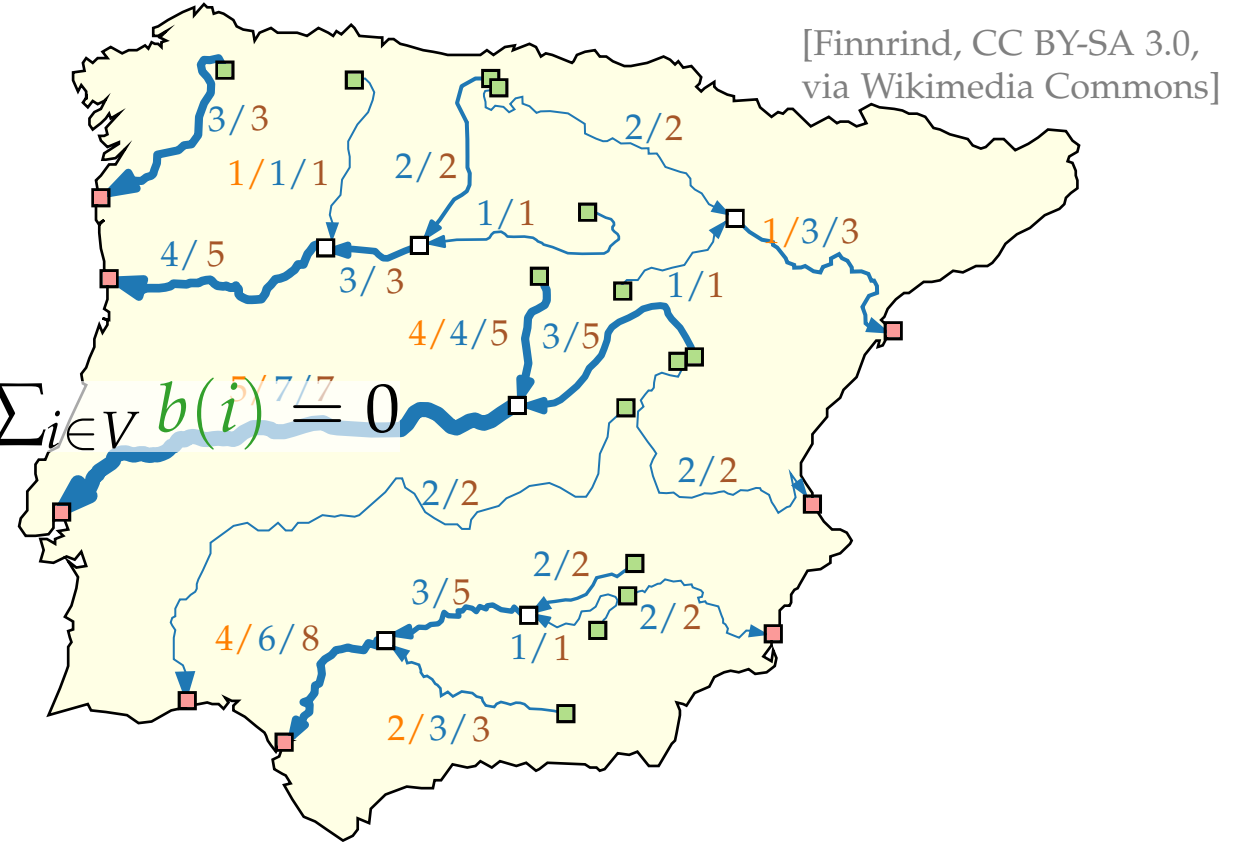
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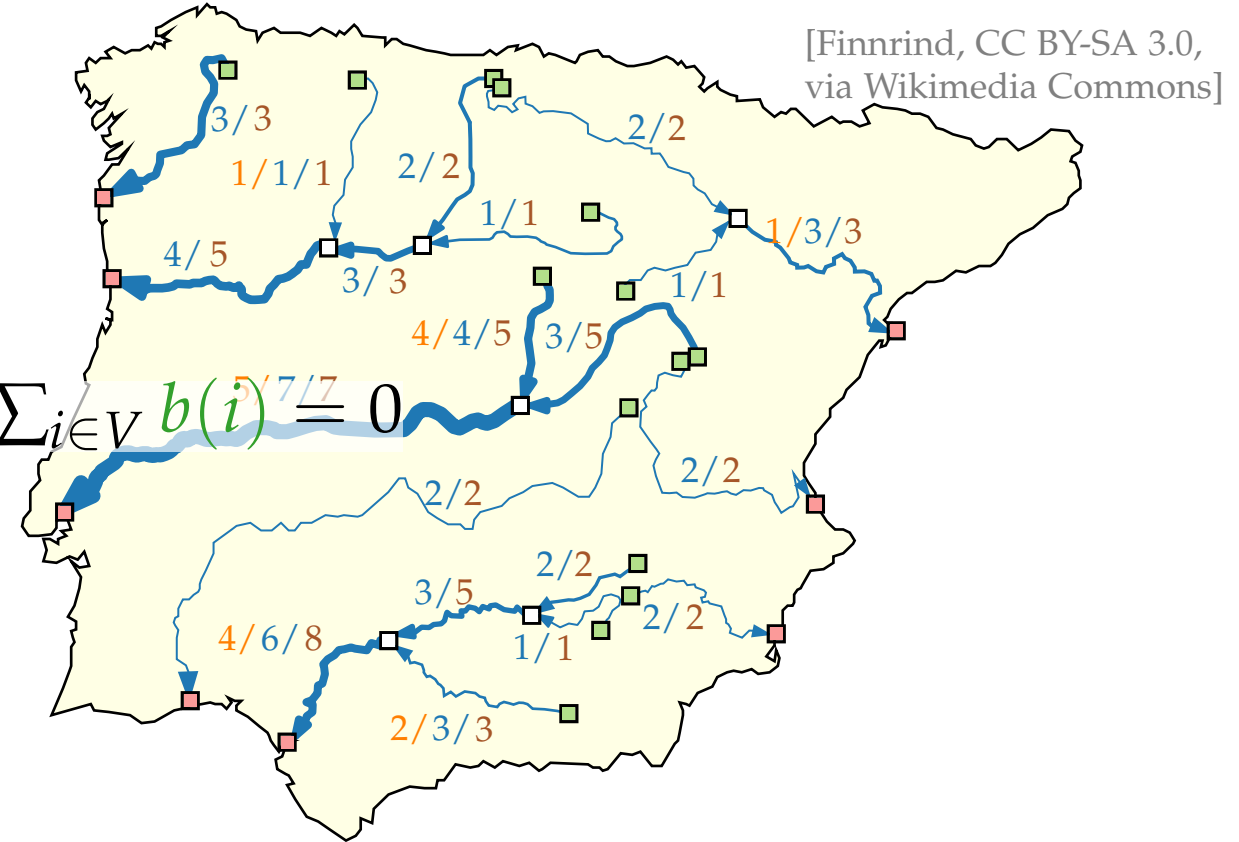
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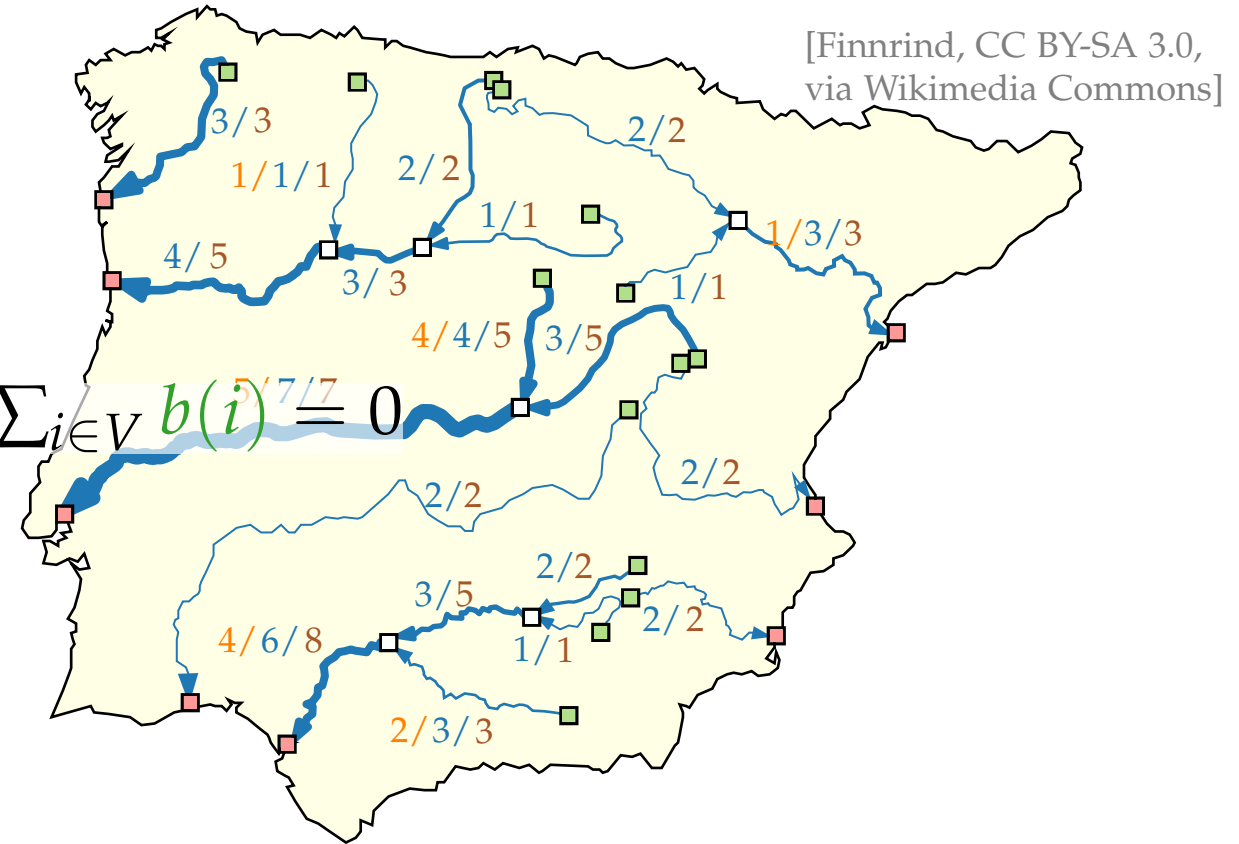
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# General Flow Network – Algorithms

## Polynomial Algorithms

#	Due to	Year	Running Time
1	Edmonds and Karp	1972	$O((n + m) \log U S(n, m, nC))$
2	Rock	1980	$O((n + m) \log U S(n, m, nC))$
3	Rock	1980	$O(n \log C M(n, m, U))$
4	Bland and Jensen	1985	$O(m \log C M(n, m, U))$
5	Goldberg and Tarjan	1987	$O(nm \log (n^2/m) \log (nC))$
6	Goldberg and Tarjan	1988	$O(nm \log n \log (nC))$
7	Ahuja, Goldberg, Orlin and Tarjan	1988	$O(nm \log \log U \log (nC))$

## Strongly Polynomial Algorithms

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6	Goldberg and Tarjan	1988	$O(nm^2 \log^2 n)$
7	Orlin (this paper)	1988	$O((n + m) \log n S(n, m))$

$S(n, m)$	= $O(m + n \log n)$	Fredman and Tarjan [1984]
$S(n, m, C)$	= $O(\text{Min}(m + n\sqrt{\log C}, (m \log \log C)))$	Ahuja, Mehlhorn, Orlin and Tarjan [1990] Van Emde Boas, Kaas and Zijlstra[1977]
$M(n, m)$	= $O(\text{min}(nm + n^{2+\epsilon}, nm \log n))$ where $\epsilon$ is any fixed constant.	King, Rao, and Tarjan [1991]
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## Theorem.

[Orlin 1991]

The minimum cost flow problem can be solved in  $O(n^2 \log^2 n + m^2 \log n)$  time.



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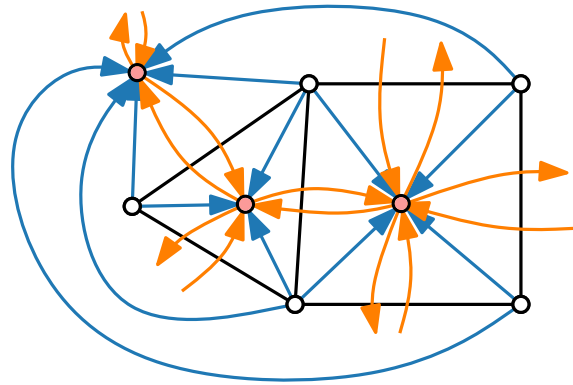
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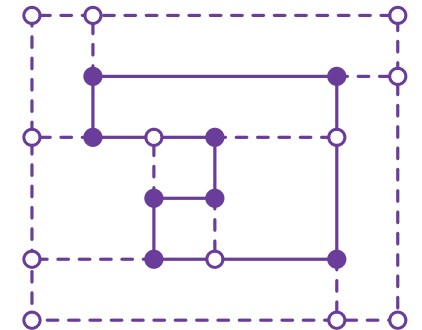
[Cornelsen & Karrenbauer 2011]

The minimum cost flow problem for planar graphs with bounded costs and face sizes can be solved in  $O(n^{3/2})$  time.

# Visualization of Graphs

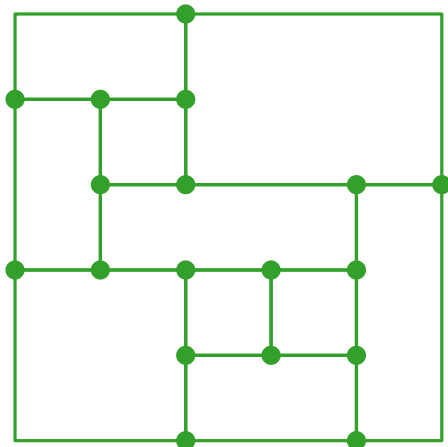


## Lecture 6: Orthogonal Layouts



## Part IV: Bend Minimization

Philipp Kindermann



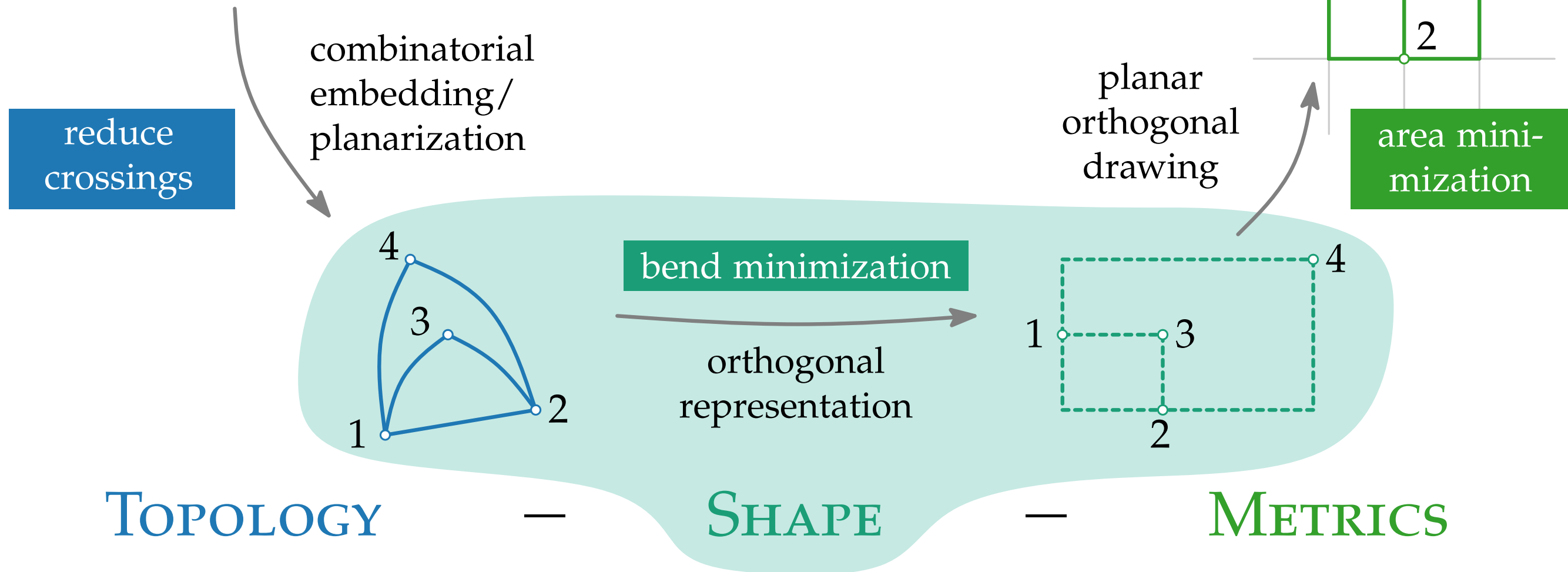
# Topology – Shape – Metrics

Three-step approach:

[Tamassia 1987]

$$V = \{v_1, v_2, v_3, v_4\}$$

$$E = \{v_1v_2, v_1v_3, v_1v_4, v_2v_3, v_2v_4\}$$



# Bend Minimization with Given Embedding

Geometric bend minimization.

Given:

Find:

# Bend Minimization with Given Embedding

## Geometric bend minimization.

Given: ■ Plane graph  $G = (V, E)$  with maximum degree 4

Find:

# Bend Minimization with Given Embedding

## Geometric bend minimization.

Given:

- Plane graph  $G = (V, E)$  with maximum degree 4
- Combinatorial embedding  $F$  and outer face  $f_0$

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# Bend Minimization with Given Embedding

## Geometric bend minimization.

Given: ■ Plane graph  $G = (V, E)$  with maximum degree 4

■ Combinatorial embedding  $F$  and outer face  $f_0$

Find: Orthogonal drawing with minimum number of bends that preserves the embedding.

# Bend Minimization with Given Embedding

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Given: ■ Plane graph  $G = (V, E)$  with maximum degree 4

■ Combinatorial embedding  $F$  and outer face  $f_0$

Find: Orthogonal drawing with minimum number of bends that preserves the embedding.

Compare with the following variation.

## Combinatorial bend minimization.

Given:

Find:



# Bend Minimization with Given Embedding

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Given: ■ Plane graph  $G = (V, E)$  with maximum degree 4

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Compare with the following variation.

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## Combinatorial bend minimization.

Given: ■ Plane graph  $G = (V, E)$  with maximum degree 4

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Find: **Orthogonal representation**  $H(G)$  with minimum number of bends that preserves the embedding.

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## Idea.

Formulate as a network flow problem:

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- a unit of flow =  $\angle \frac{\pi}{2}$

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Formulate as a network flow problem:

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- vertices  $\xrightarrow{\sphericalangle}$  faces ( $\# \sphericalangle \frac{\pi}{2}$  per face)

# Combinatorial Bend Minimization

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## Idea.

Formulate as a network flow problem:

- a unit of flow =  $\sphericalangle \frac{\pi}{2}$
- vertices  $\xrightarrow{\sphericalangle}$  faces ( $\# \sphericalangle \frac{\pi}{2}$  per face)
- faces  $\xrightarrow{\sphericalangle}$  neighbouring faces ( $\#$  bends toward the neighbour)

# Flow Network for Bend Minimization

(H1)  $H(G)$  corresponds to  $F, f_0$ .

(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$ , sequence  $\delta_1$  is reversed and inverted  $\delta_2$ .

(H3) For each **face**  $f$  it holds that:

$$\sum_{r \in H(f)} C(r) = \begin{cases} -4 & \text{if } f = f_0 \\ +4 & \text{otherwise.} \end{cases}$$

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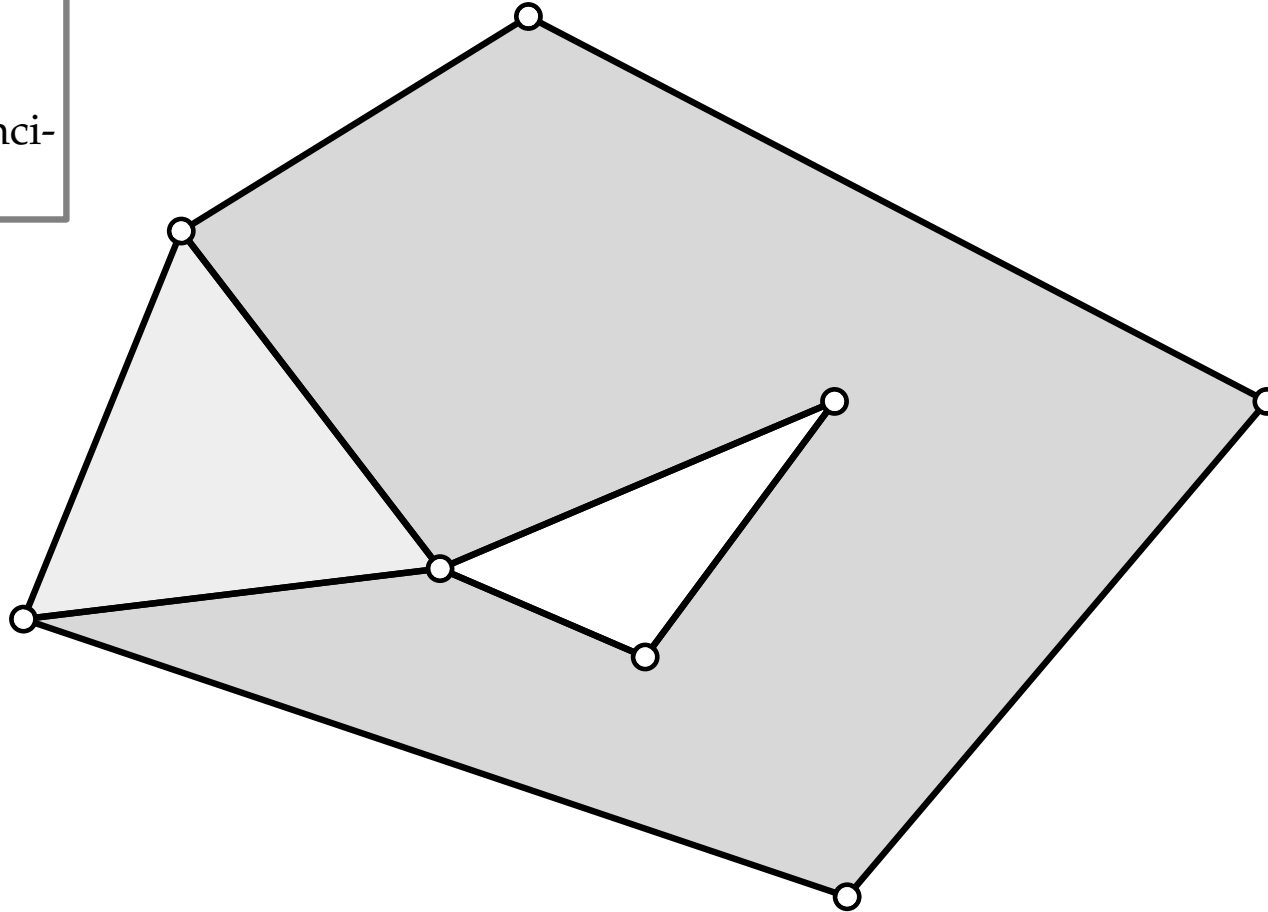
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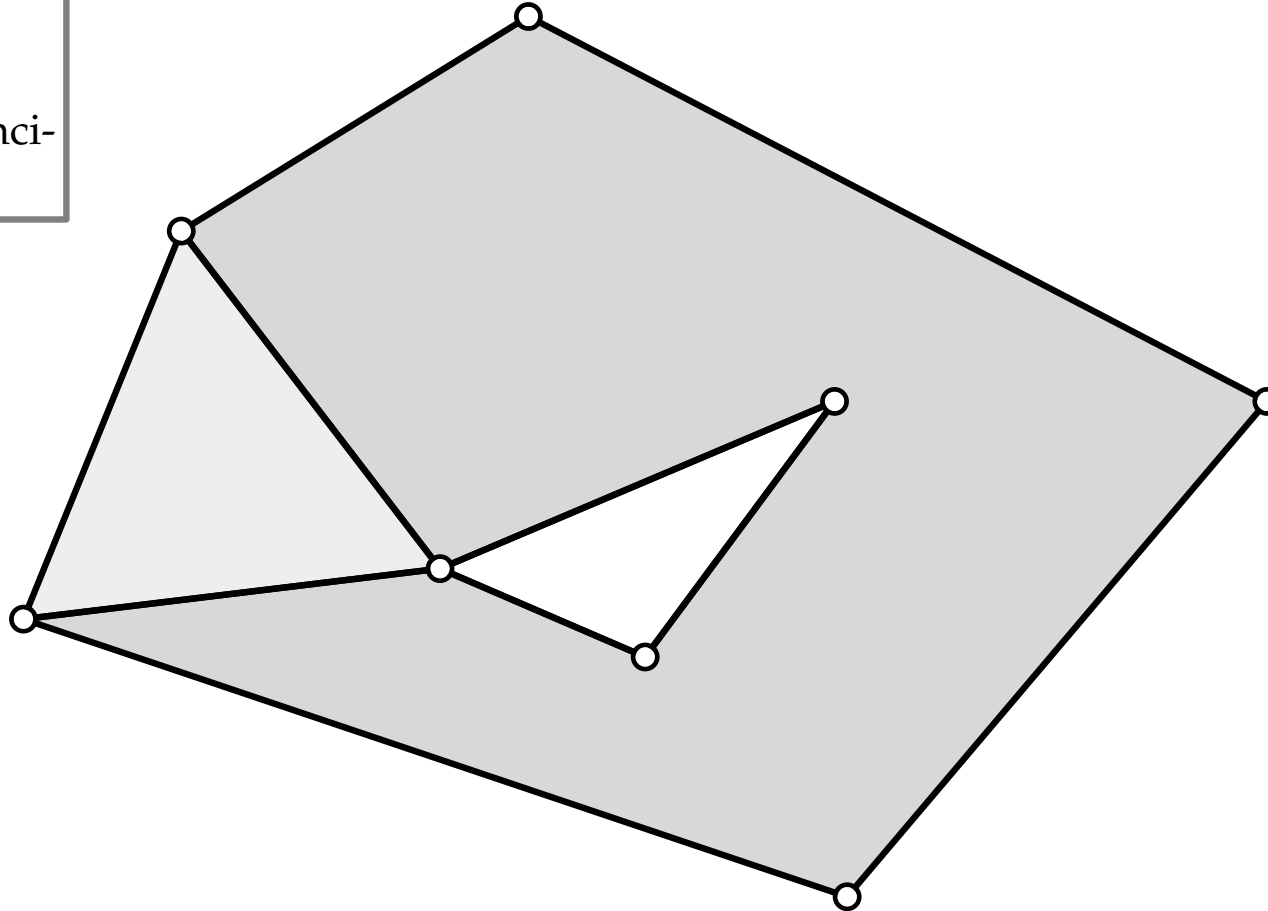
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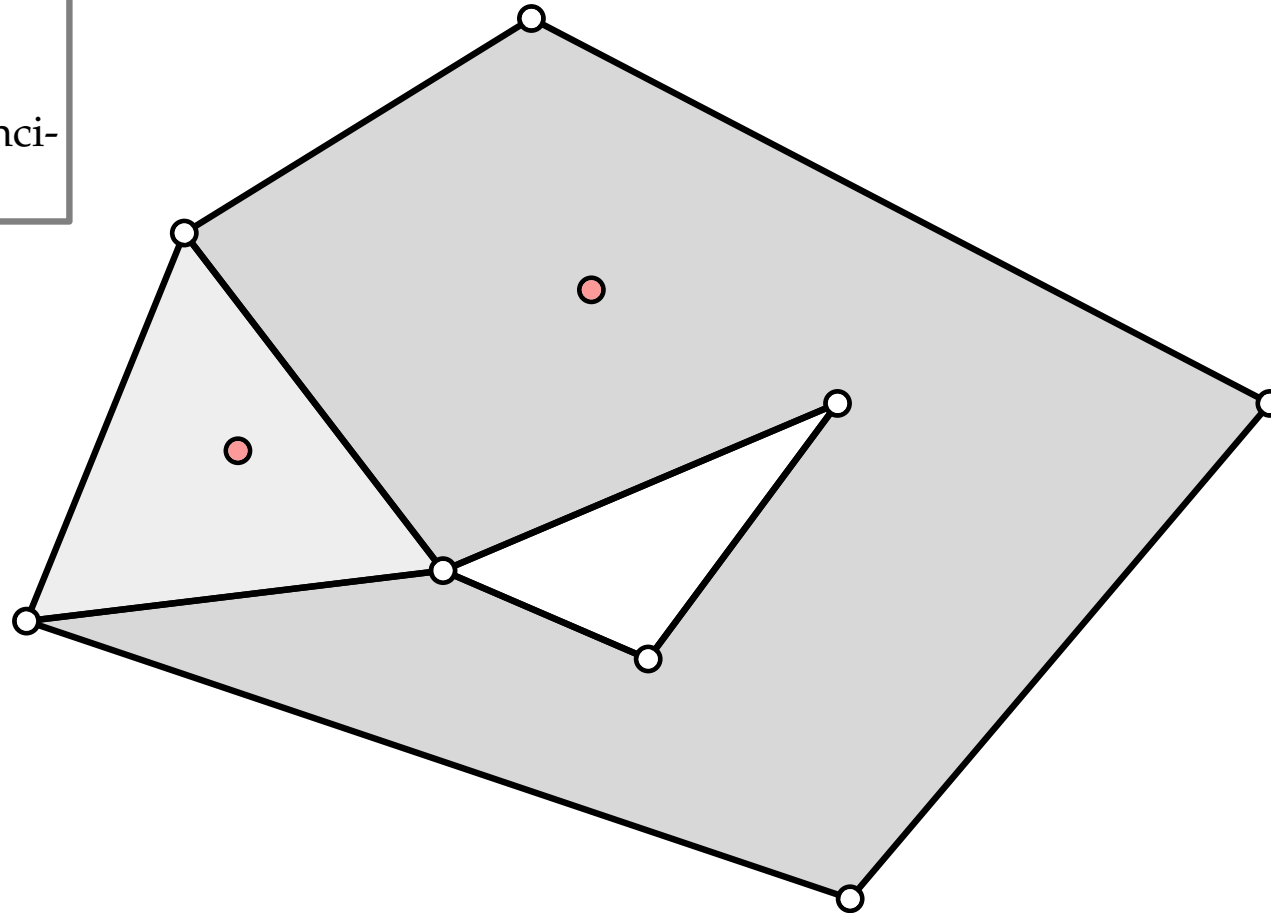
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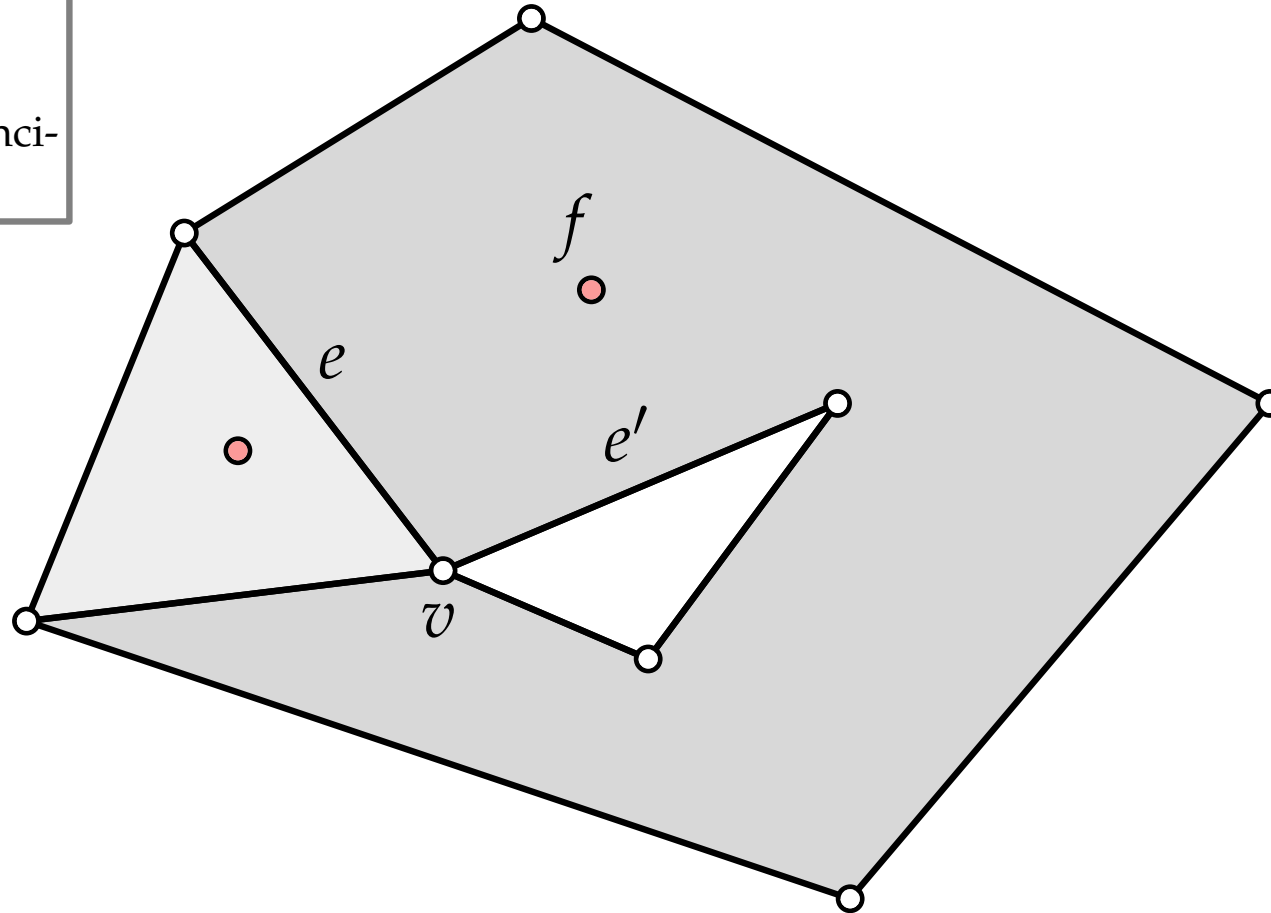
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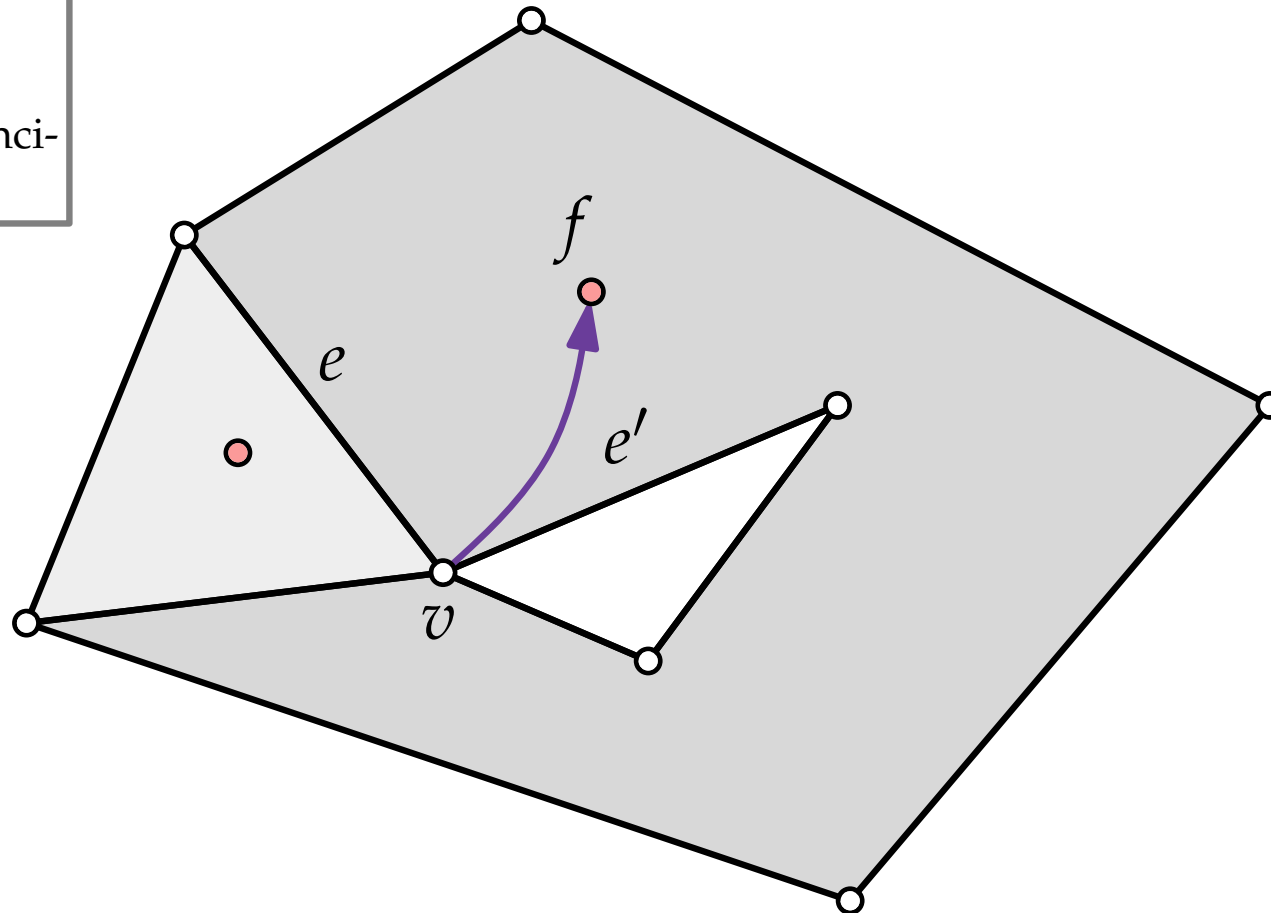
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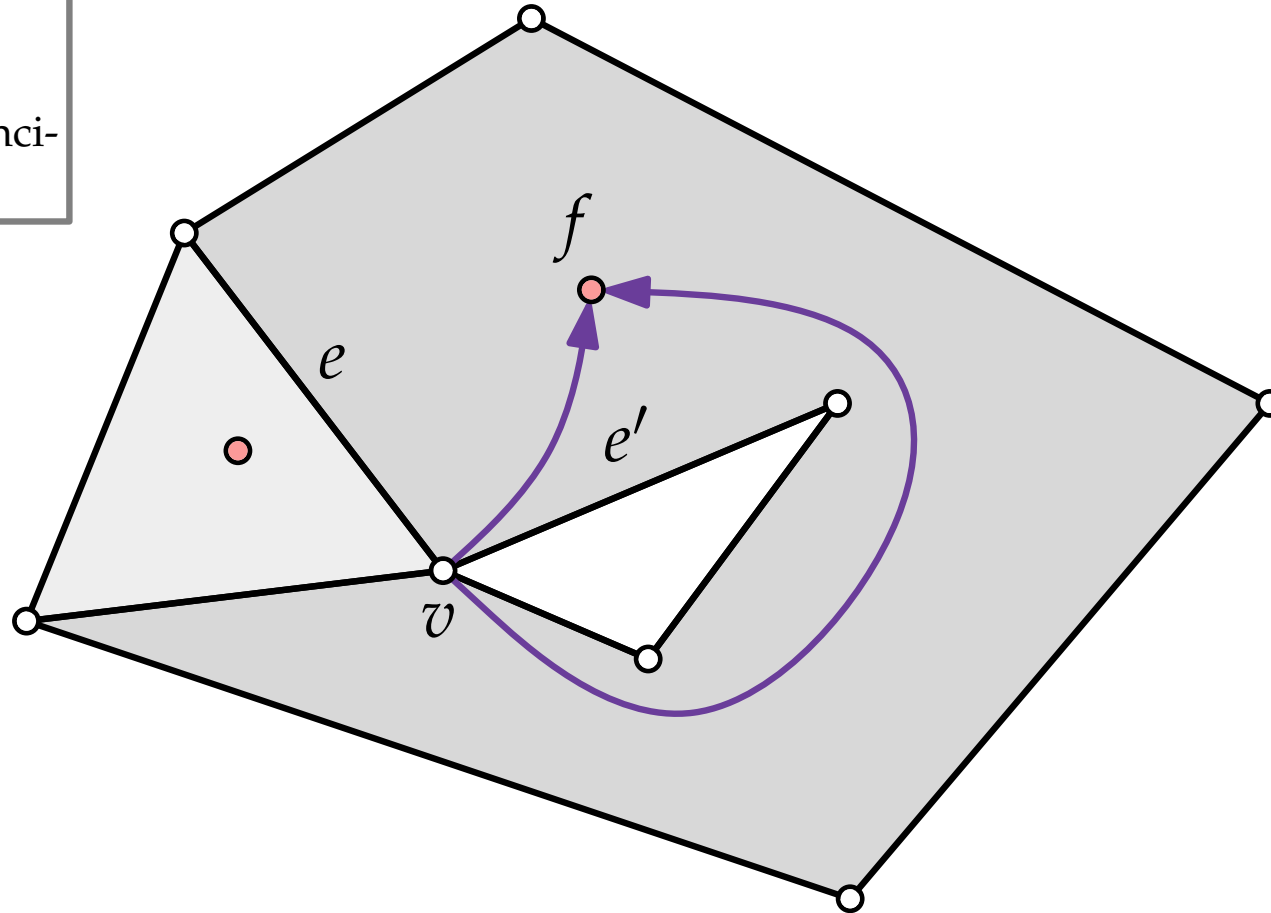
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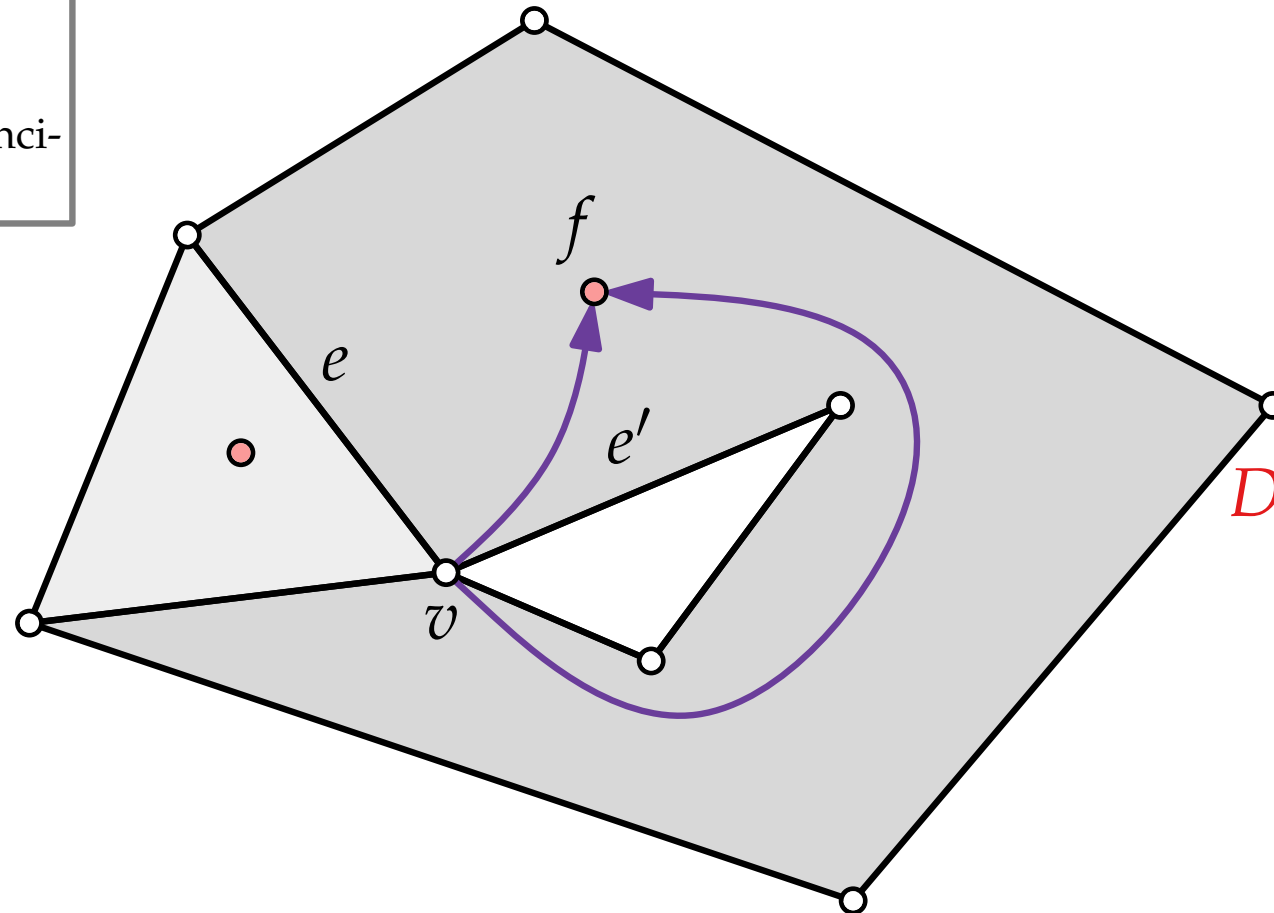
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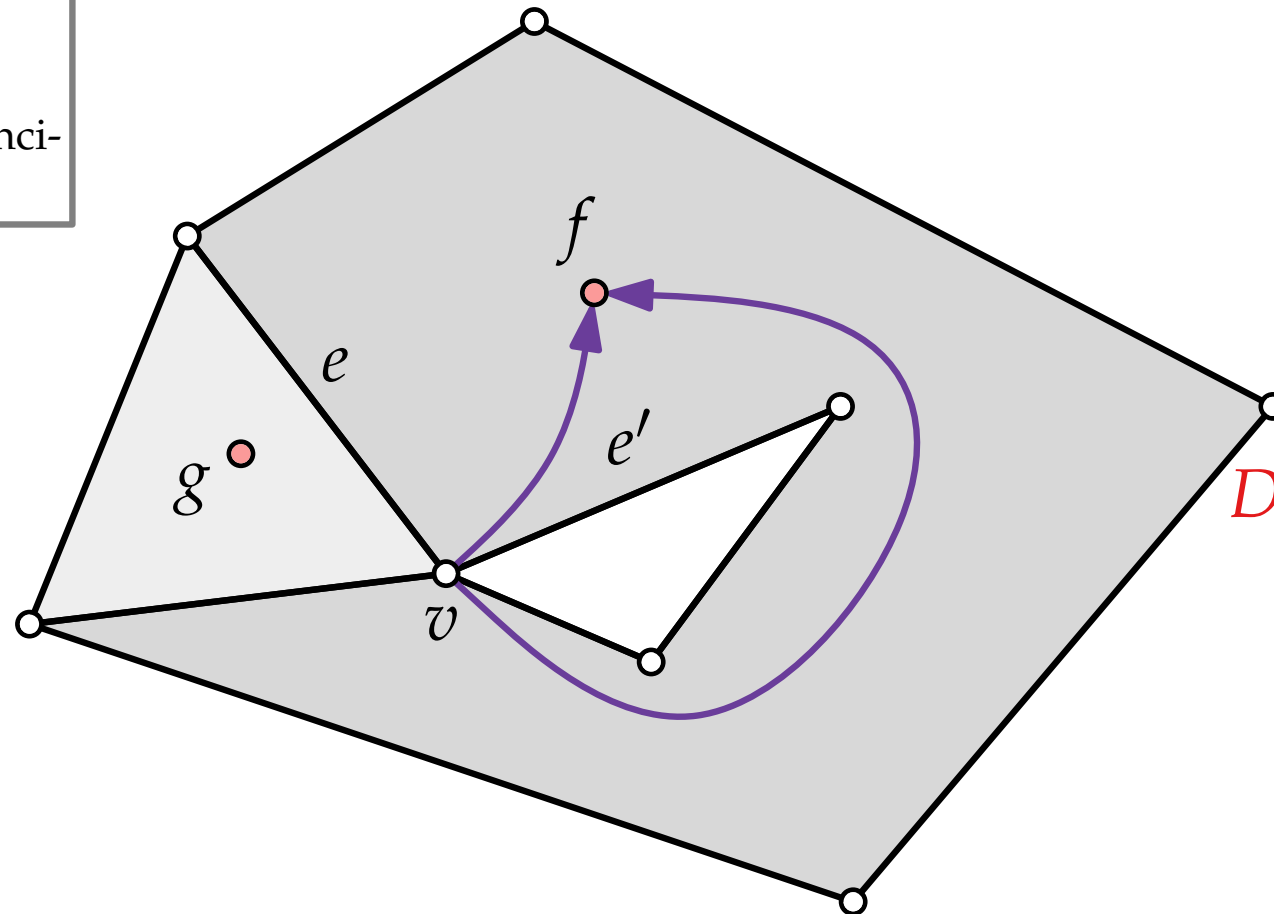
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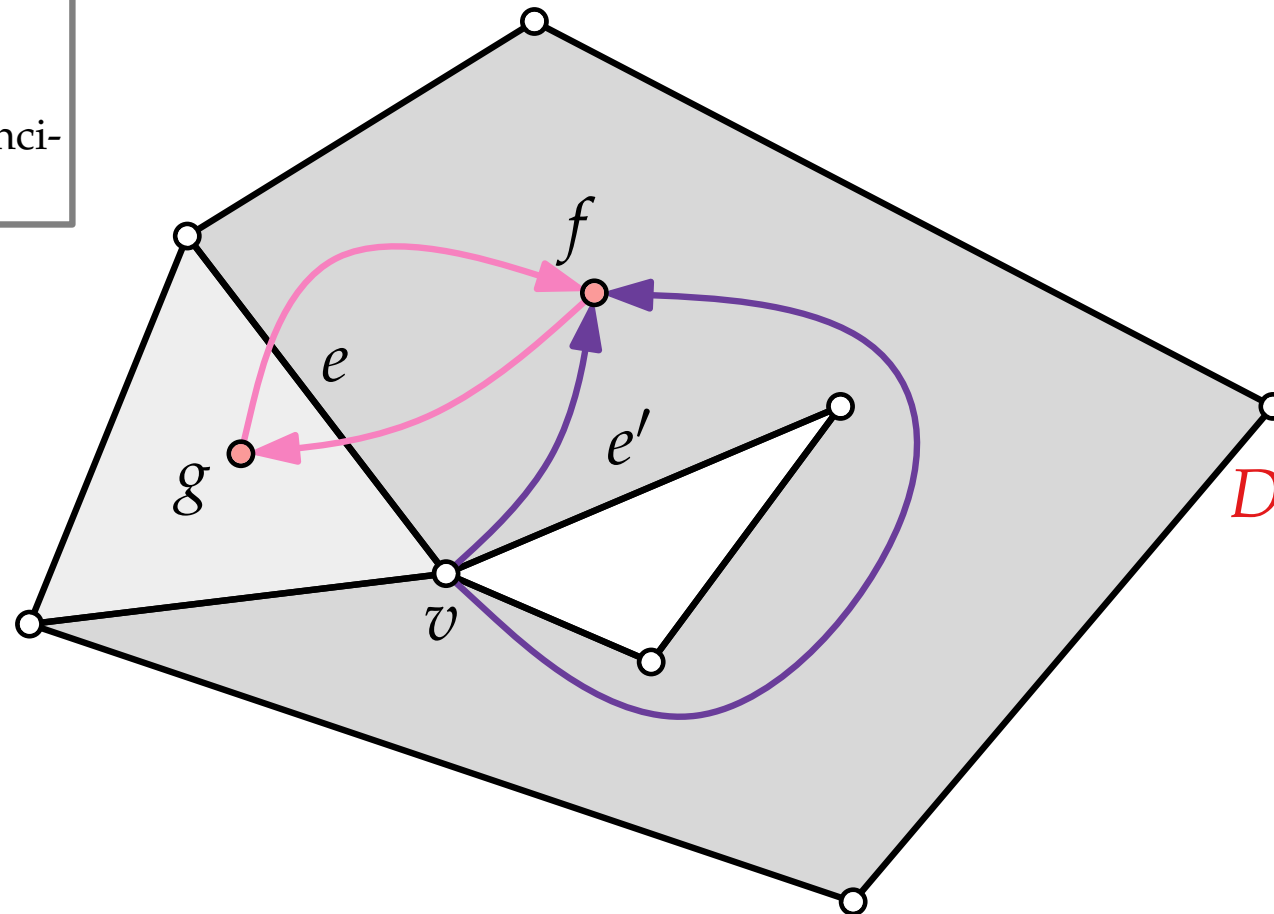
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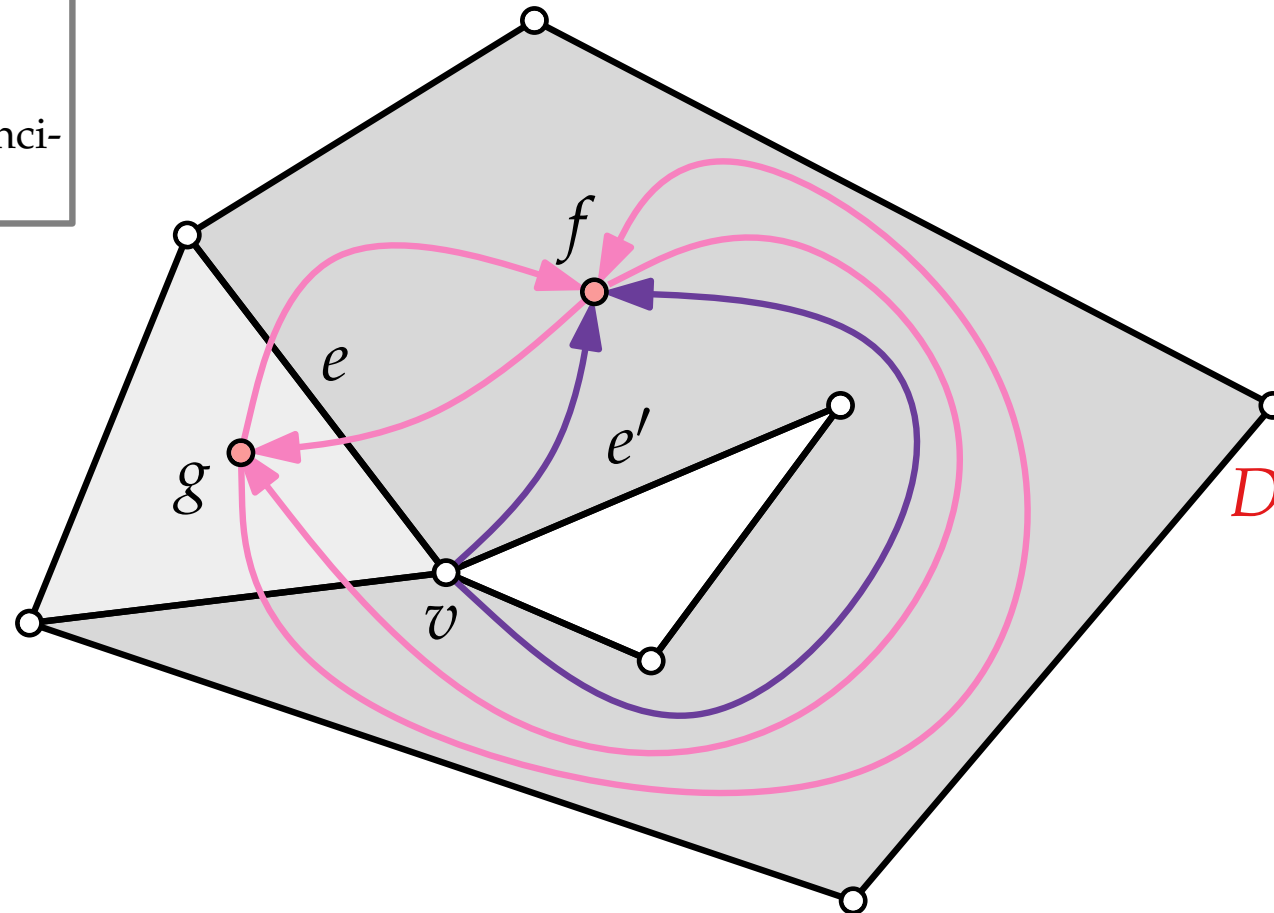
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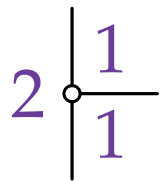
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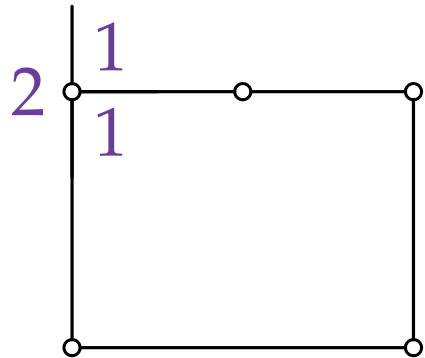
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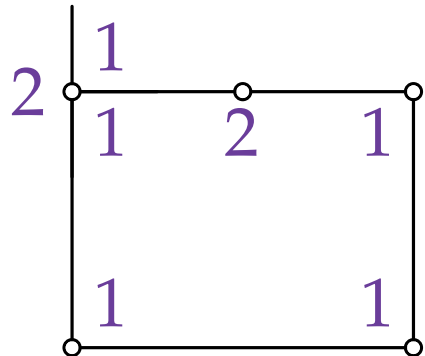
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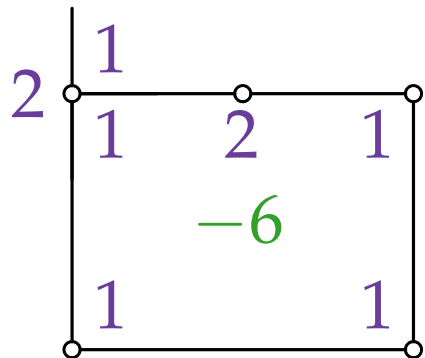
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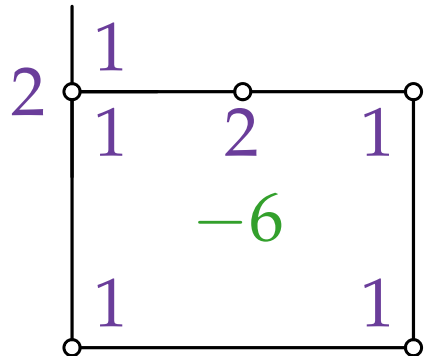
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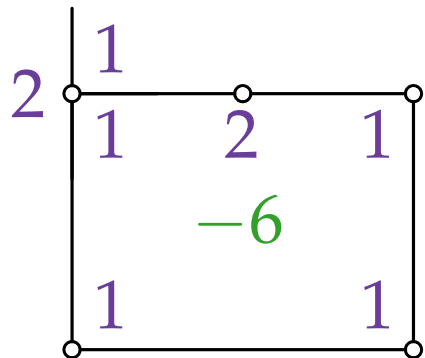
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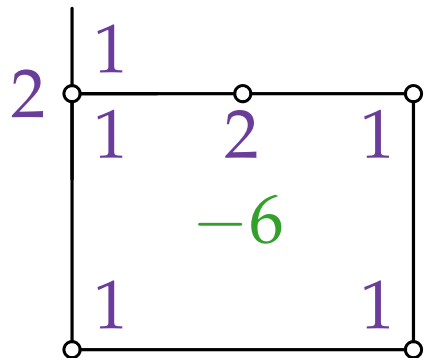
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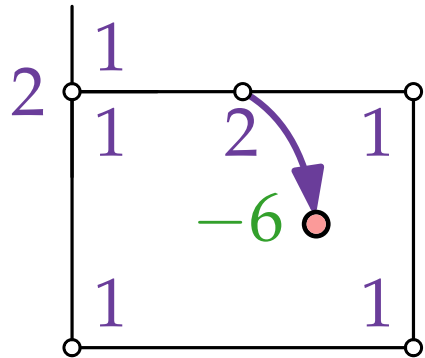
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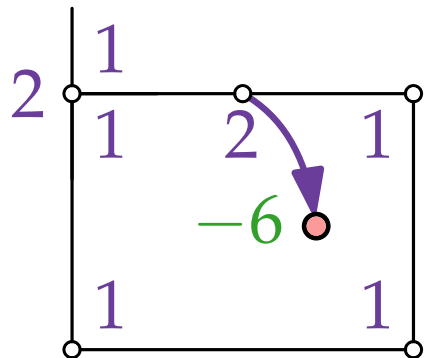
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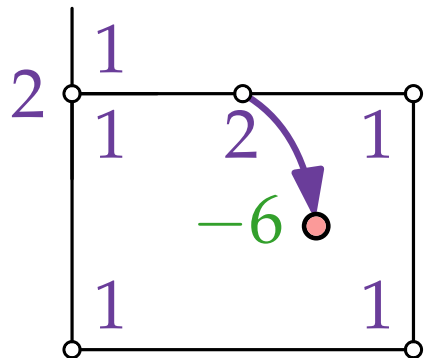
$$\blacksquare b(v) = 4 \quad \forall v \in V$$

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$$\forall (v, f) \in E, v \in V, f \in F$$

$$\ell(v, f) := 1 \leq X(v, f) \leq 4 =: u(v, f)$$

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# Flow Network for Bend Minimization

(H1)  $H(G)$  corresponds to  $F, f_0$ .

(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$ , sequence  $\delta_1$  is reversed and inverted  $\delta_2$ .

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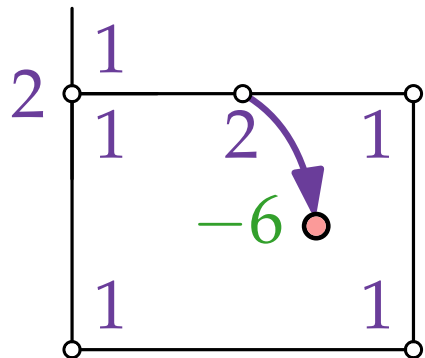
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# Flow Network for Bend Minimization

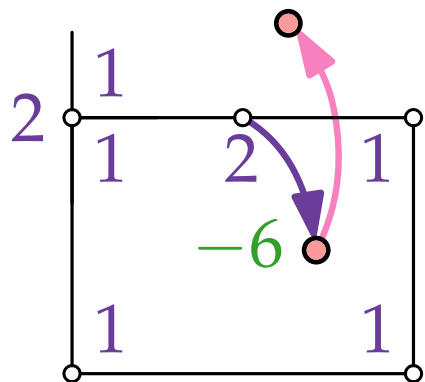
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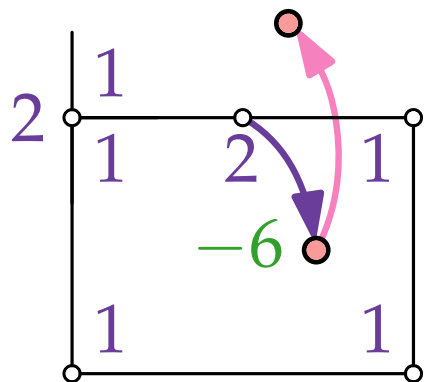
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# Flow Network for Bend Minimization

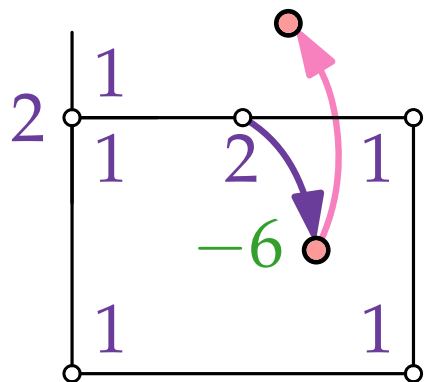
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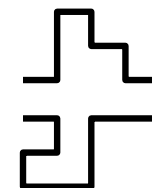
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We model only the number of bends.  
Why is it enough?



# Flow Network for Bend Minimization

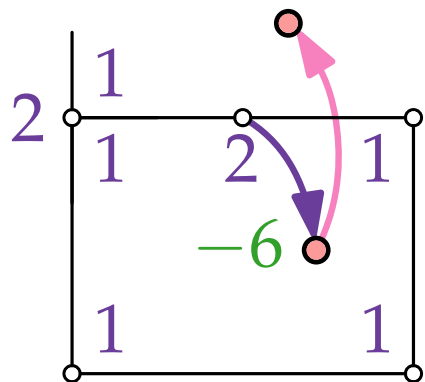
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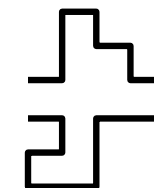
$$\text{cost}(v, f) = 0$$

$$\forall (f, g) \in E, f, g \in F$$

$$\ell(f, g) := 0 \leq X(f, g) \leq \infty =: u(f, g)$$

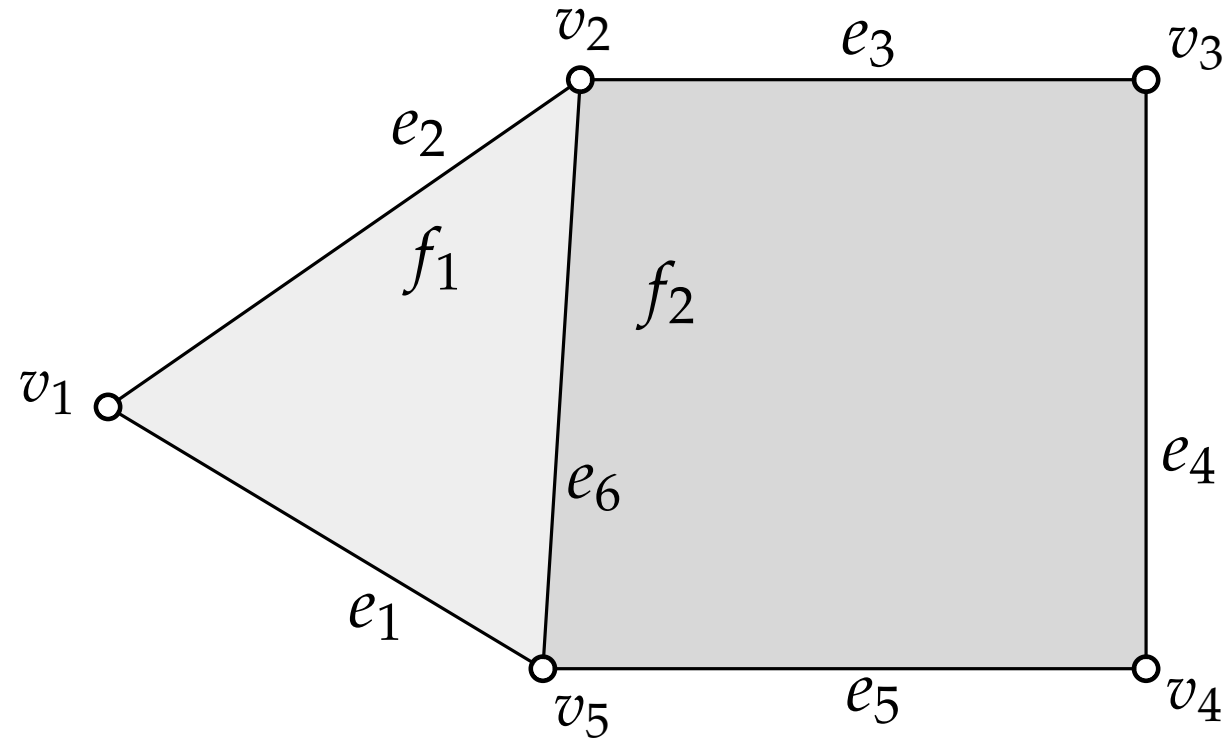
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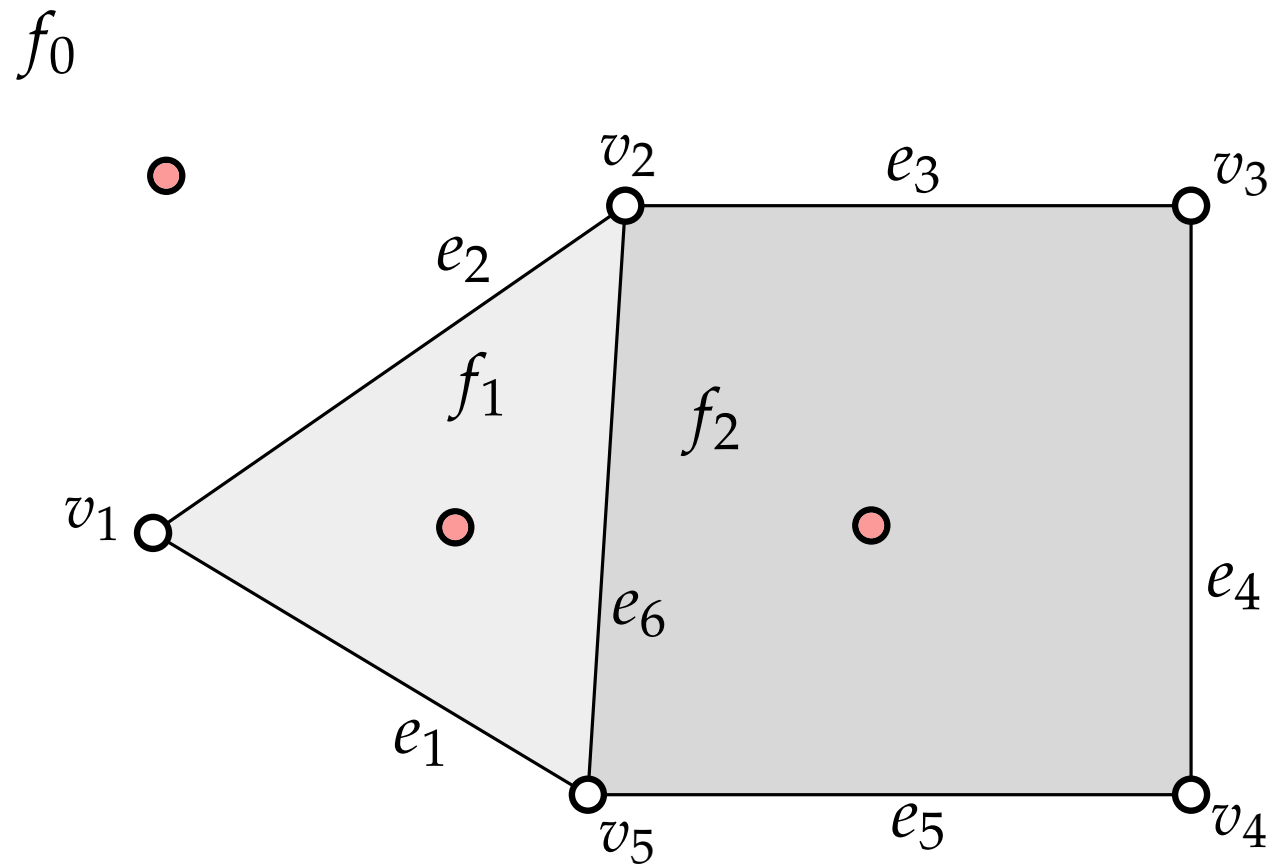


→ Exercise

# Flow Network Example

 $f_0$ 

# Flow Network Example

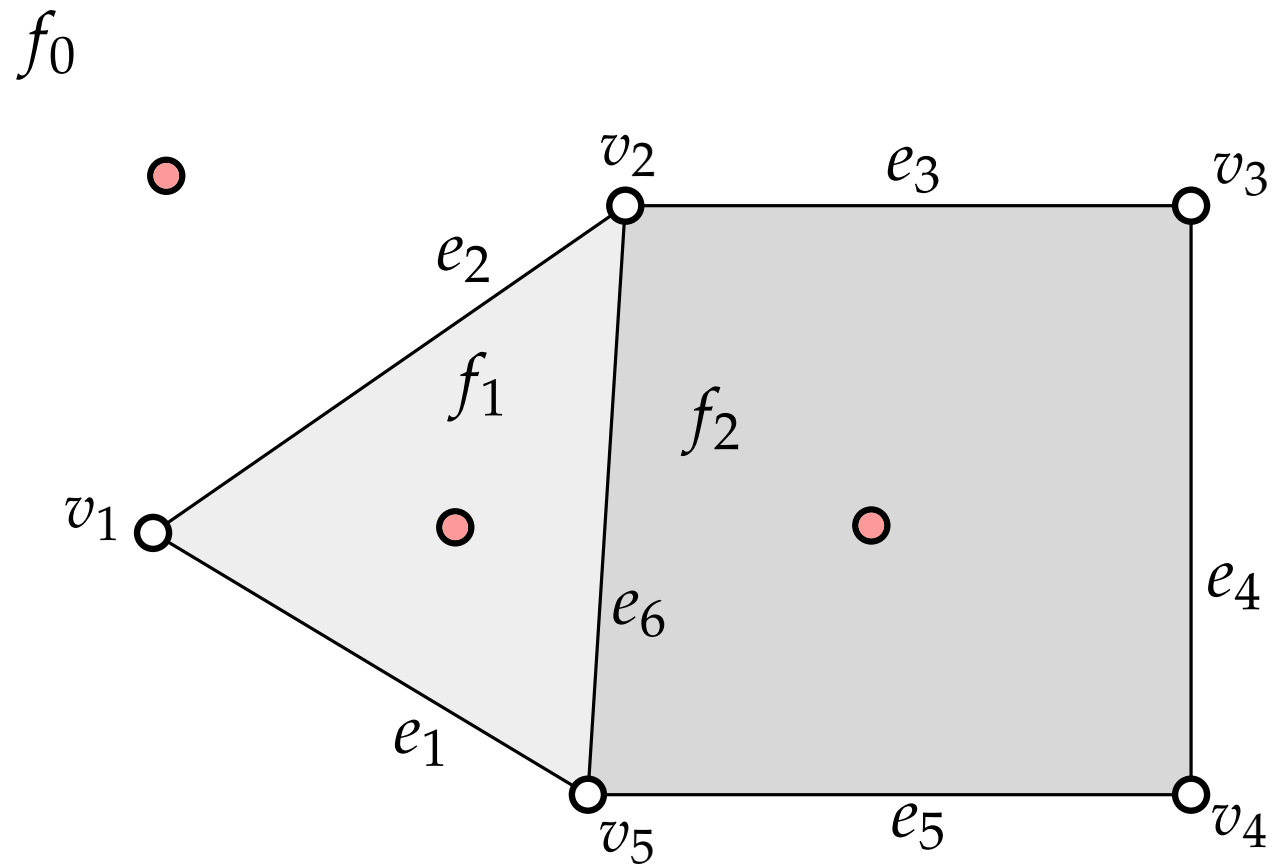


Legend

$V$  ○

$F$  ●

# Flow Network Example



Legend

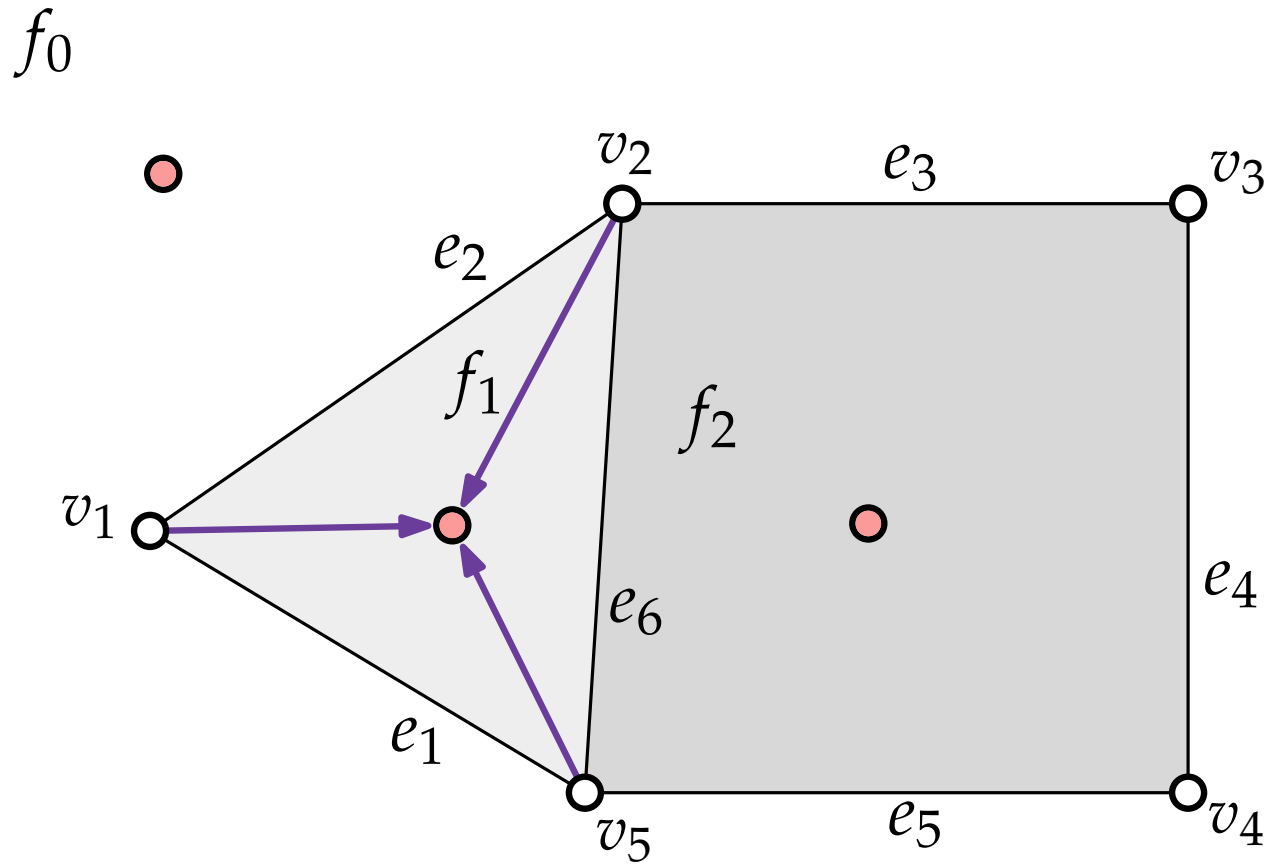
$V$  ○

$F$  ●

$\ell/u/\text{cost}$

$V \times F \supseteq \xrightarrow{1/4/0}$

# Flow Network Example



Legend

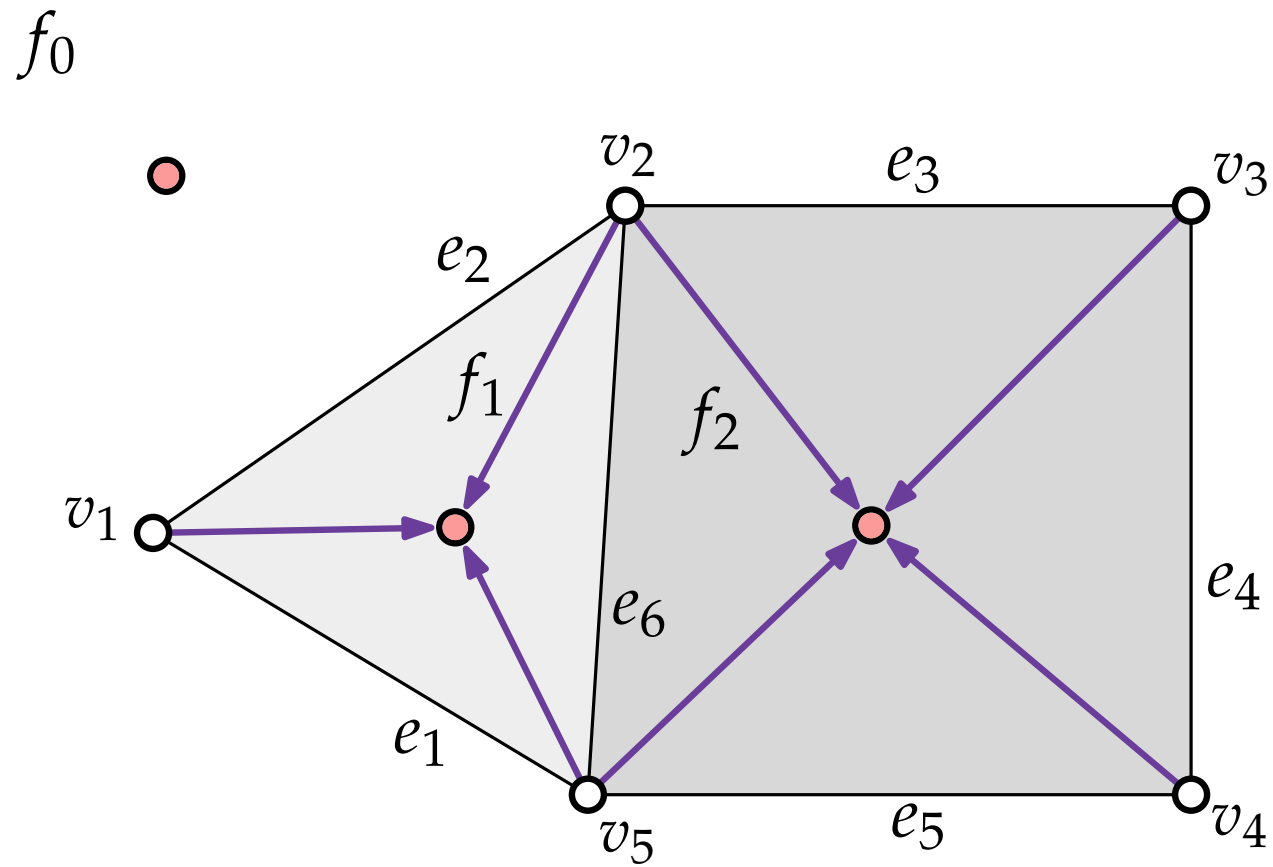
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$F$  ●

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# Flow Network Example



Legend

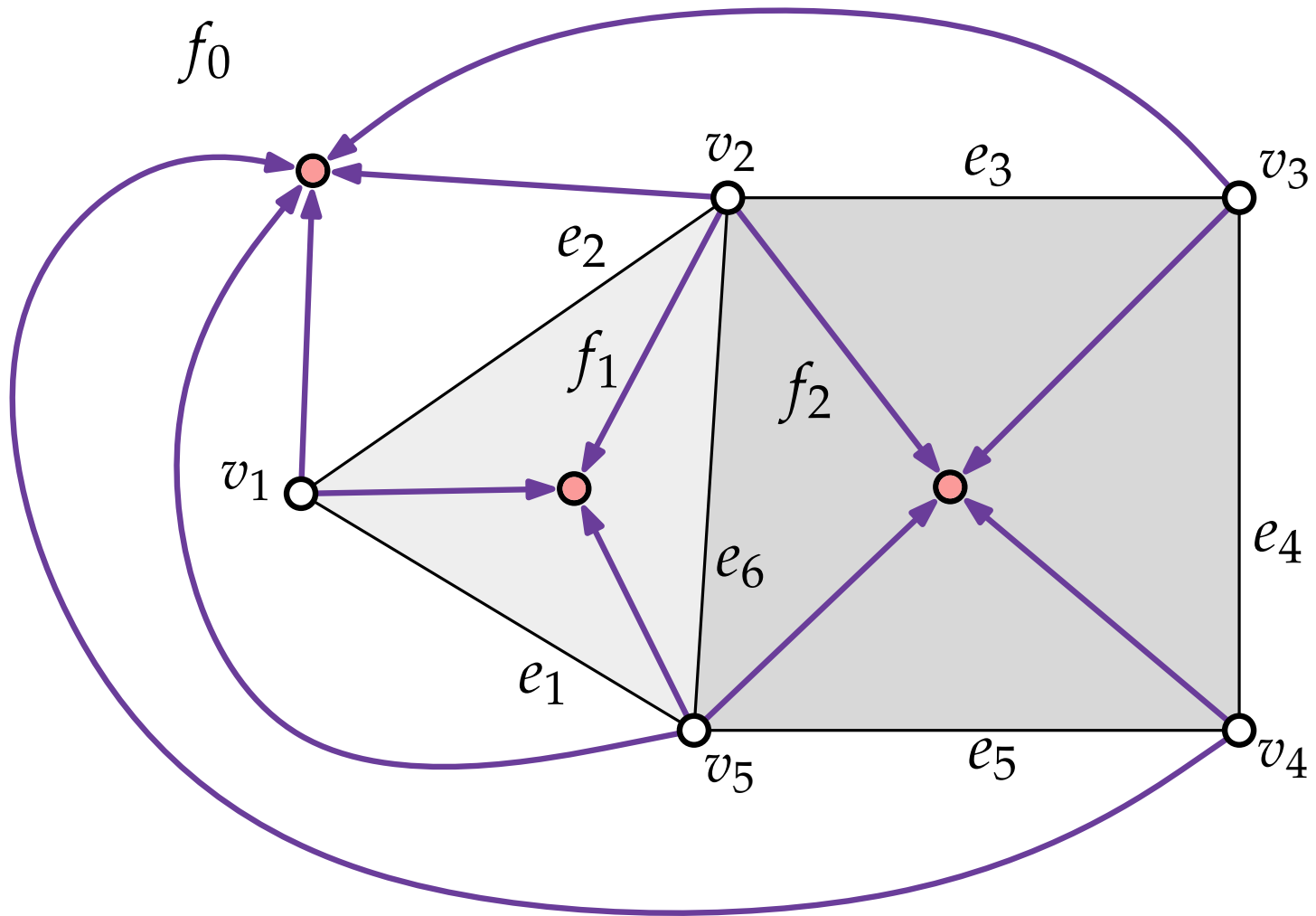
$V$  ○

$F$  ●

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# Flow Network Example



Legend

$V$  ○

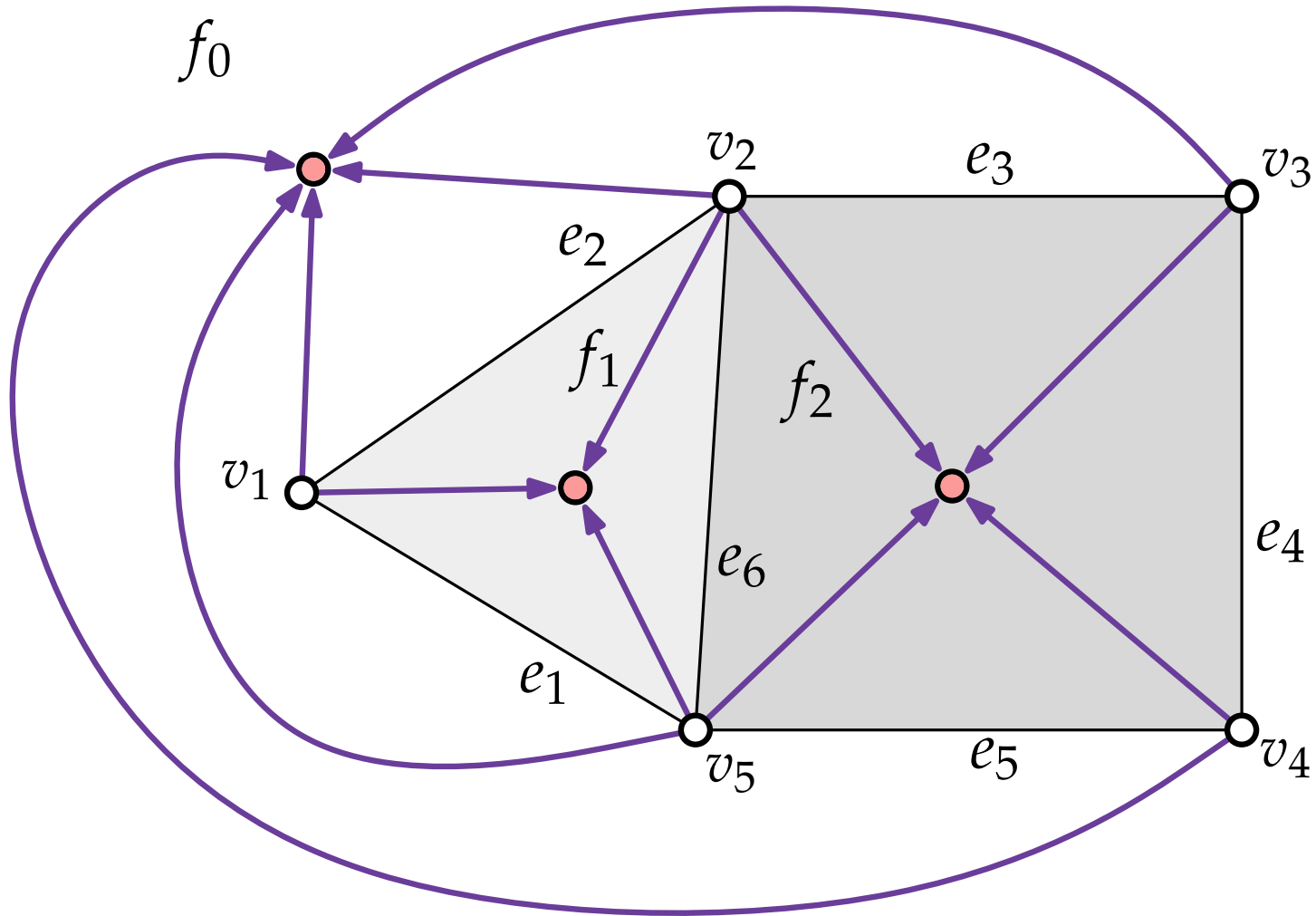
$F$  ●

$\ell/u/\text{cost}$

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# Flow Network Example



## Legend

$V$  ○

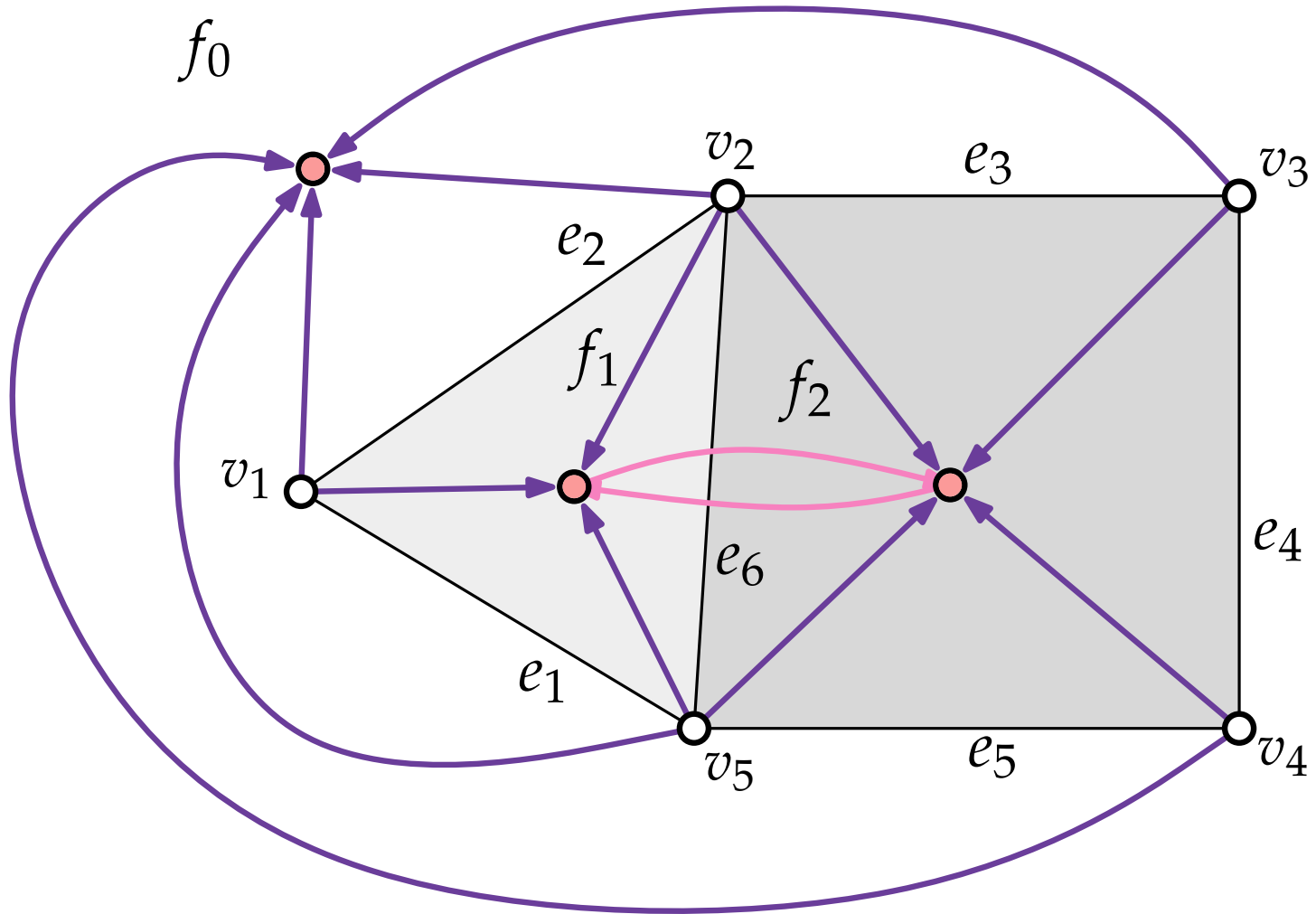
$F$  ●

$\ell/u/\text{cost}$

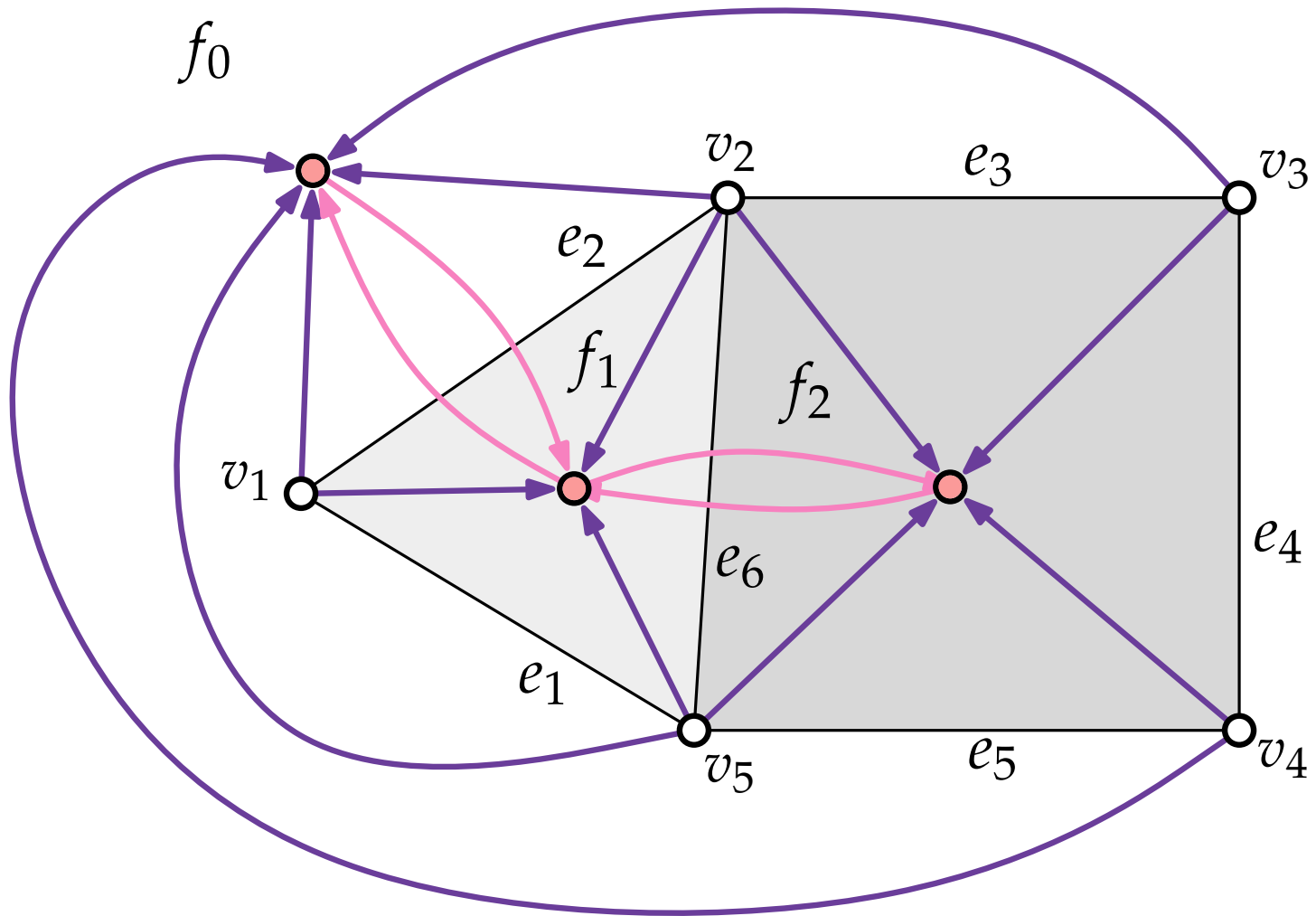
$V \times F \supseteq$   $\xrightarrow{1/4/0}$

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# Flow Network Example



# Flow Network Example



## Legend

$V$  ○

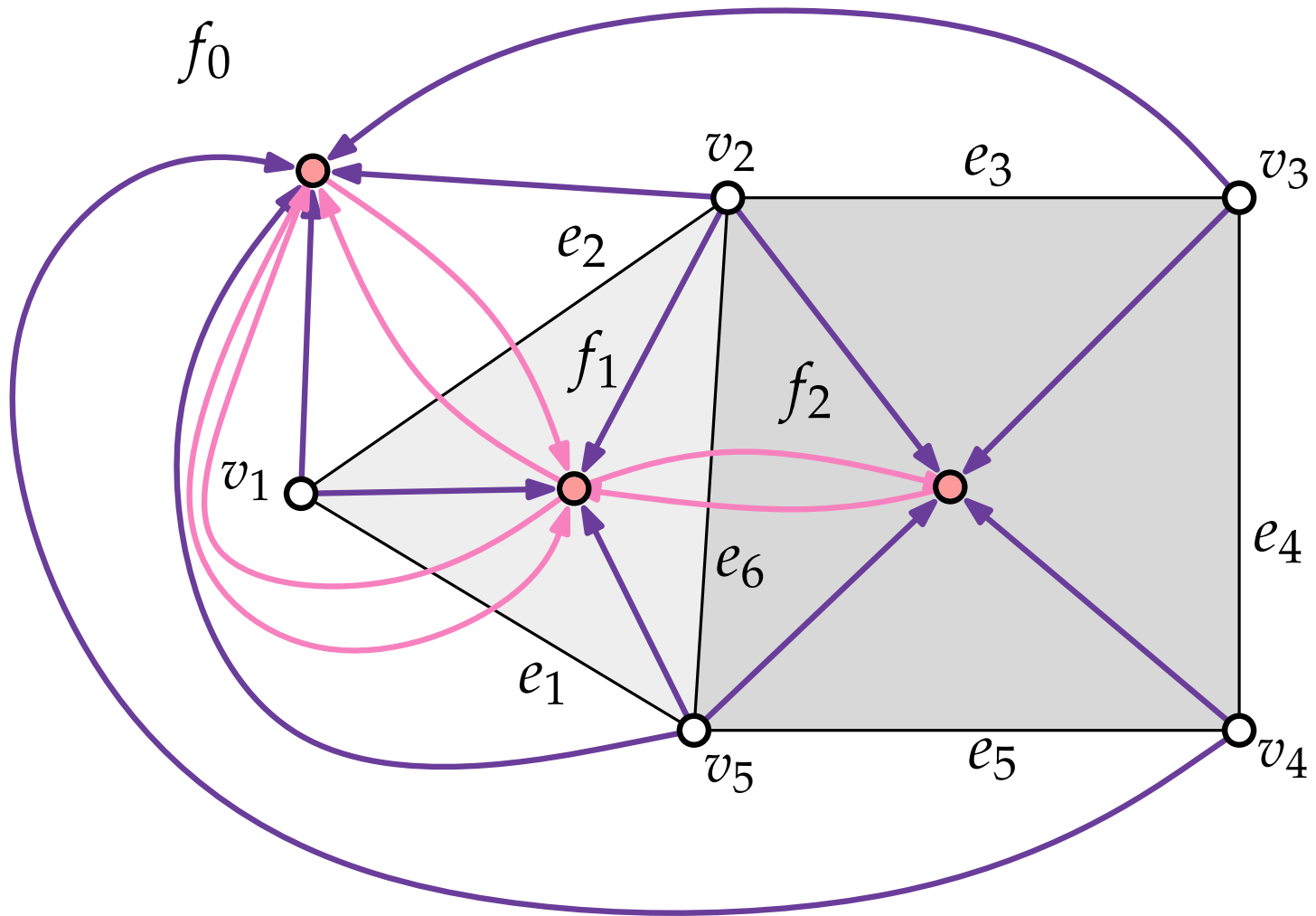
$F$  ●

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# Flow Network Example



## Legend

$V$  ○

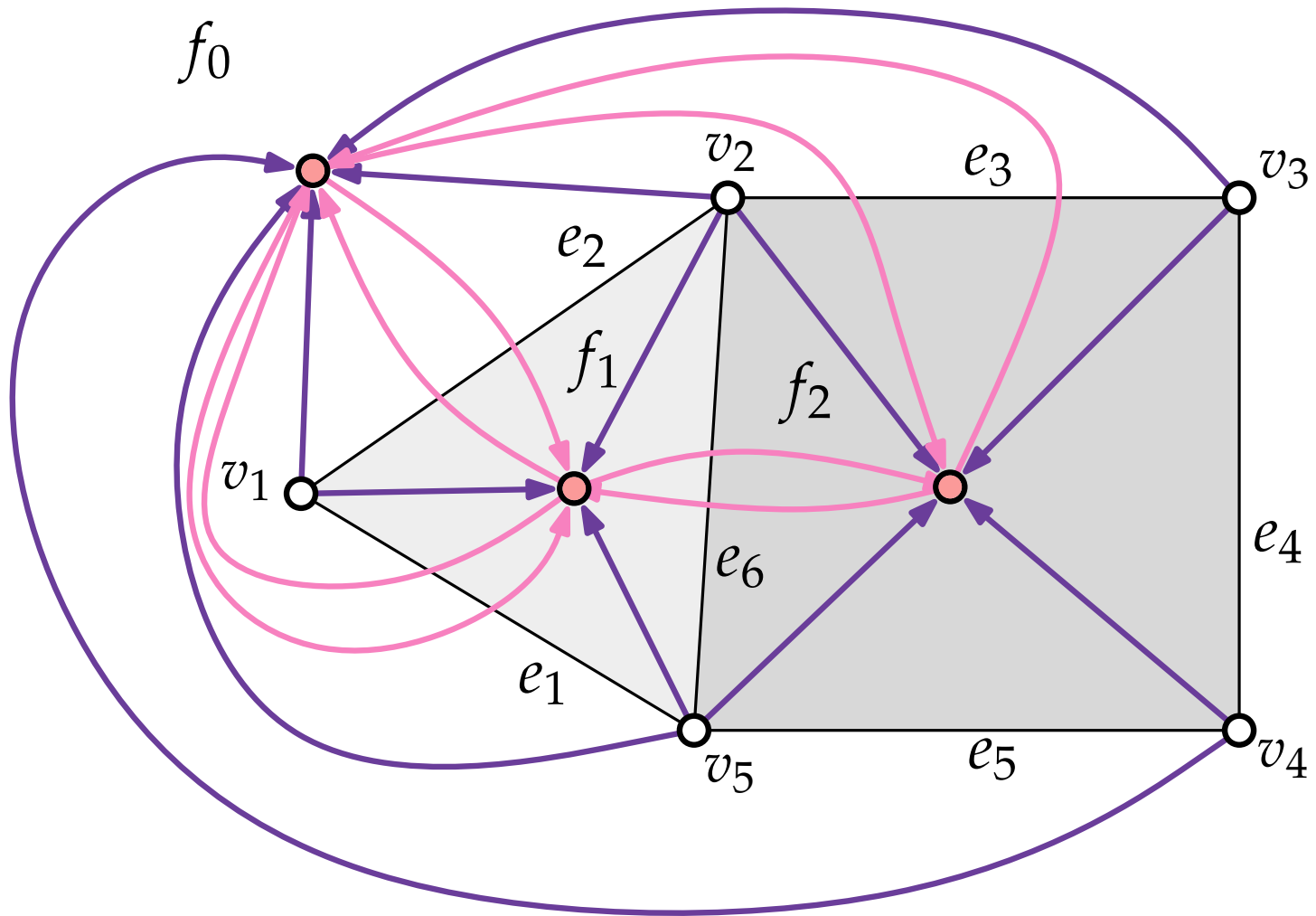
$F$  ●

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# Flow Network Example



## Legend

$V$  ○

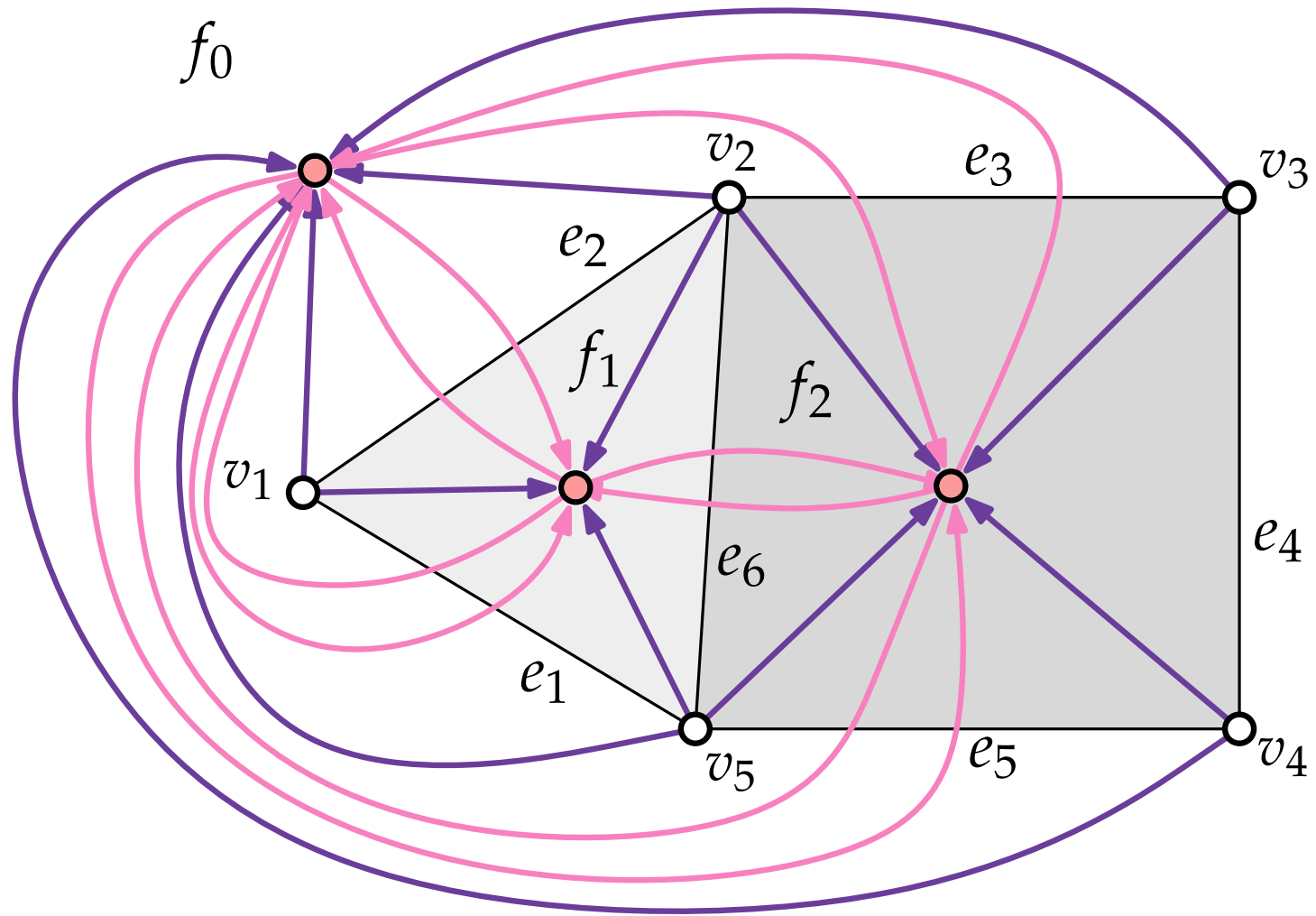
$F$  ●

$\ell/u/\text{cost}$

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# Flow Network Example



## Legend

$V$  ○

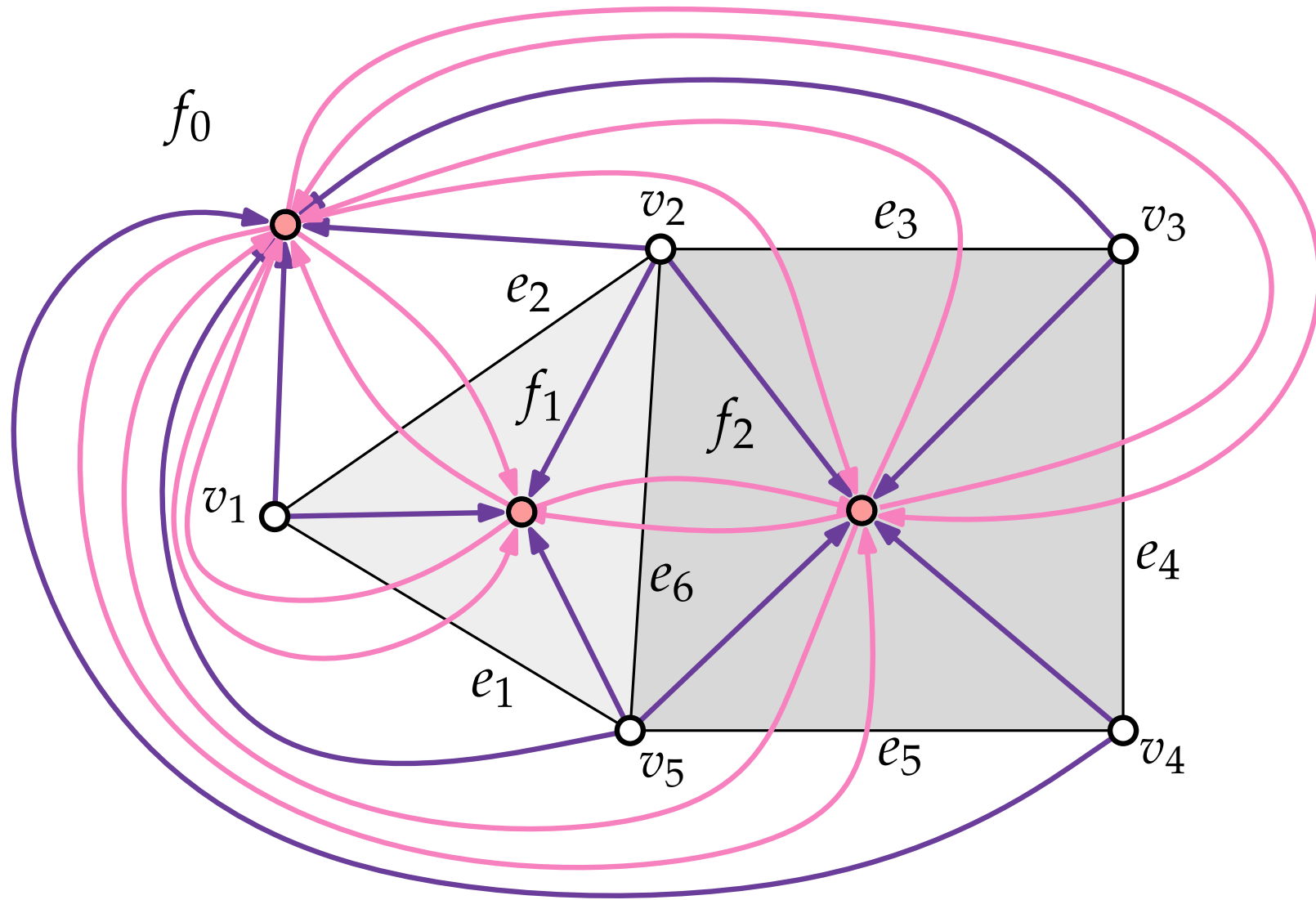
$F$  ●

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# Flow Network Example



## Legend

$V$  ○

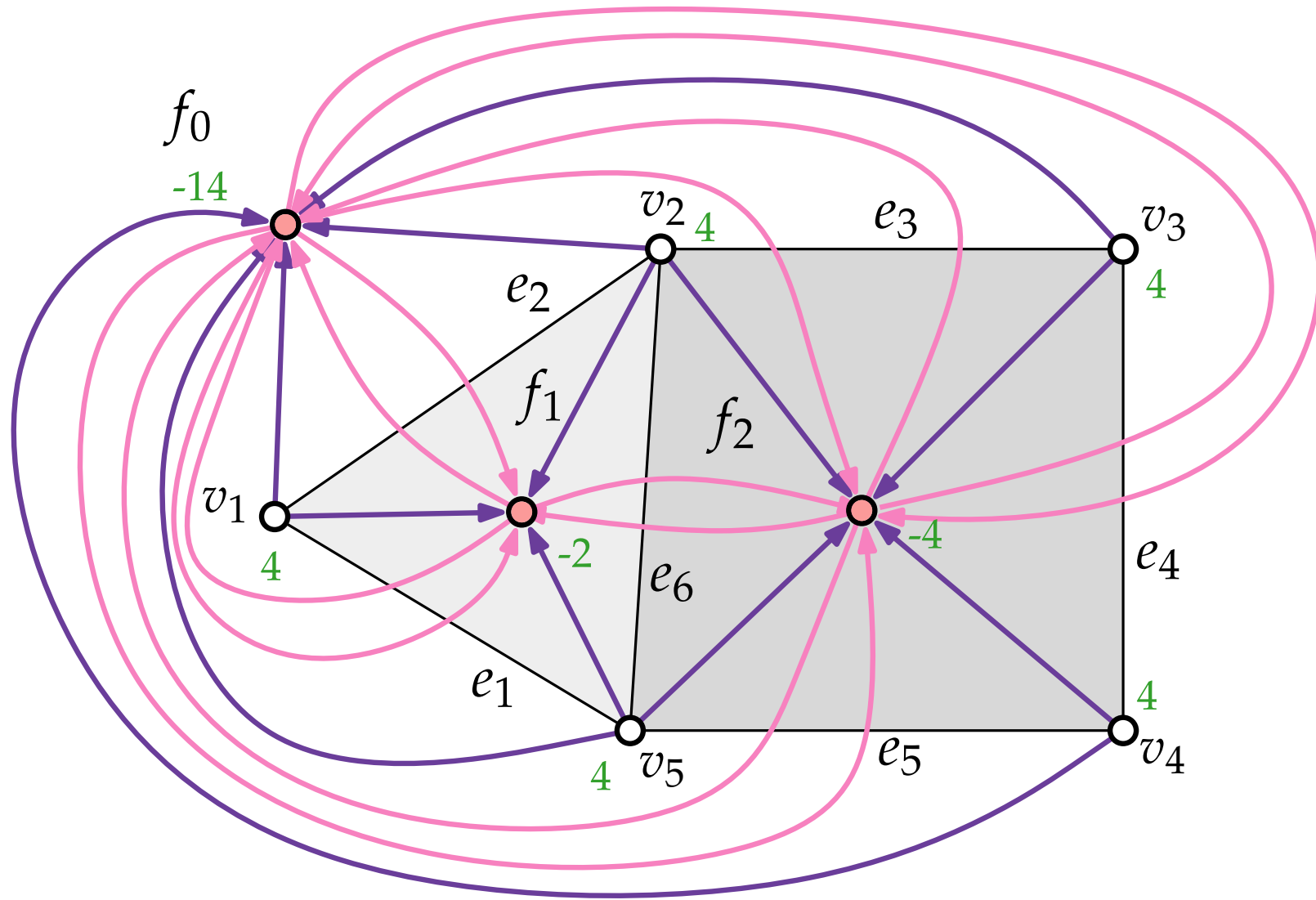
$F$  ●

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# Flow Network Example



## Legend

$V$  ○

$F$  ●

$\ell/u/\text{cost}$

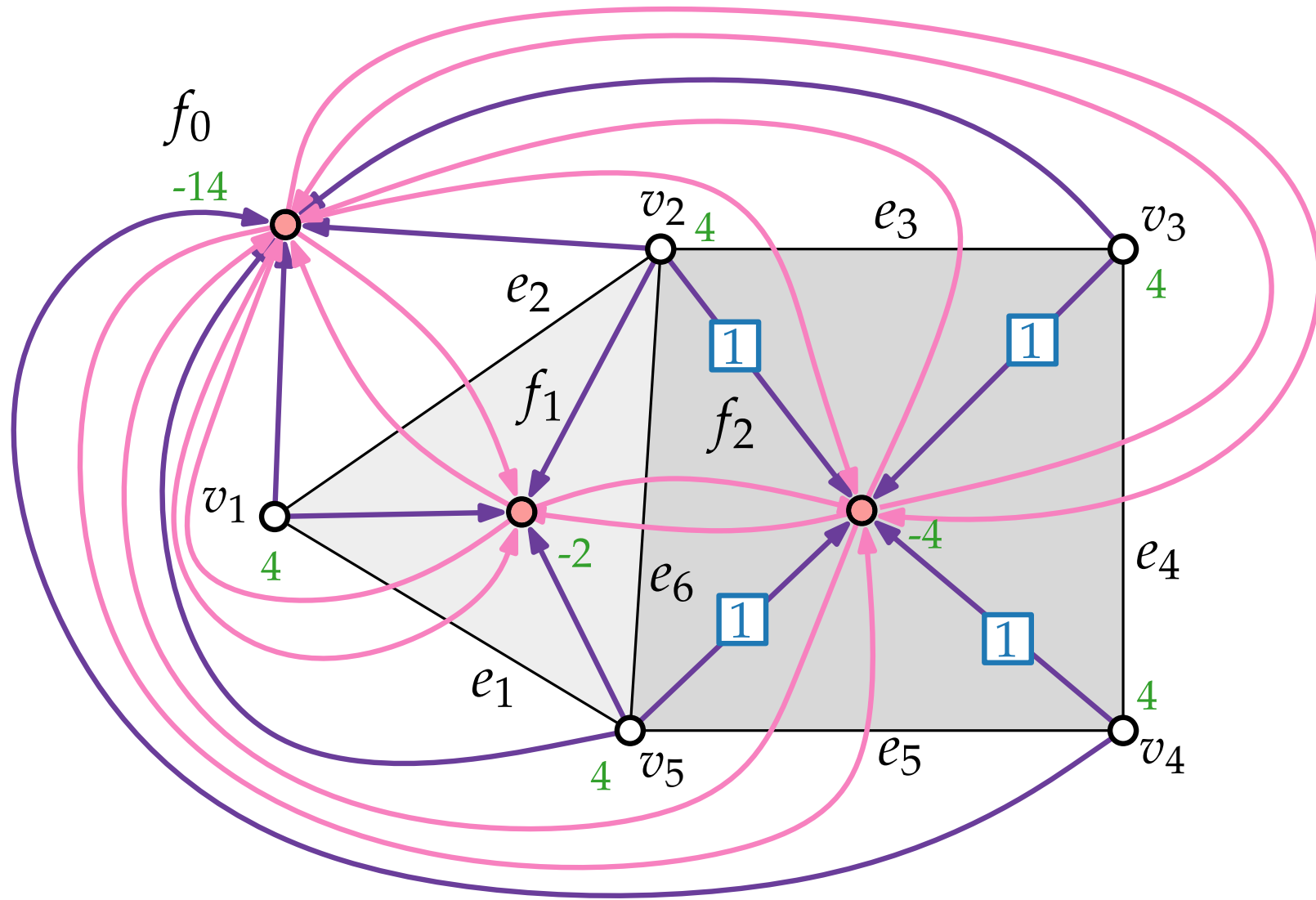
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$F \times F \supseteq$   $\xrightarrow{0/\infty/1}$

4 =  $b$ -value



# Flow Network Example



## Legend

$V$  ○

$F$  ●

$l/u/cost$

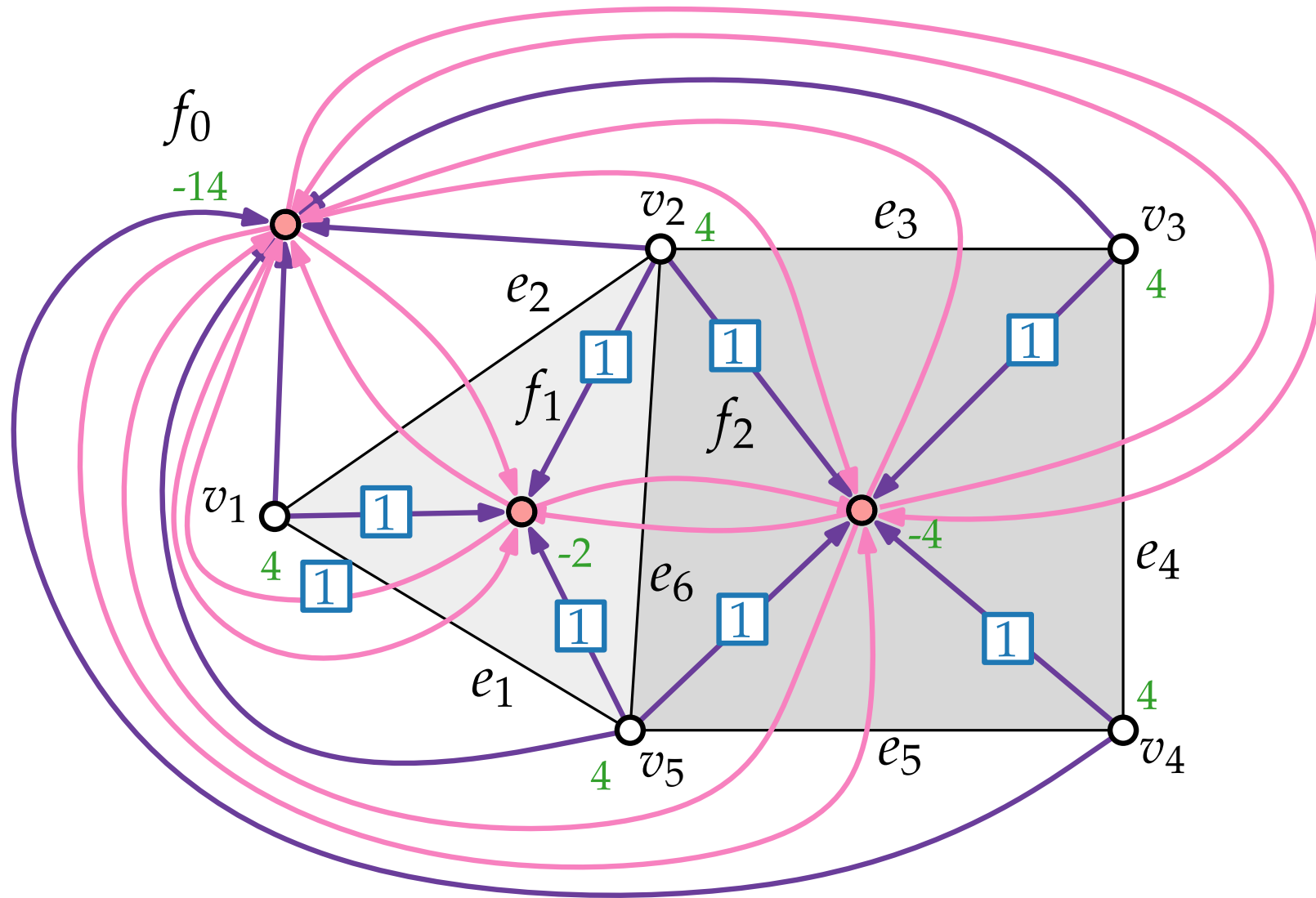
$V \times F \supseteq$   $\xrightarrow{1/4/0}$

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$4 = b\text{-value}$

$\boxed{3}$  flow

# Flow Network Example



## Legend

$V$  ○

$F$  ●

$\ell/u/\text{cost}$

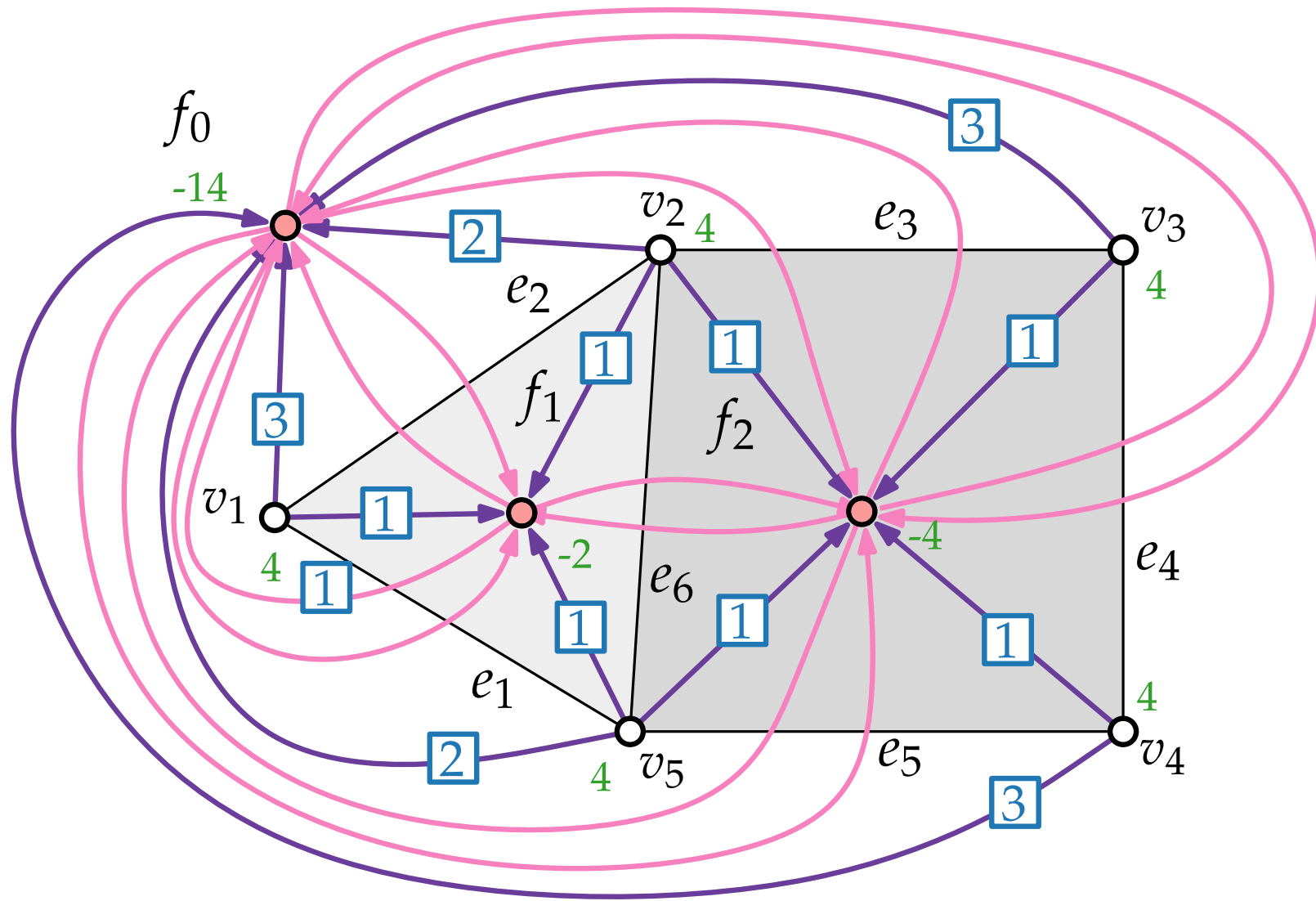
$V \times F \supseteq$   $\xrightarrow{1/4/0}$

$F \times F \supseteq$   $\xrightarrow{0/\infty/1}$

$4 = b\text{-value}$

$3$  flow

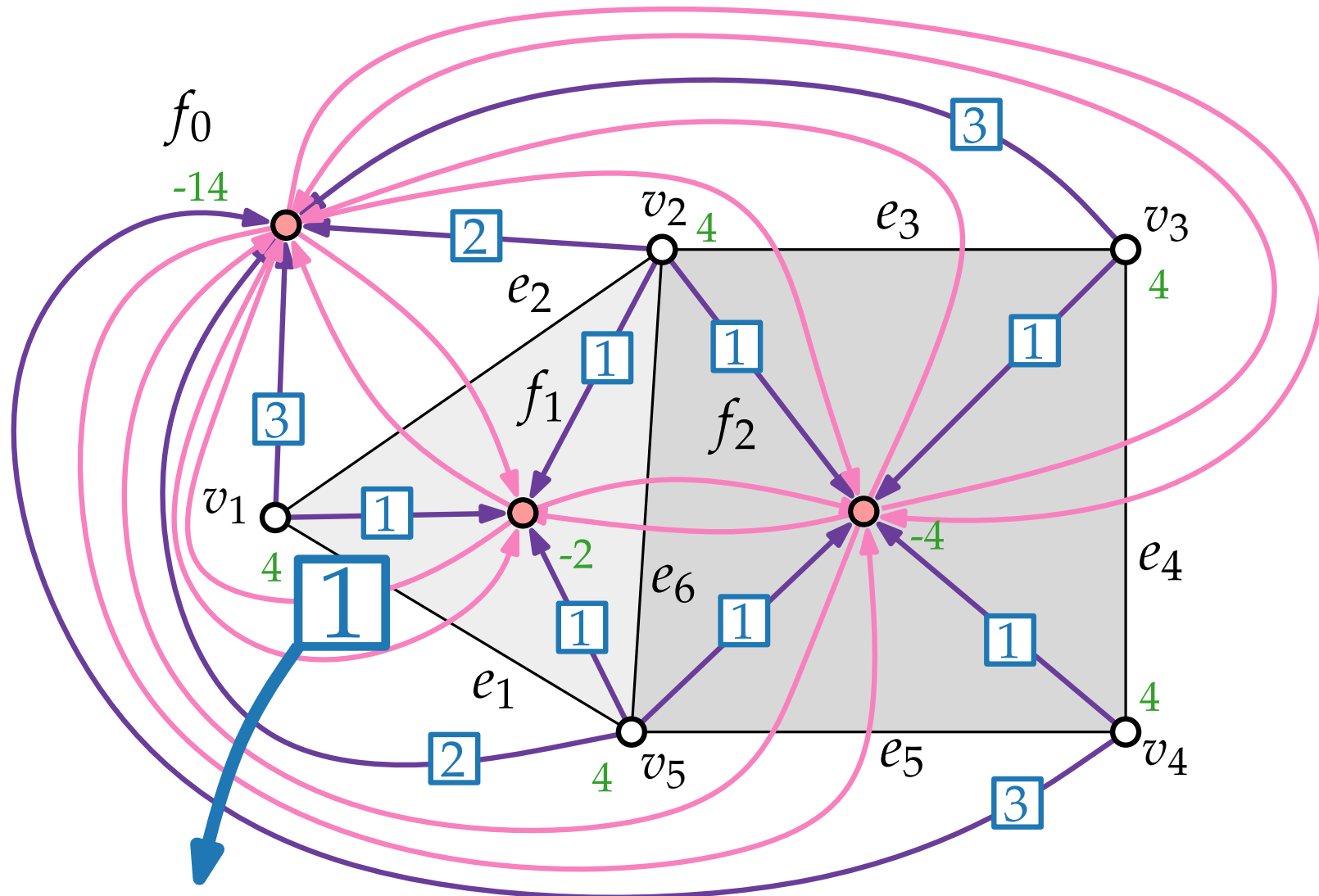
# Flow Network Example



## Legend

- $V$      $\circ$
- $F$      $\bullet$
- $V \times F \supseteq$      $\xrightarrow{1/4/0}$
- $F \times F \supseteq$      $\xrightarrow{0/\infty/1}$
- 4 = *b*-value
- 3 flow

# Flow Network Example



## Legend

$V$  ○

$F$  ●

$l/u/cost$

$V \times F \supseteq \xrightarrow{1/4/0}$

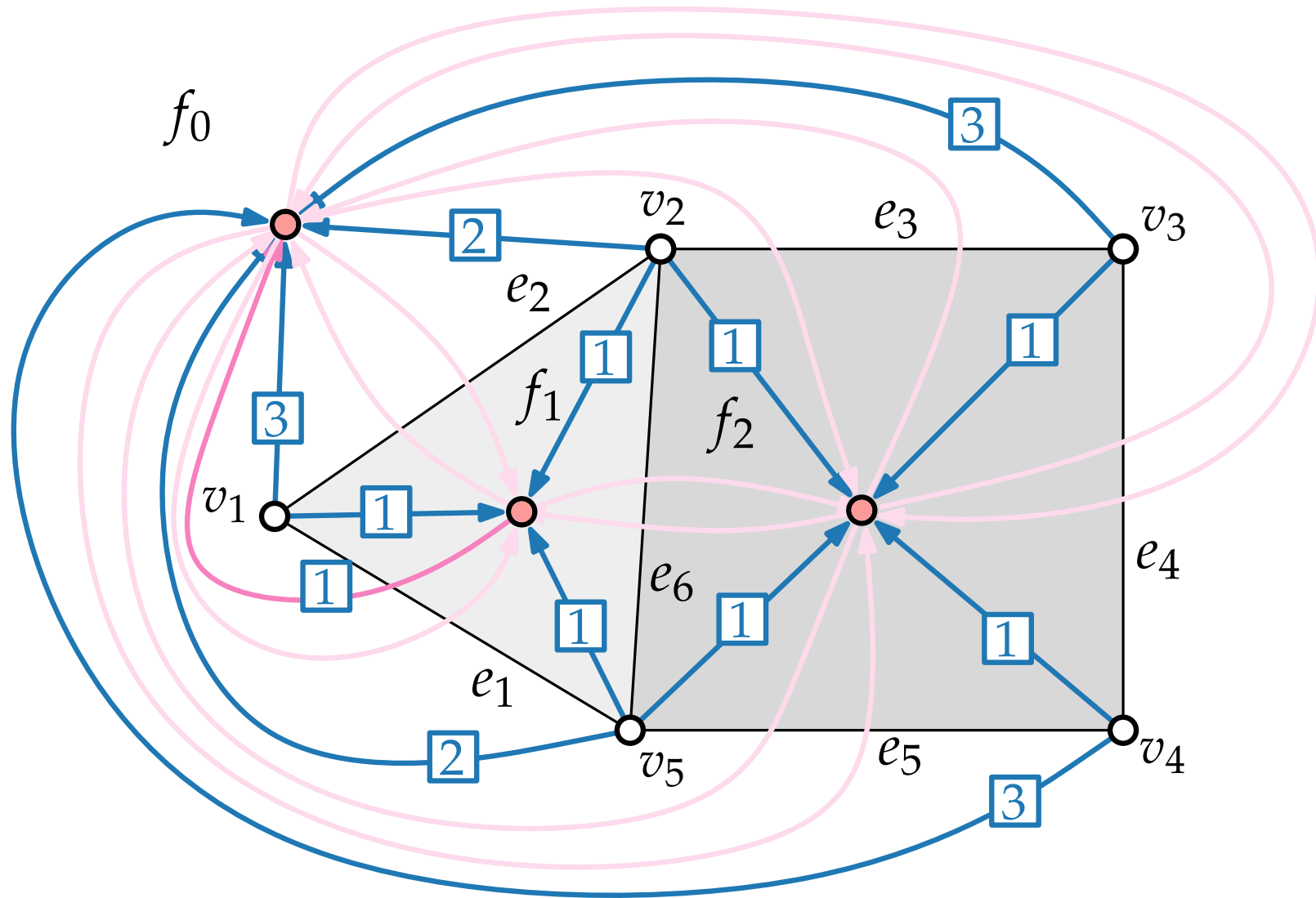
$F \times F \supseteq \xrightarrow{0/\infty/1}$

4 =  $b$ -value

3 flow

cost = 1  
one bend  
(outward)

# Flow Network Example



## Legend

$V$  ○

$F$  ●

$l/u/cost$

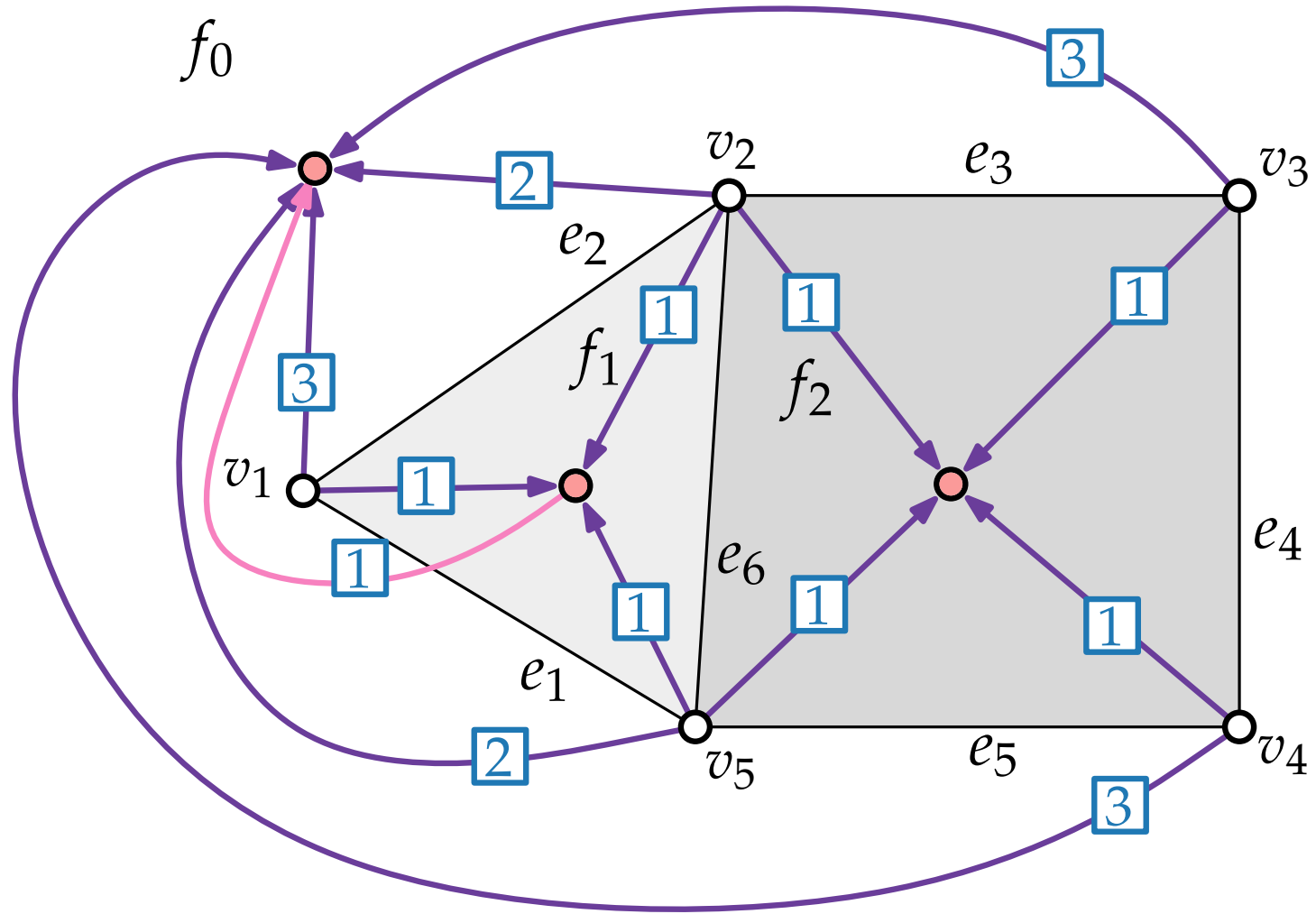
$V \times F \supseteq$   $\xrightarrow{1/4/0}$

$F \times F \supseteq$   $\xrightarrow{0/\infty/1}$

4 =  $b$ -value

3 flow

# Flow Network Example



## Legend

$V$  ○

$F$  ●

$l/u/cost$

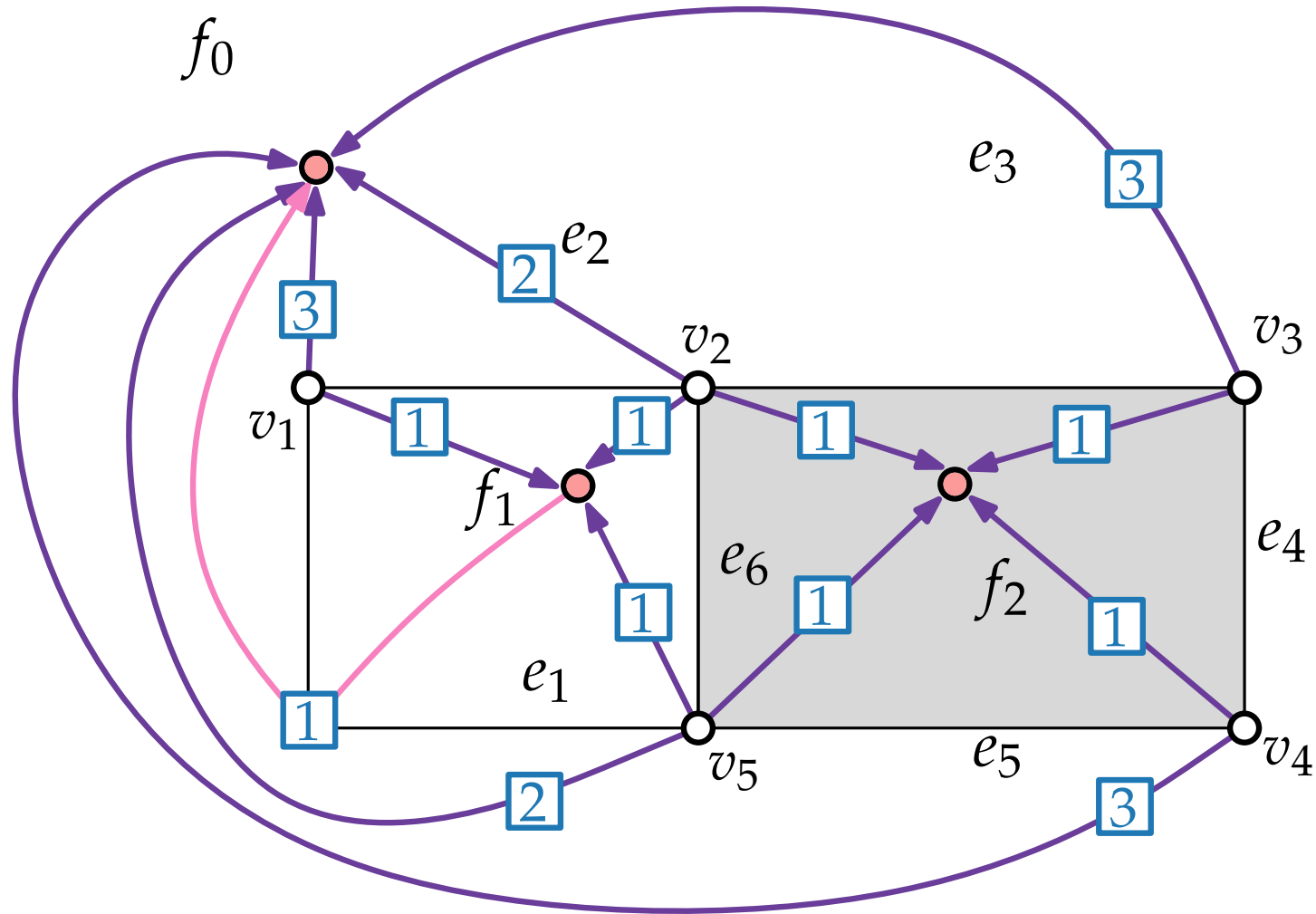
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3 flow

# Flow Network Example



## Legend

$V$  ○

$F$  ●

$l/u/cost$

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4 =  $b$ -value

3 flow

# Bend Minimization – Result

## Theorem.

[Tamassia '87]

A plane graph  $(G, F, f_0)$  has a valid orthogonal representation  $H(G)$  with  $k$  bends iff the flow network  $N(G)$  has a valid flow  $X$  with cost  $k$ .



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$\Leftarrow$  Given valid flow  $X$  in  $N(G)$  with cost  $k$ .

Construct  $\circ$

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- Transform from flow to orthogonal description.
- Show properties (H1)–(H4).

(H1)

(H2)

(H3)

(H4)

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(H4) For each **vertex**  $v$  the sum of incident angles is  $2\pi$ .

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(H1)  $H(G)$  matches  $F, f_0$

(H2)

(H3)

(H4)

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- Transform from flow to orthogonal description.
- Show properties (H1)–(H4).

(H1)  $H(G)$  matches  $F, f_0$



(H2)

(H3)

(H4) Total angle at each vertex =  $2\pi$



(H1)  $H(G)$  corresponds to  $F, f_0$ .

(H2) For each **edge**  $\{u, v\}$  shared by faces  $f$  and  $g$ , sequence  $\delta_1$  is reversed and inverted  $\delta_2$ .

(H3) For each **face**  $f$  it holds that:

$$\sum_{r \in H(f)} C(r) = \begin{cases} -4 & \text{if } f = f_0 \\ +4 & \text{otherwise.} \end{cases}$$

(H4) For each **vertex**  $v$  the sum of incident angles is  $2\pi$ .

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Exercise.



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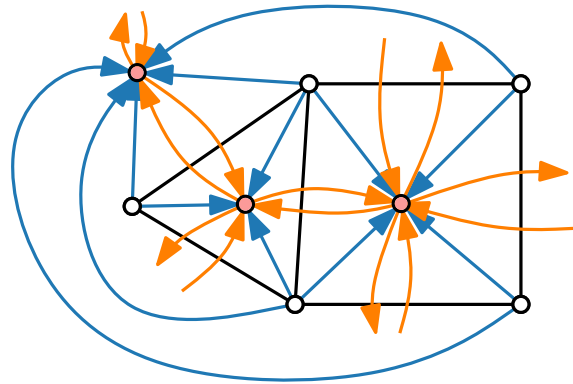
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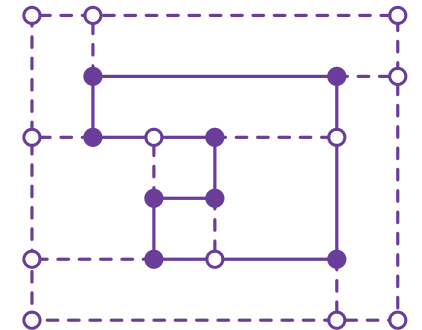
**Theorem.** [Garg & Tamassia 2001]

Bend Minimization without a given combinatorial embedding is an NP-hard problem.

# Visualization of Graphs

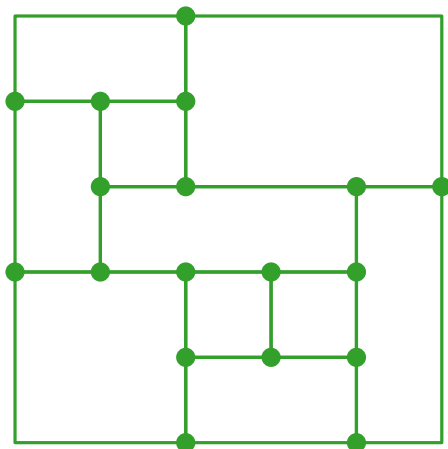


## Lecture 6: Orthogonal Layouts



## Part V: Area Minimization

Philipp Kindermann

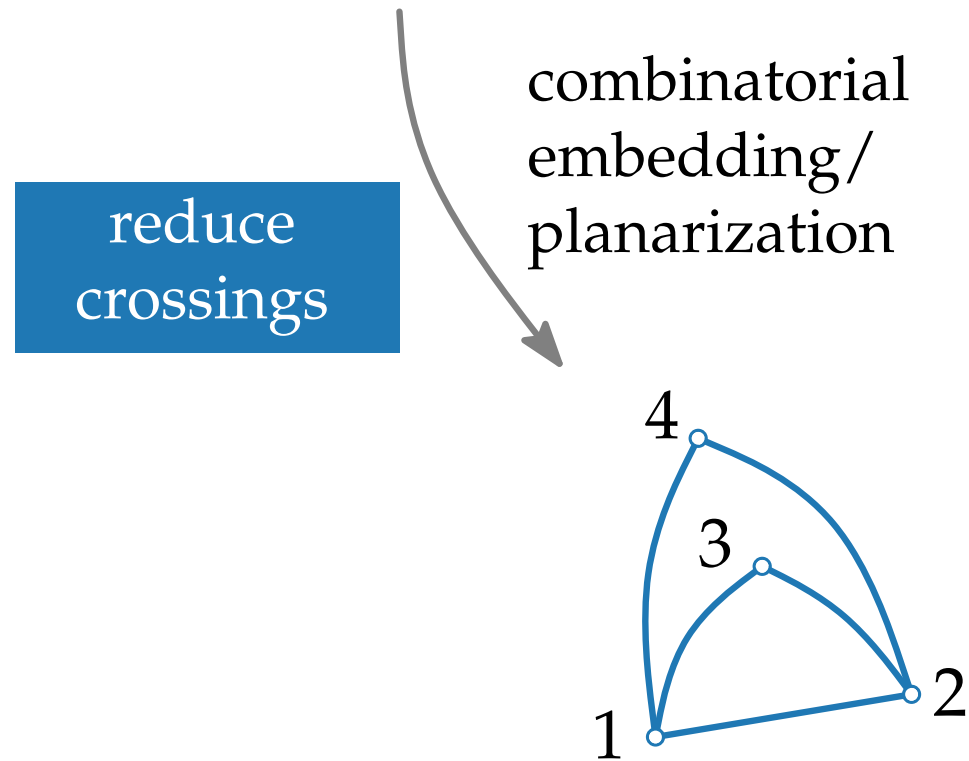


# Topology – Shape – Metrics

Three-step approach:

$$V = \{v_1, v_2, v_3, v_4\}$$

$$E = \{v_1v_2, v_1v_3, v_1v_4, v_2v_3, v_2v_4\}$$



TOPOLOGY

—

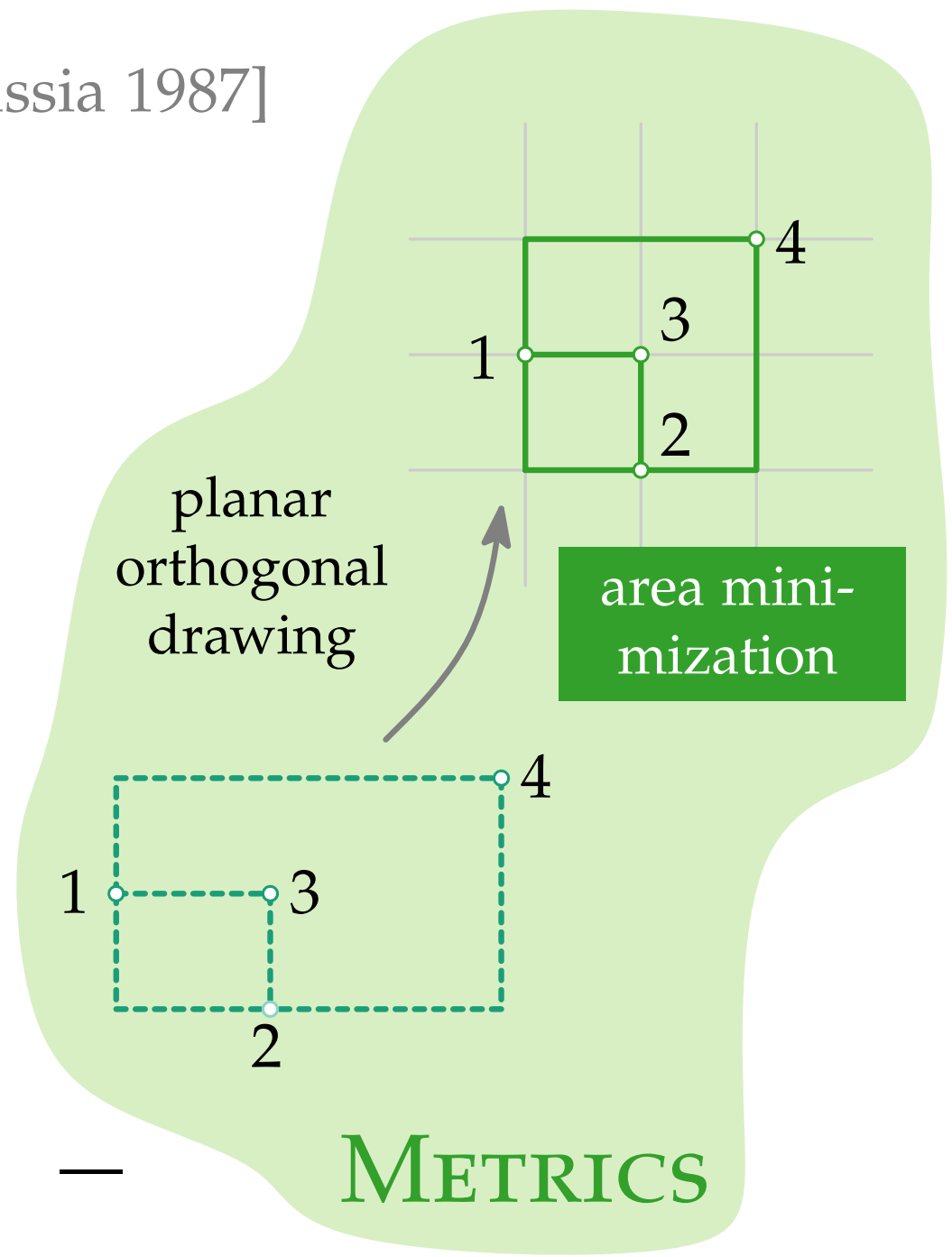
bend minimization

orthogonal representation

SHAPE

—

[Tamassia 1987]



METRICS

# Compaction

Compaction problem.

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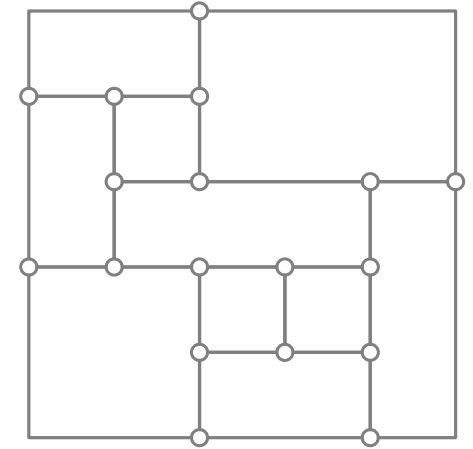
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## Idea.

- Formulate flow network for horizontal/vertical compaction

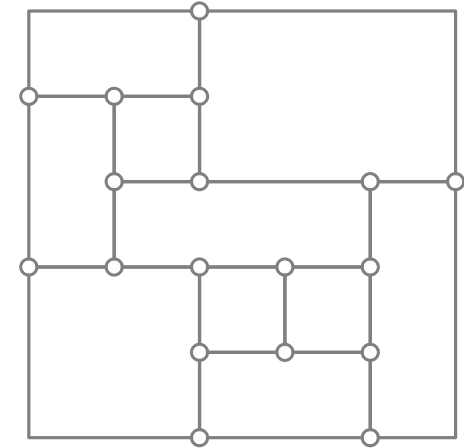
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## Definition.

Flow Network  $N_{\text{hor}} = ((W_{\text{hor}}, E_{\text{hor}}); b; \ell; u; \text{cost})$

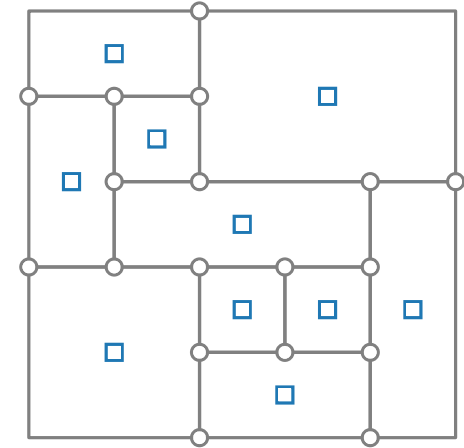


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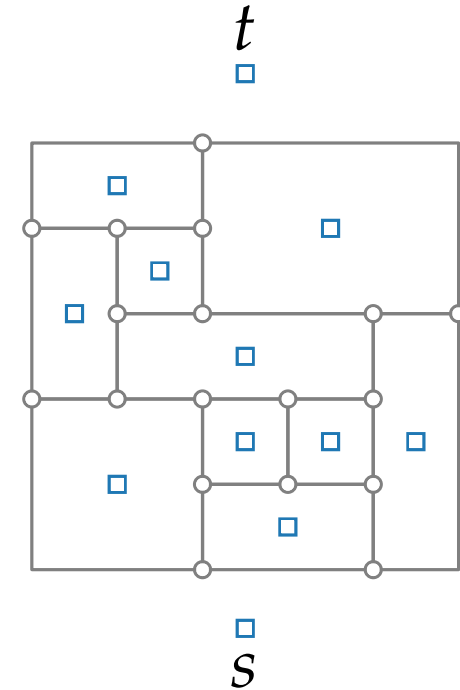


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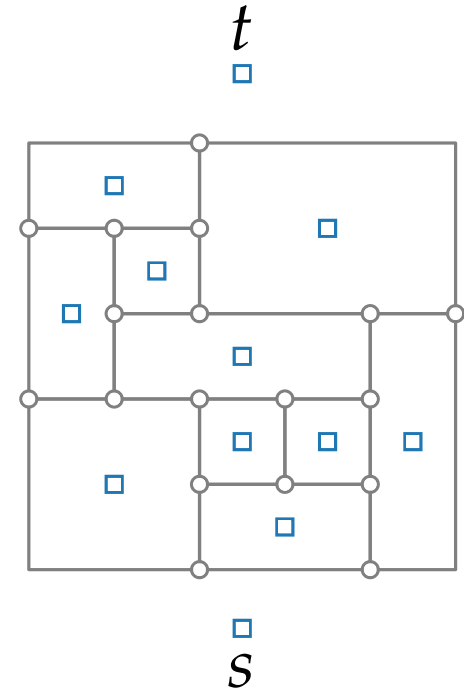


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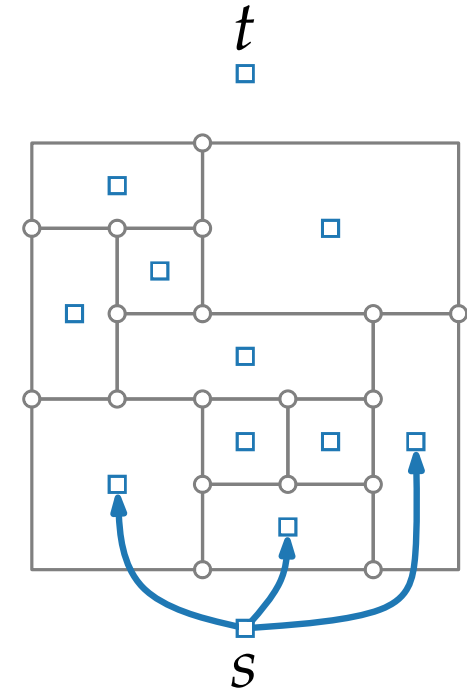


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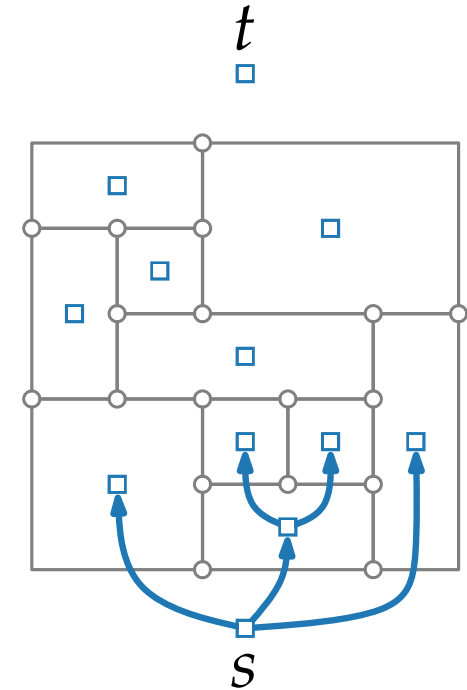


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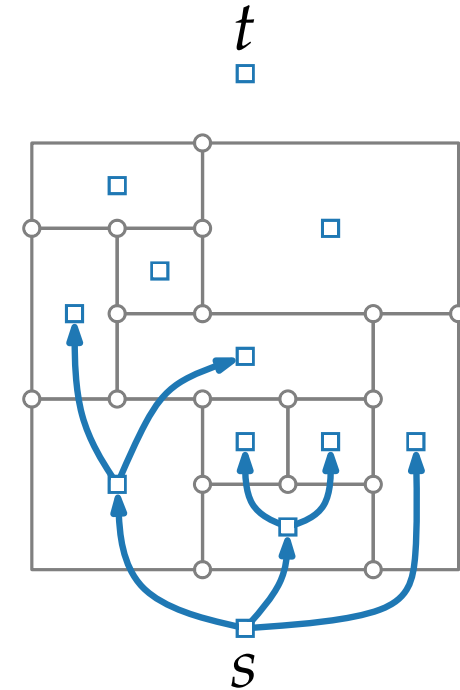


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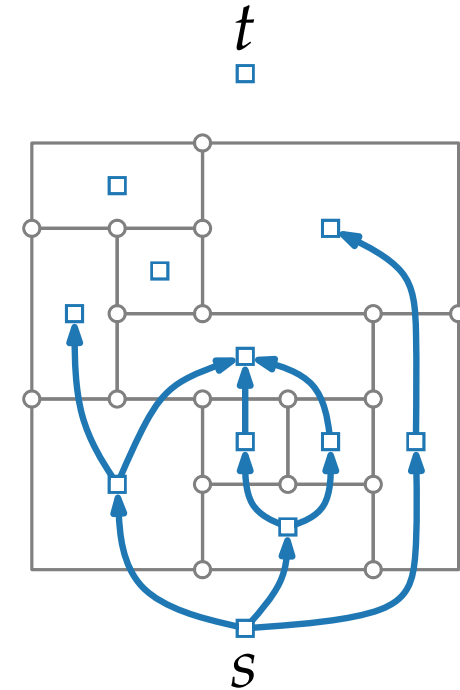


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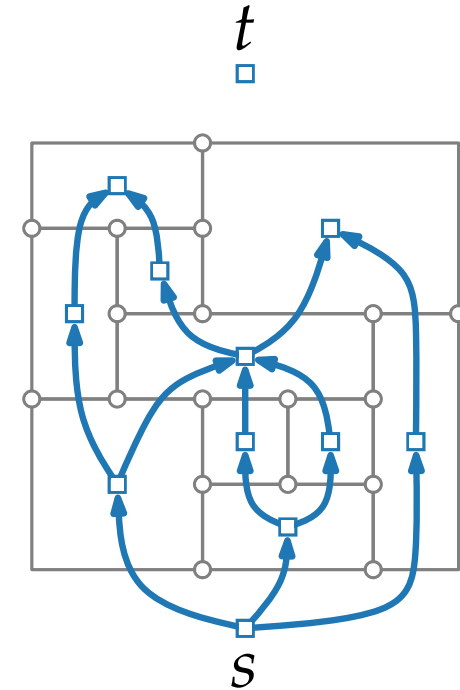


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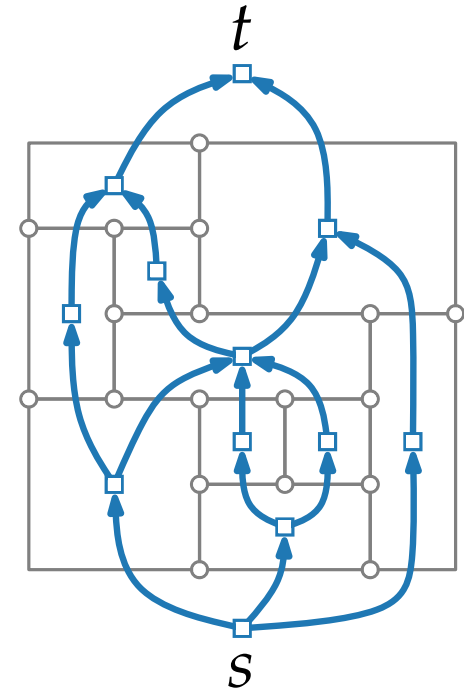


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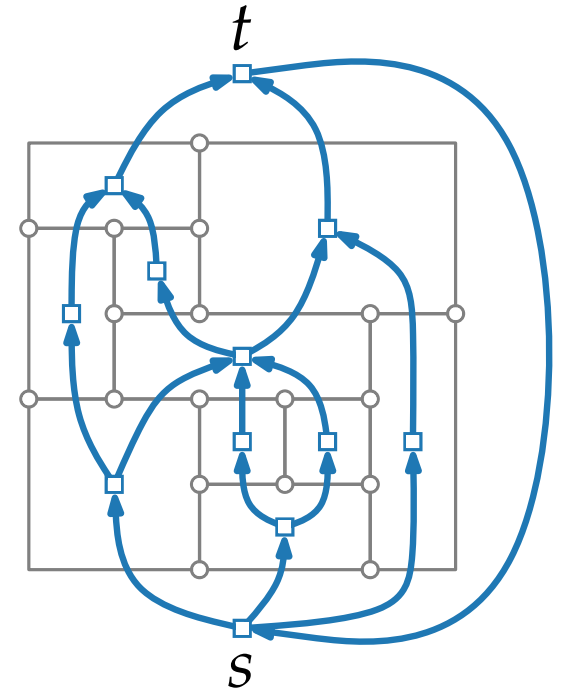


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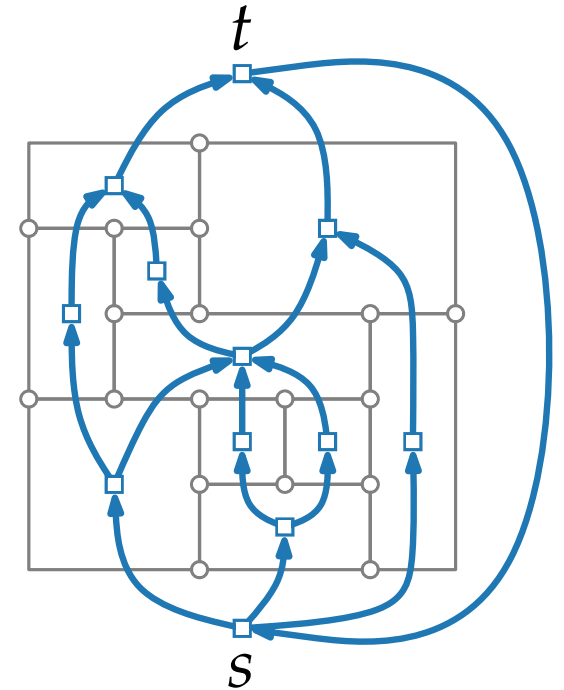


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- $\ell(a) = 1 \quad \forall a \in E_{\text{hor}}$



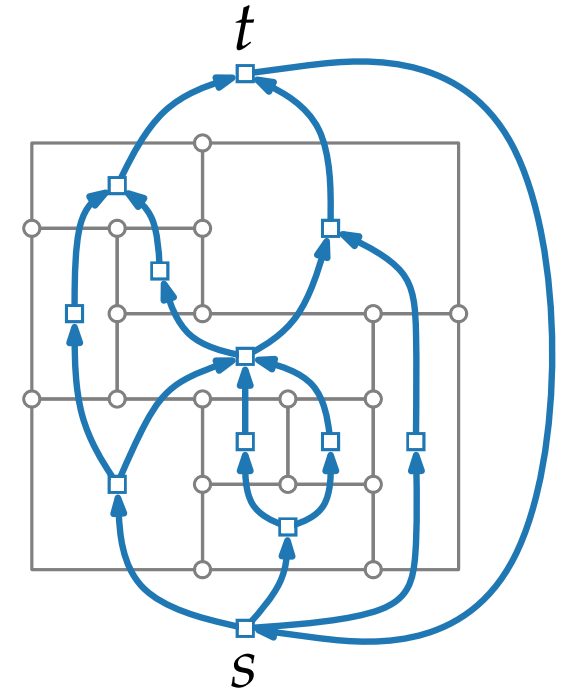


# Flow Network for Edge Length Assignment

## Definition.

Flow Network  $N_{\text{hor}} = ((W_{\text{hor}}, E_{\text{hor}}); b; \ell; u; \text{cost})$

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- $E_{\text{hor}} = \{(f, g) \mid f, g \text{ share a horizontal segment and } f \text{ lies below } g\} \cup \{(t, s)\}$
- $\ell(a) = 1 \quad \forall a \in E_{\text{hor}}$
- $u(a) = \infty \quad \forall a \in E_{\text{hor}}$

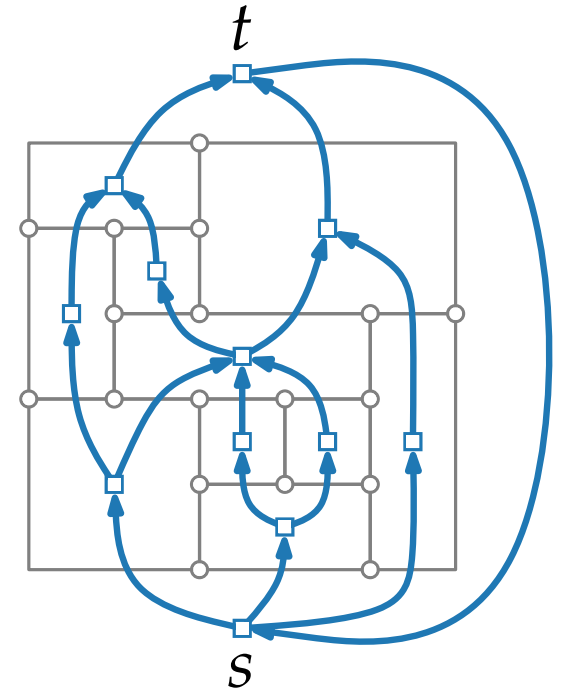


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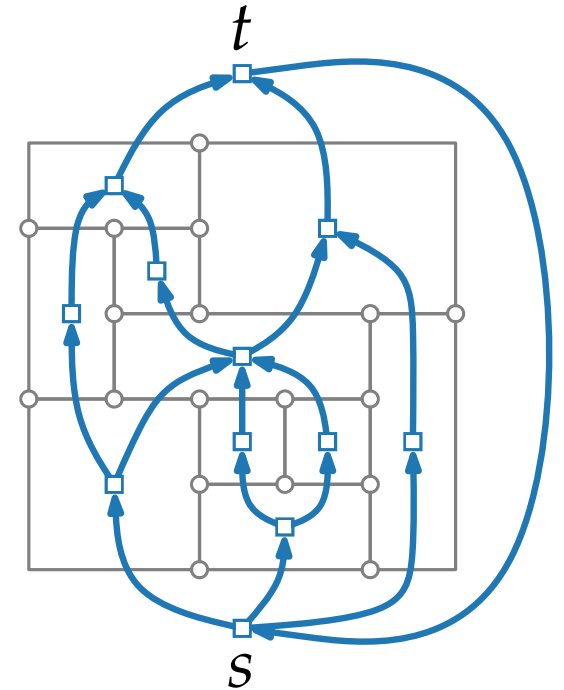


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- $b(f) = 0 \quad \forall f \in W_{\text{hor}}$

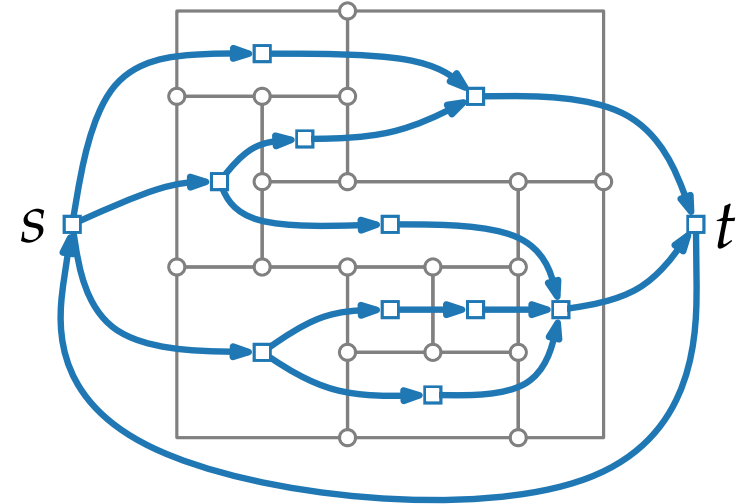


# Flow Network for Edge Length Assignment

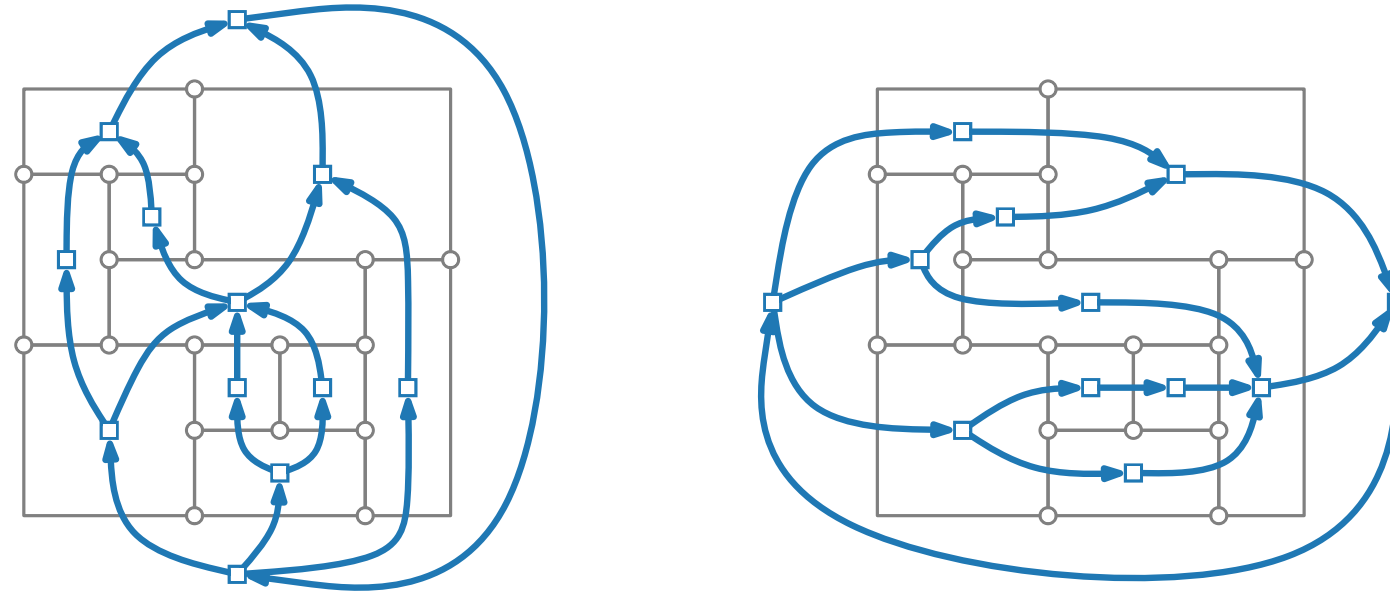
## Definition.

Flow Network  $N_{\text{ver}} = ((W_{\text{ver}}, E_{\text{ver}}); b; \ell; u; \text{cost})$

- $W_{\text{ver}} = F \setminus \{f_0\} \cup \{s, t\}$     □
- $E_{\text{ver}} = \{(f, g) \mid f, g \text{ share a } \textit{vertical} \text{ segment and } f \text{ lies to the } \textit{left} \text{ of } g\} \cup \{(t, s)\}$
- $\ell(a) = 1 \quad \forall a \in E_{\text{ver}}$
- $u(a) = \infty \quad \forall a \in E_{\text{ver}}$
- $\text{cost}(a) = 1 \quad \forall a \in E_{\text{ver}}$
- $b(f) = 0 \quad \forall f \in W_{\text{ver}}$



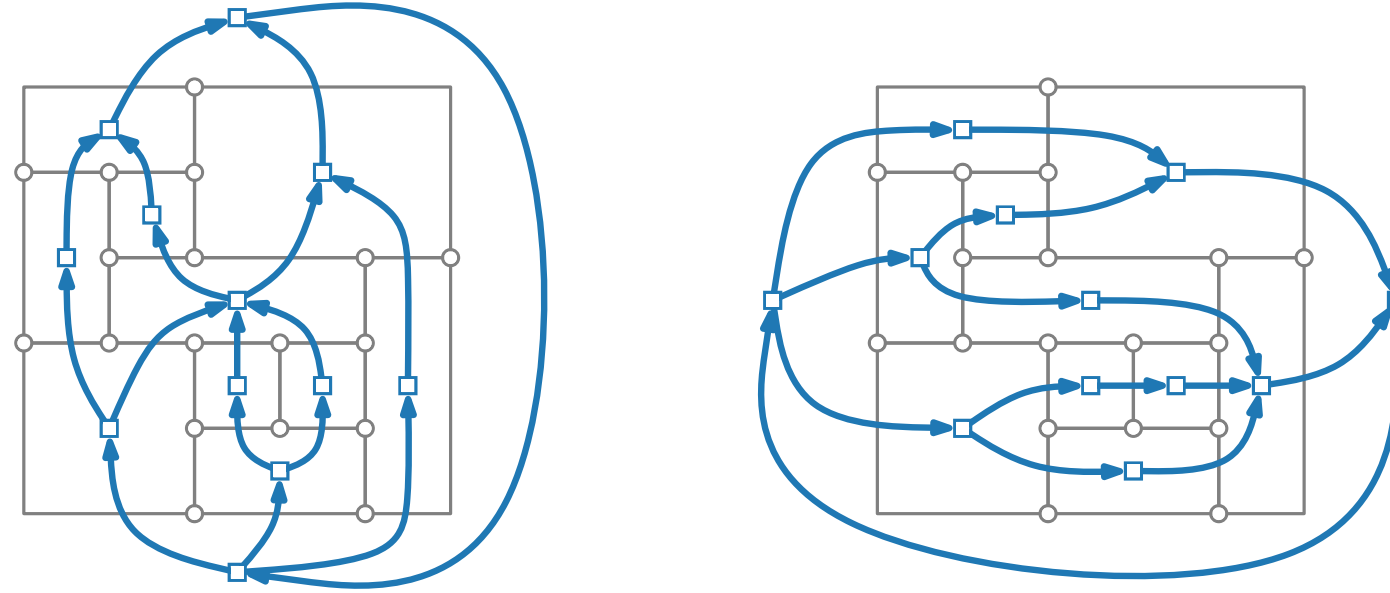
# Compaction – Result



## Theorem.

Valid min-cost-flows for  $N_{\text{hor}}$  and  $N_{\text{ver}}$  exists iff corresponding edge lengths induce orthogonal drawing.

# Compaction – Result

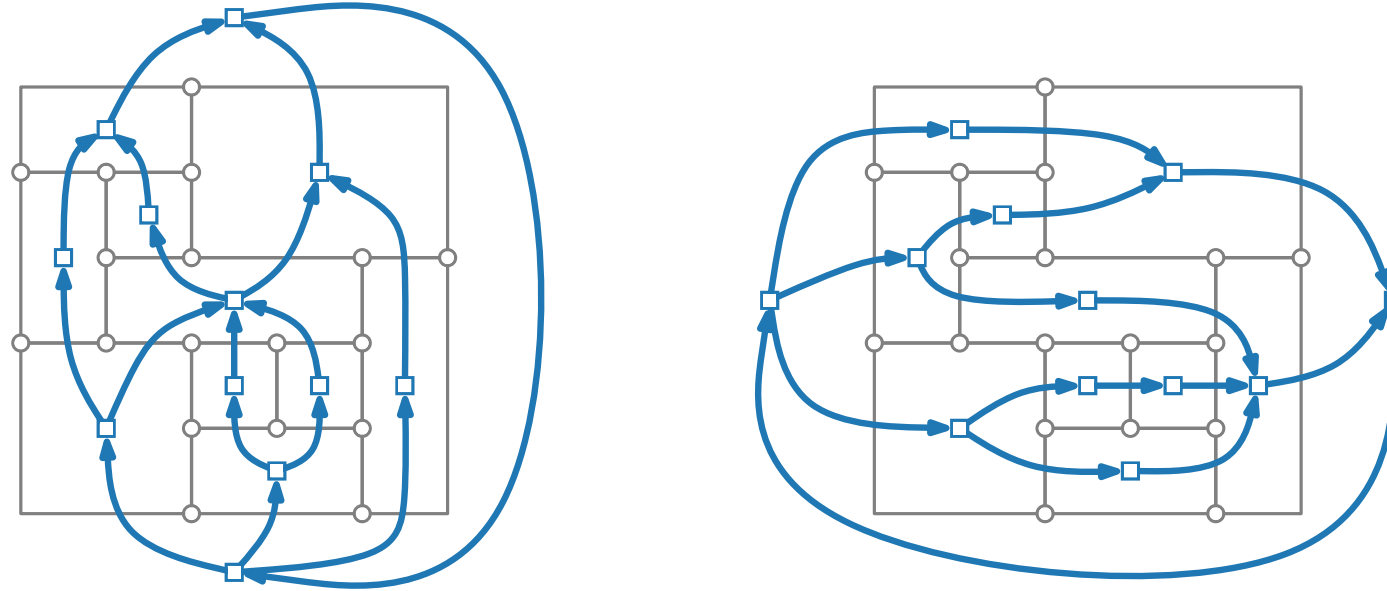


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What values of the drawing represent the following?

# Compaction – Result



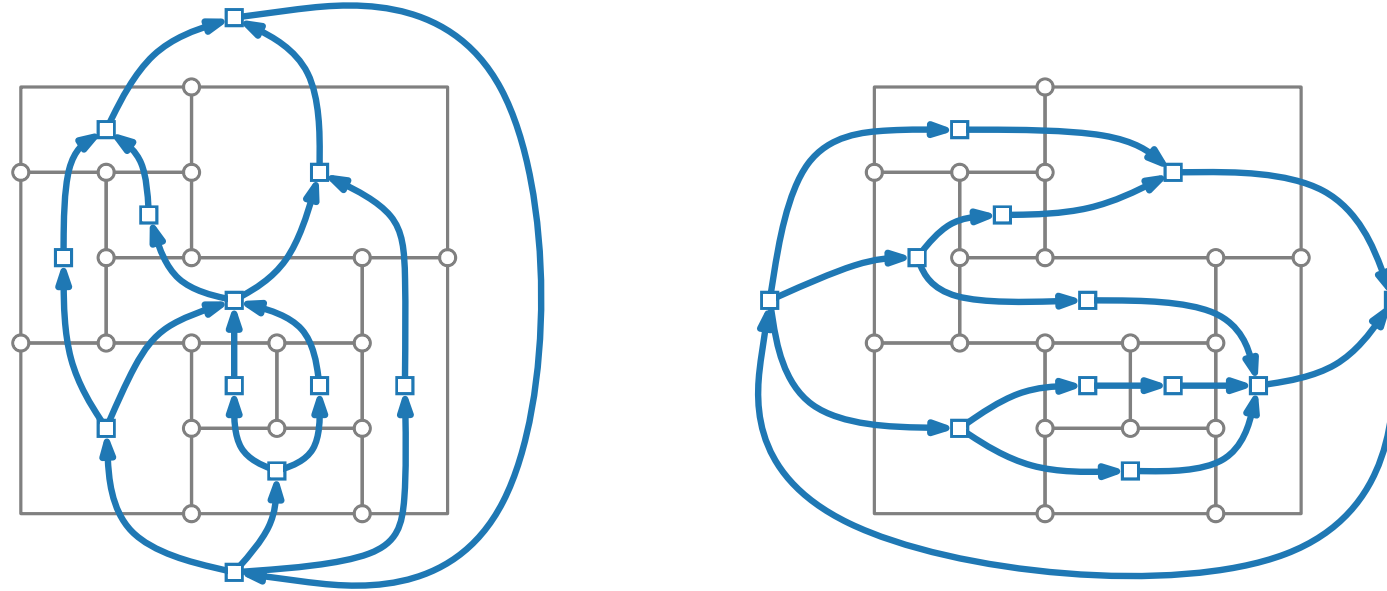
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Valid min-cost-flows for  $N_{\text{hor}}$  and  $N_{\text{ver}}$  exists iff corresponding edge lengths induce orthogonal drawing.

What values of the drawing represent the following?

- $|X_{\text{hor}}(t, s)|$  and  $|X_{\text{ver}}(t, s)|$ ?

# Compaction – Result



## Theorem.

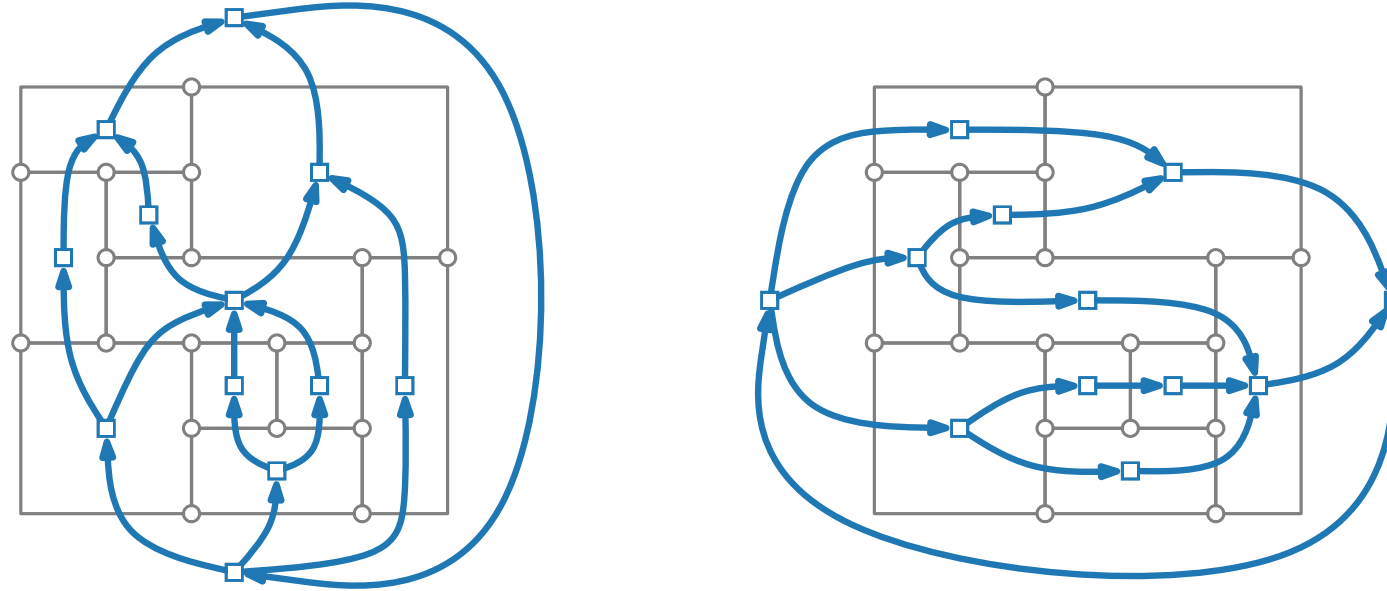
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- $|X_{\text{hor}}(t, s)|$  and  $|X_{\text{ver}}(t, s)|$       width and height of drawing



# Compaction – Result



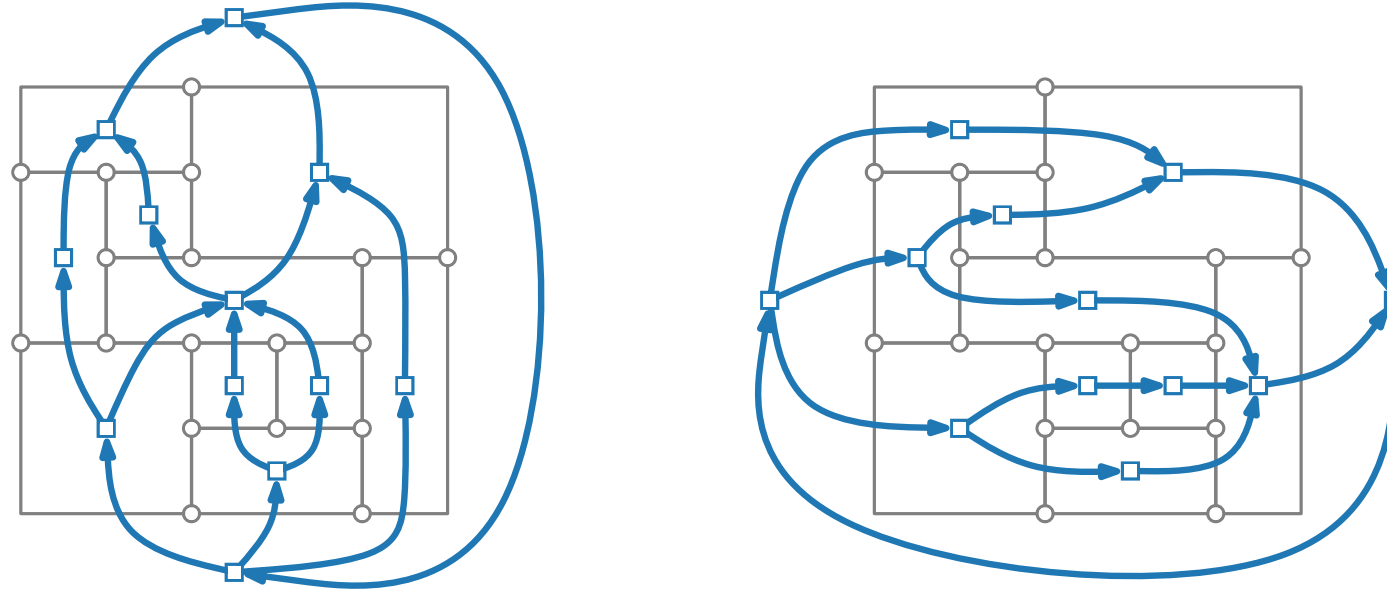
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- $|X_{\text{hor}}(t, s)|$  and  $|X_{\text{ver}}(t, s)|$       width and height of drawing
- $\sum_{e \in E_{\text{hor}}} X_{\text{hor}}(e) + \sum_{e \in E_{\text{ver}}} X_{\text{ver}}(e)$

# Compaction – Result



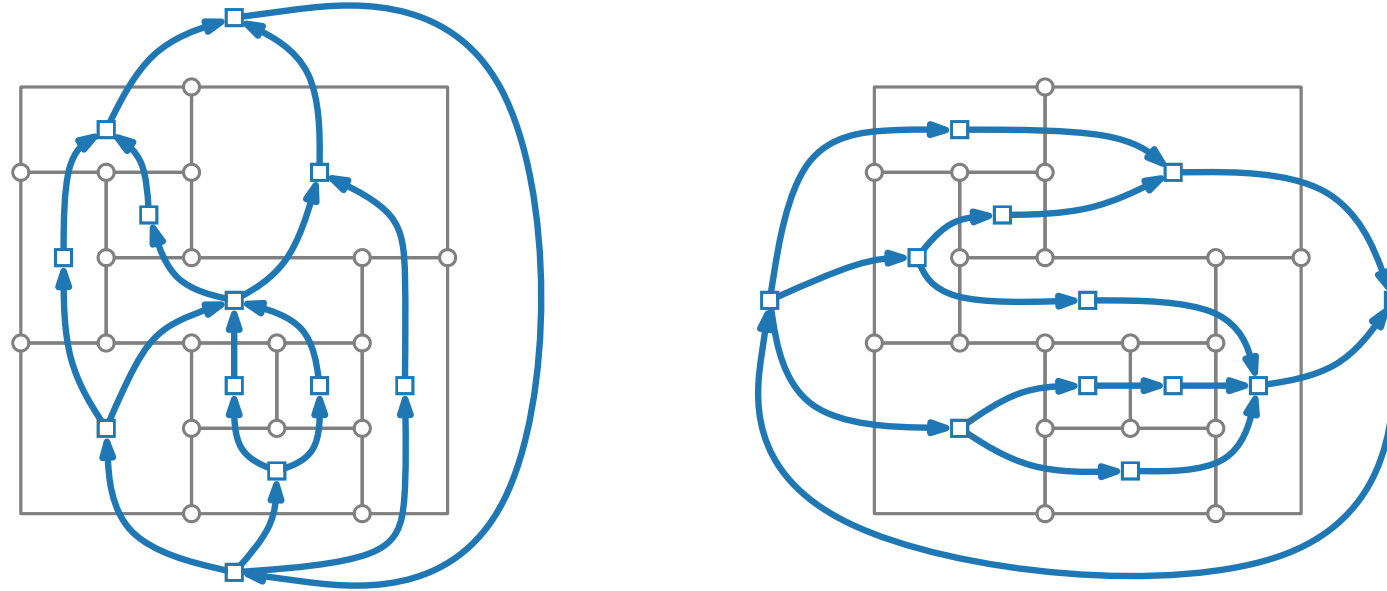
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- $\sum_{e \in E_{\text{hor}}} X_{\text{hor}}(e) + \sum_{e \in E_{\text{ver}}} X_{\text{ver}}(e)$       total edge length

# Compaction – Result



What if not all faces rectangular?

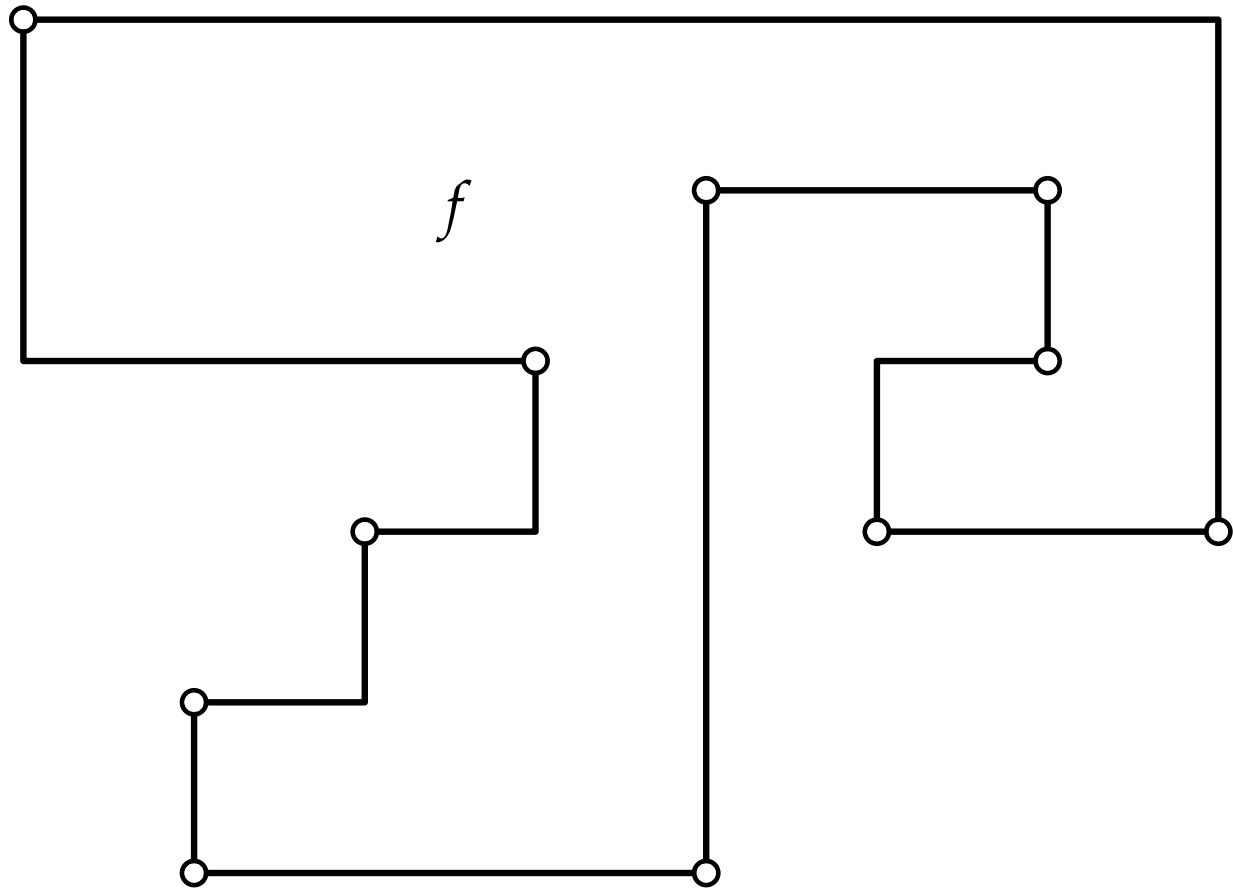
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Valid min-cost-flows for  $N_{\text{hor}}$  and  $N_{\text{ver}}$  exists iff corresponding edge lengths induce orthogonal drawing.

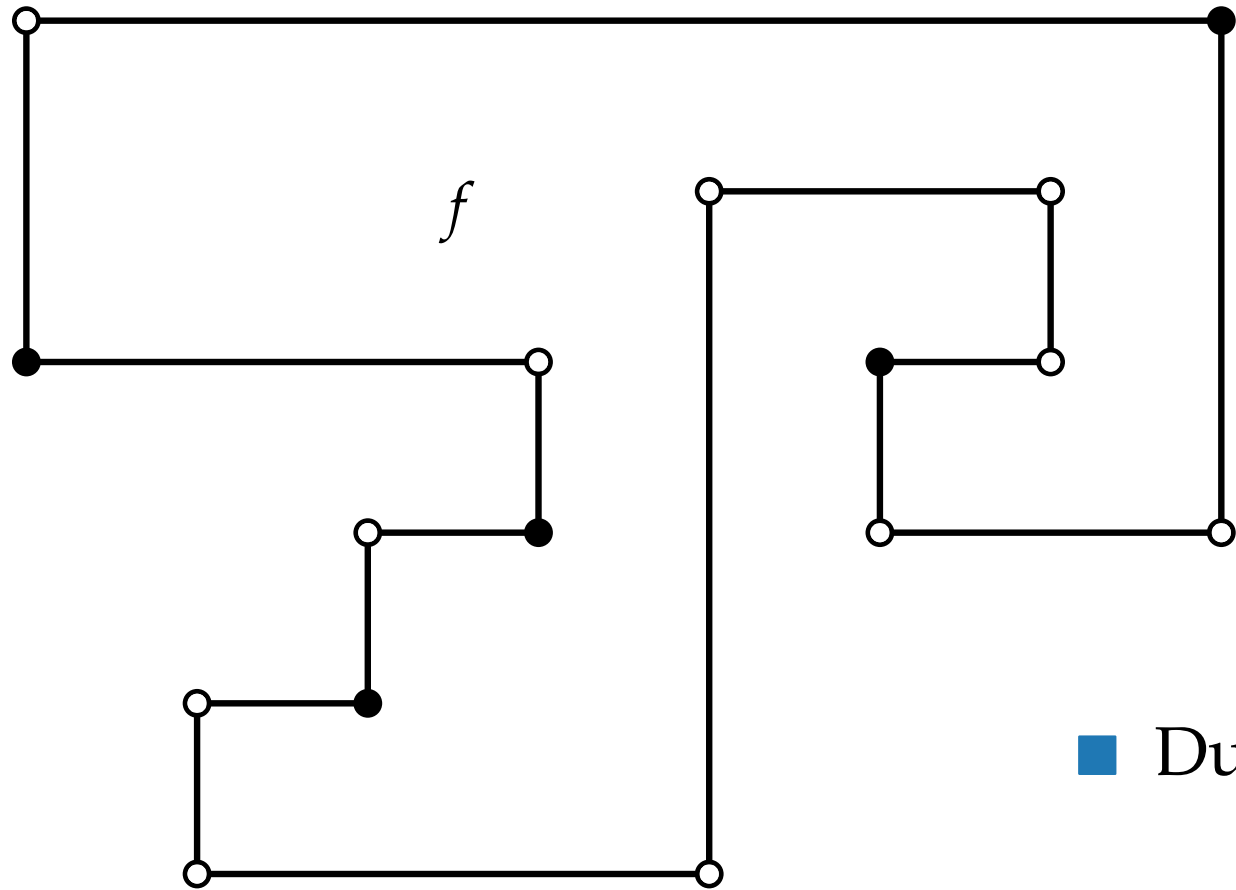
What values of the drawing represent the following?

- $|X_{\text{hor}}(t, s)|$  and  $|X_{\text{ver}}(t, s)|$       width and height of drawing
- $\sum_{e \in E_{\text{hor}}} X_{\text{hor}}(e) + \sum_{e \in E_{\text{ver}}} X_{\text{ver}}(e)$       total edge length

# Refinement of $(G, H)$ – Inner Face

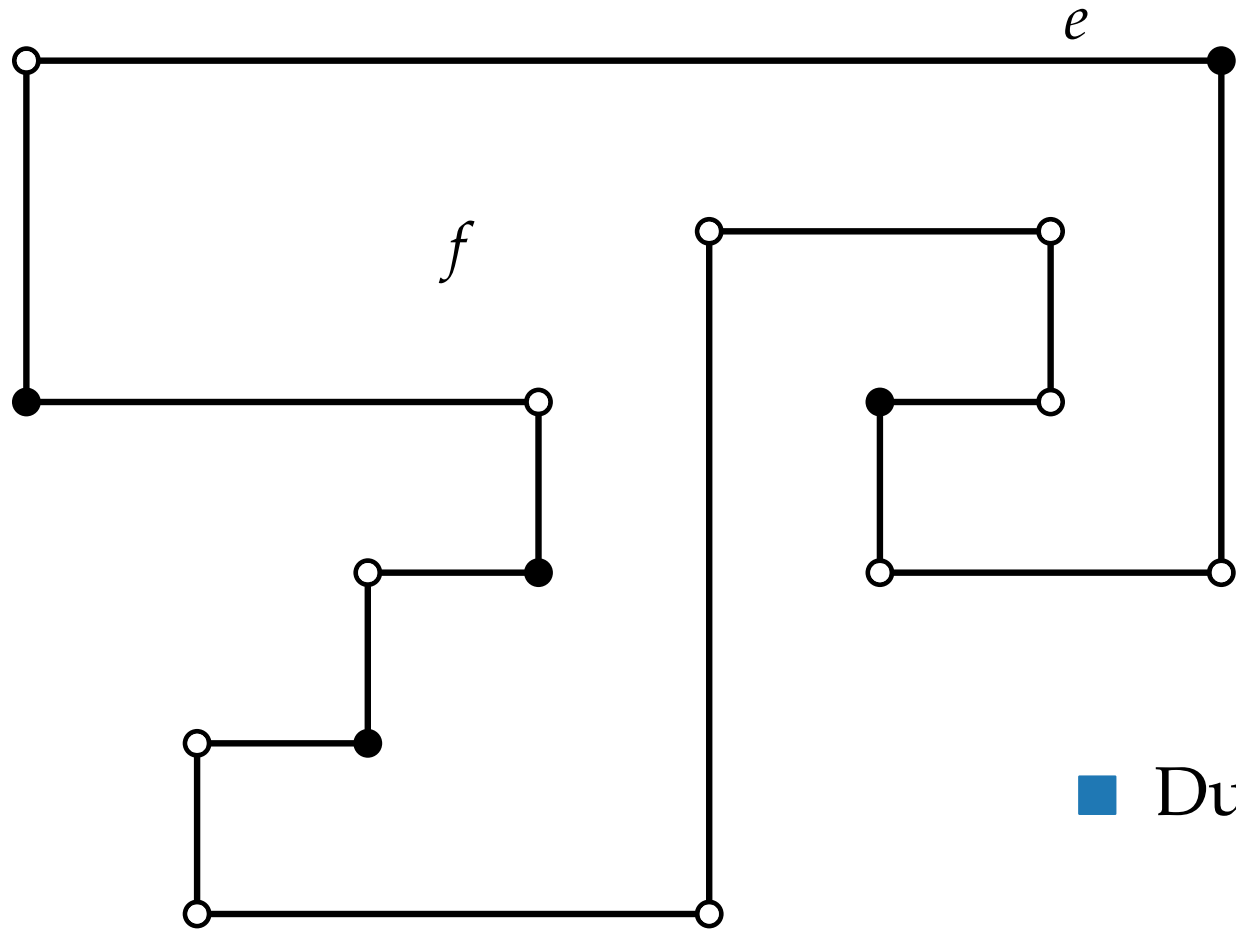


# Refinement of $(G, H)$ – Inner Face



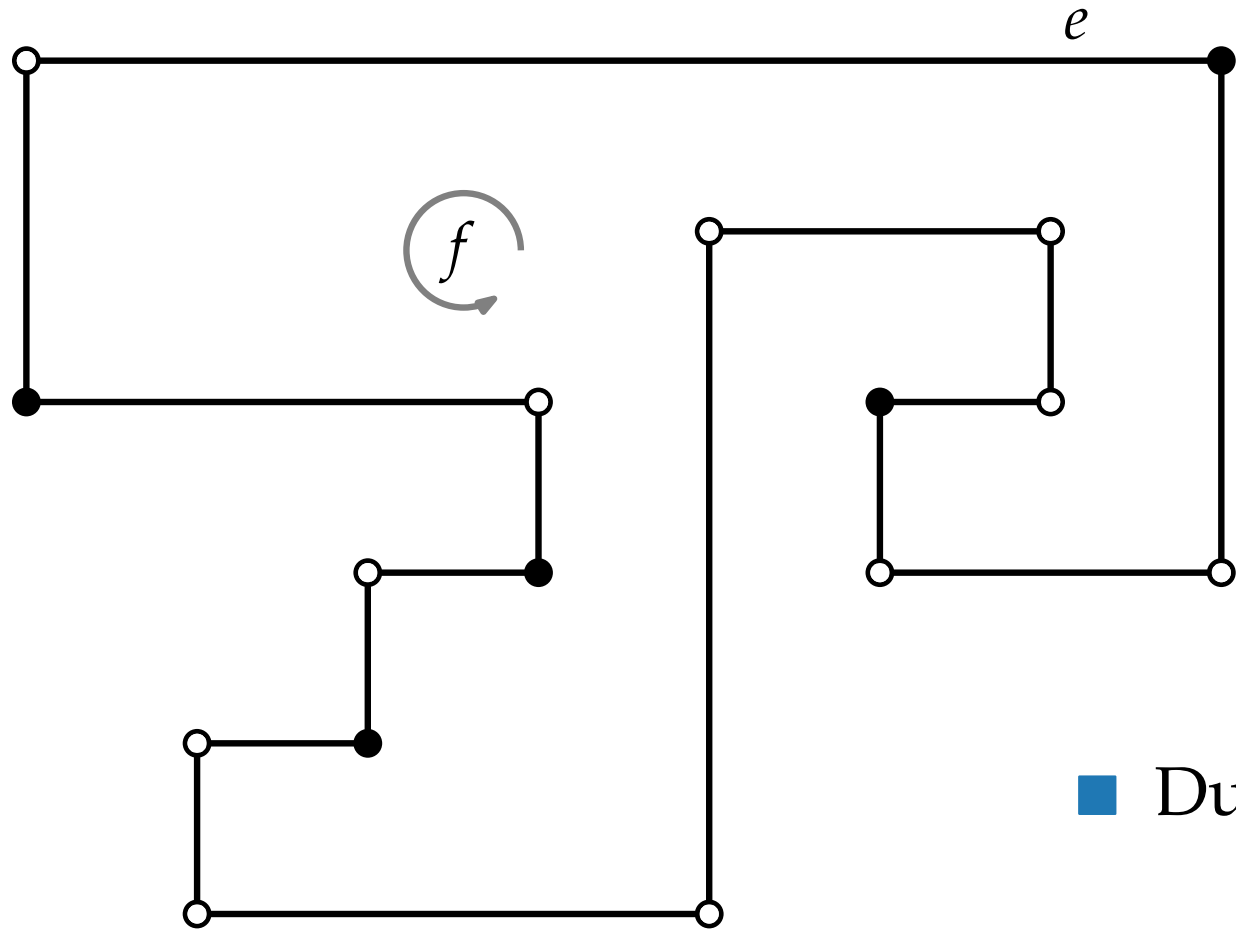
■ Dummy vertices for bends

# Refinement of $(G, H)$ – Inner Face



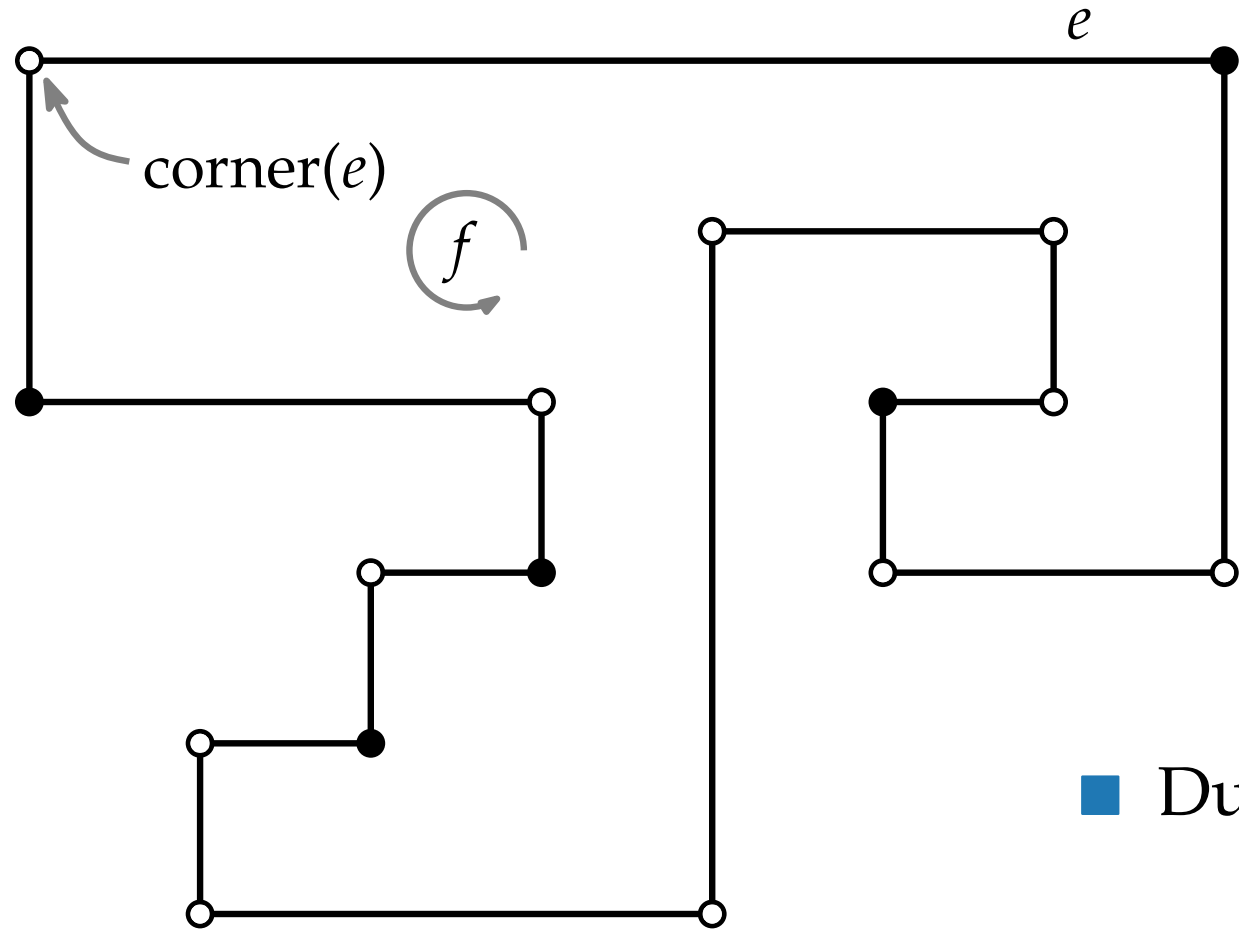
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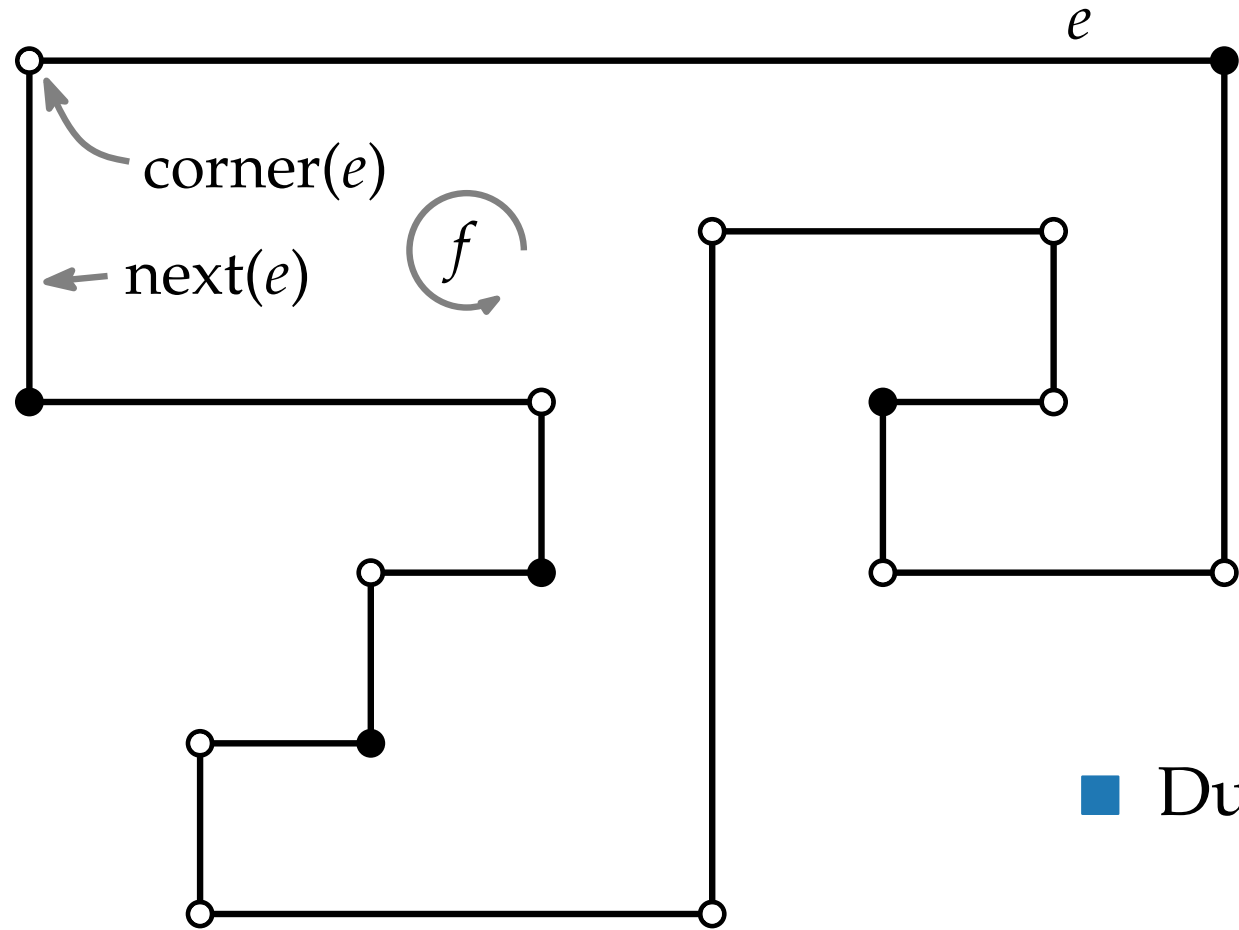
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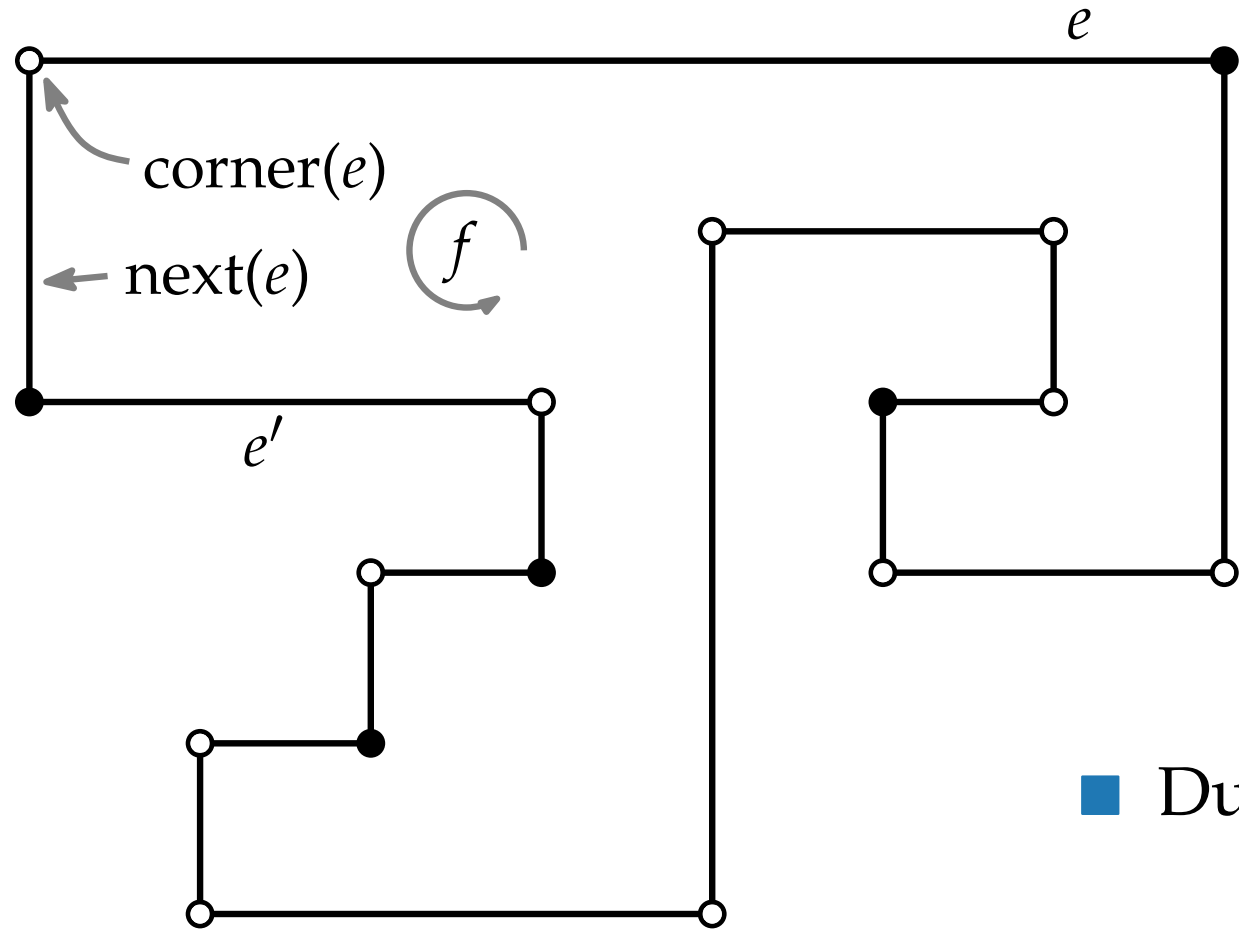


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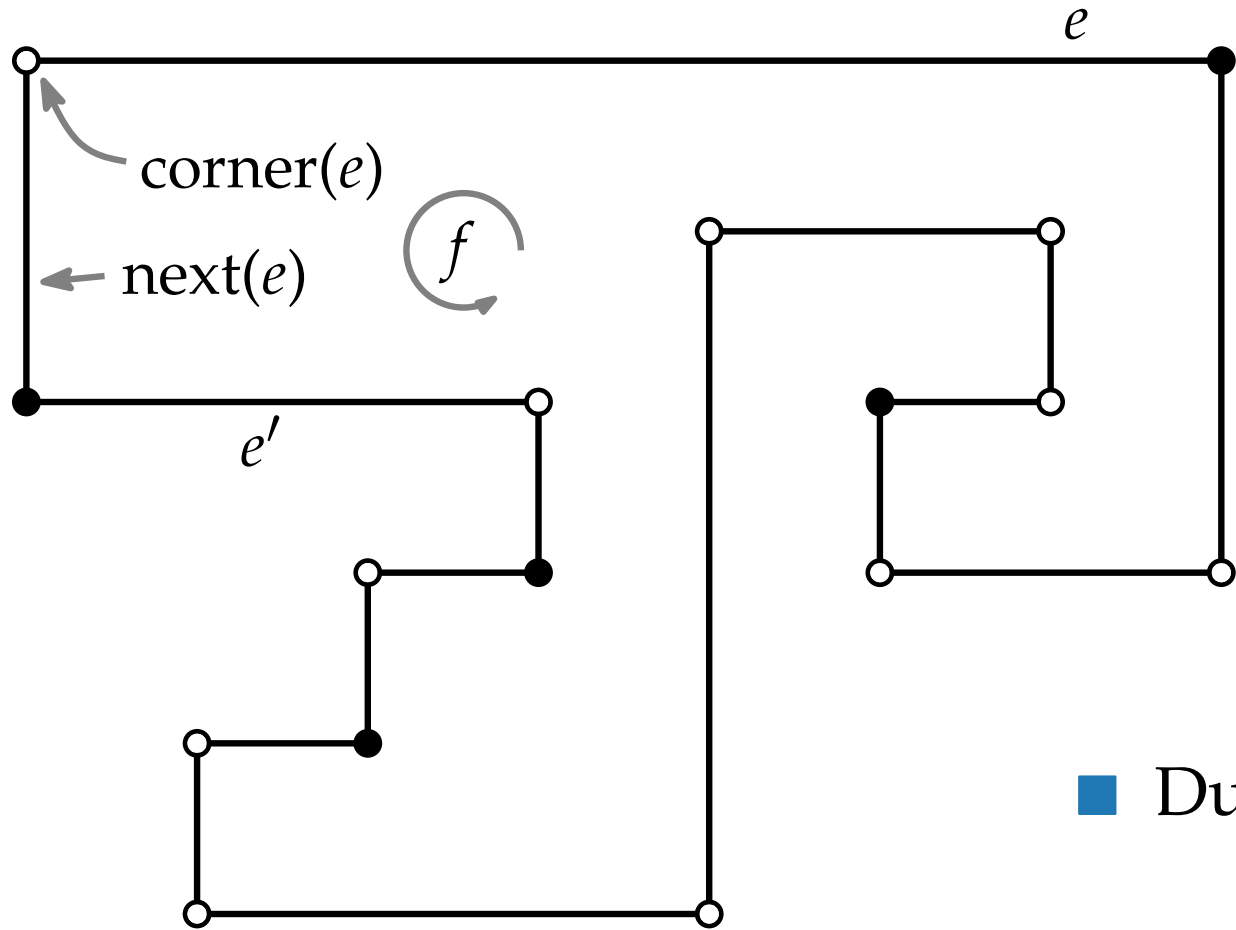
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# Refinement of $(G, H)$ – Inner Face



■ Dummy vertices for bends

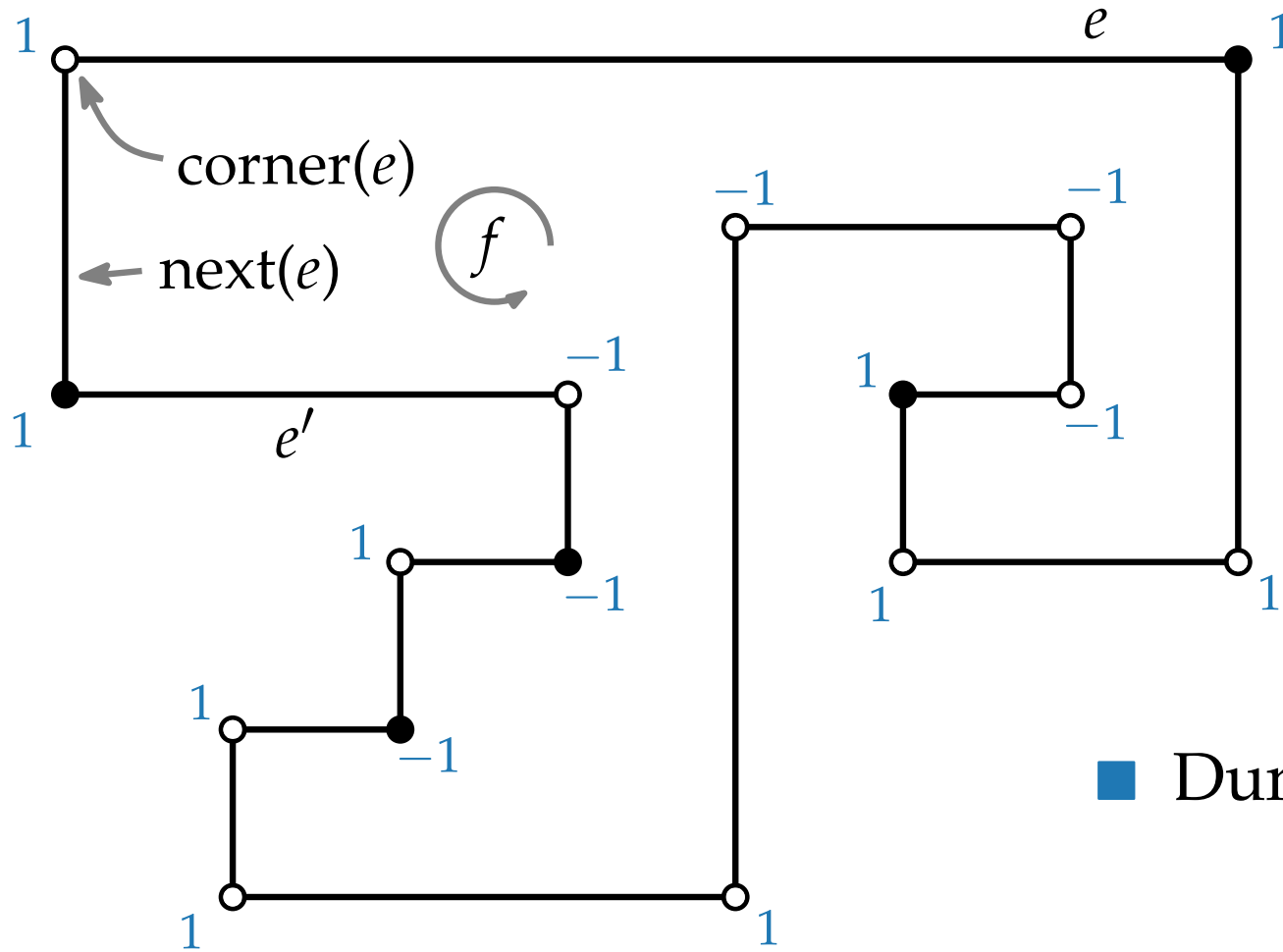
# Refinement of $(G, H)$ – Inner Face



■ Dummy vertices for bends

$$\text{turn}(e) = \begin{cases} 1 & \text{left turn} \\ 0 & \text{no turn} \\ -1 & \text{right turn} \end{cases}$$

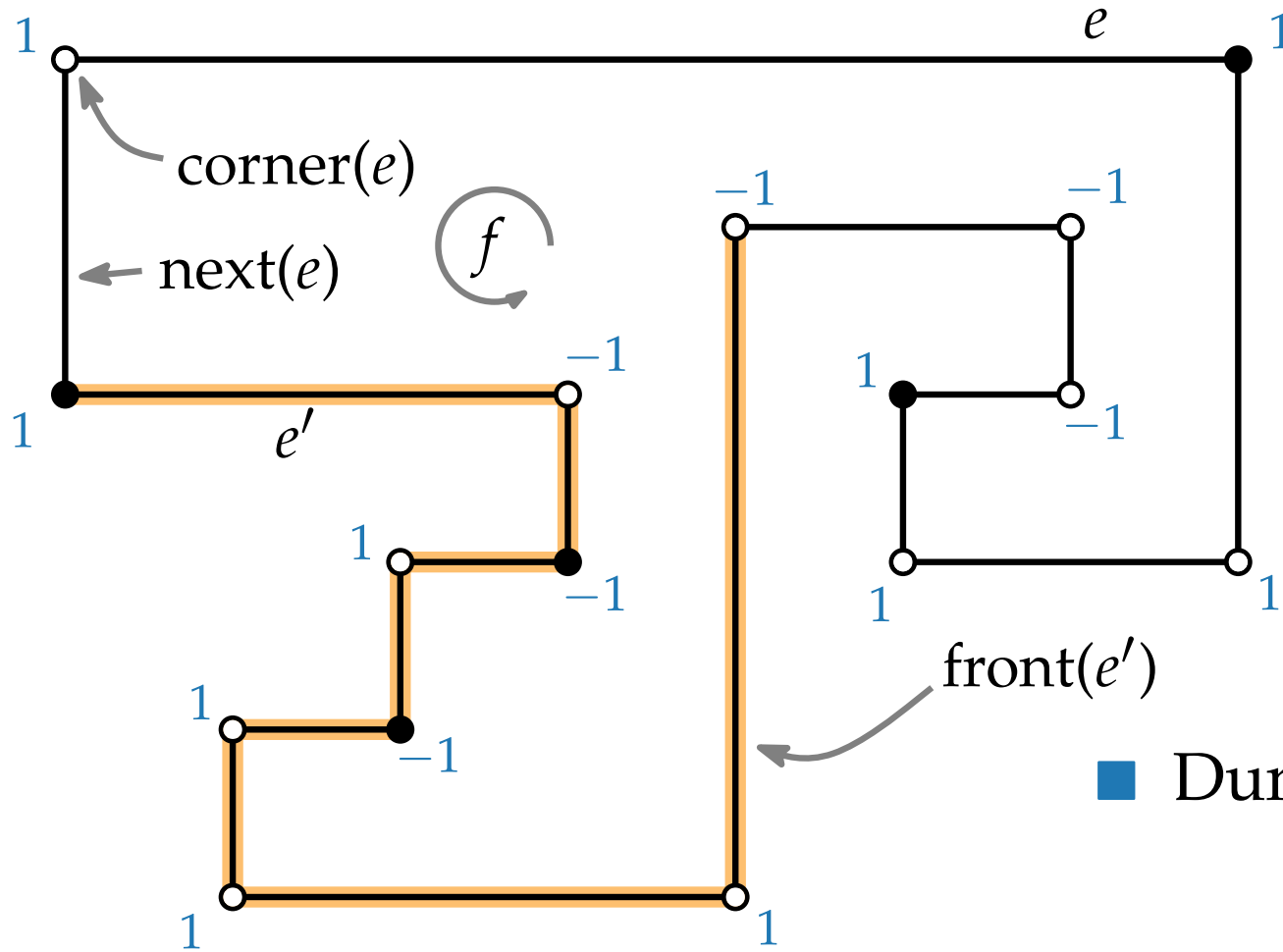
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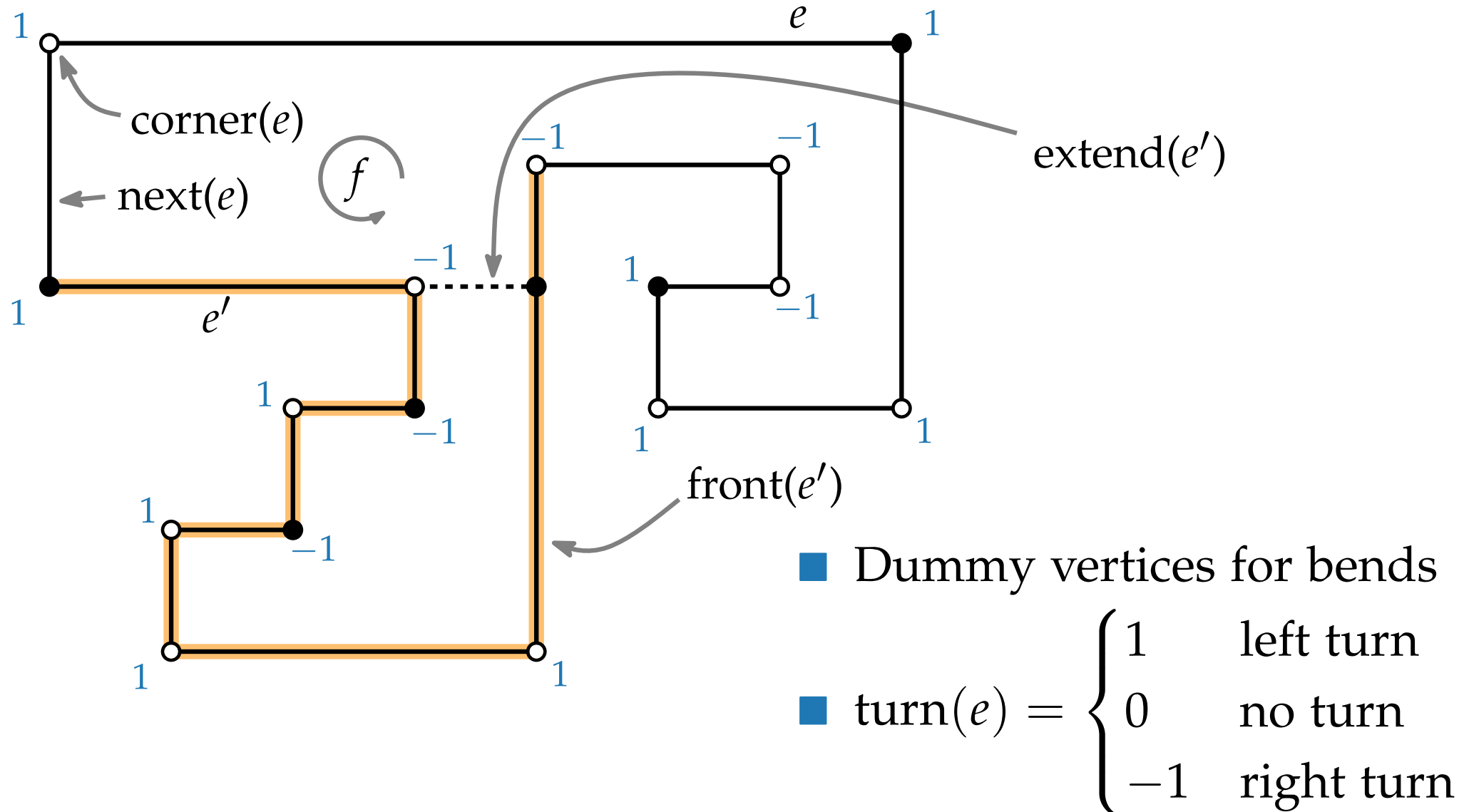
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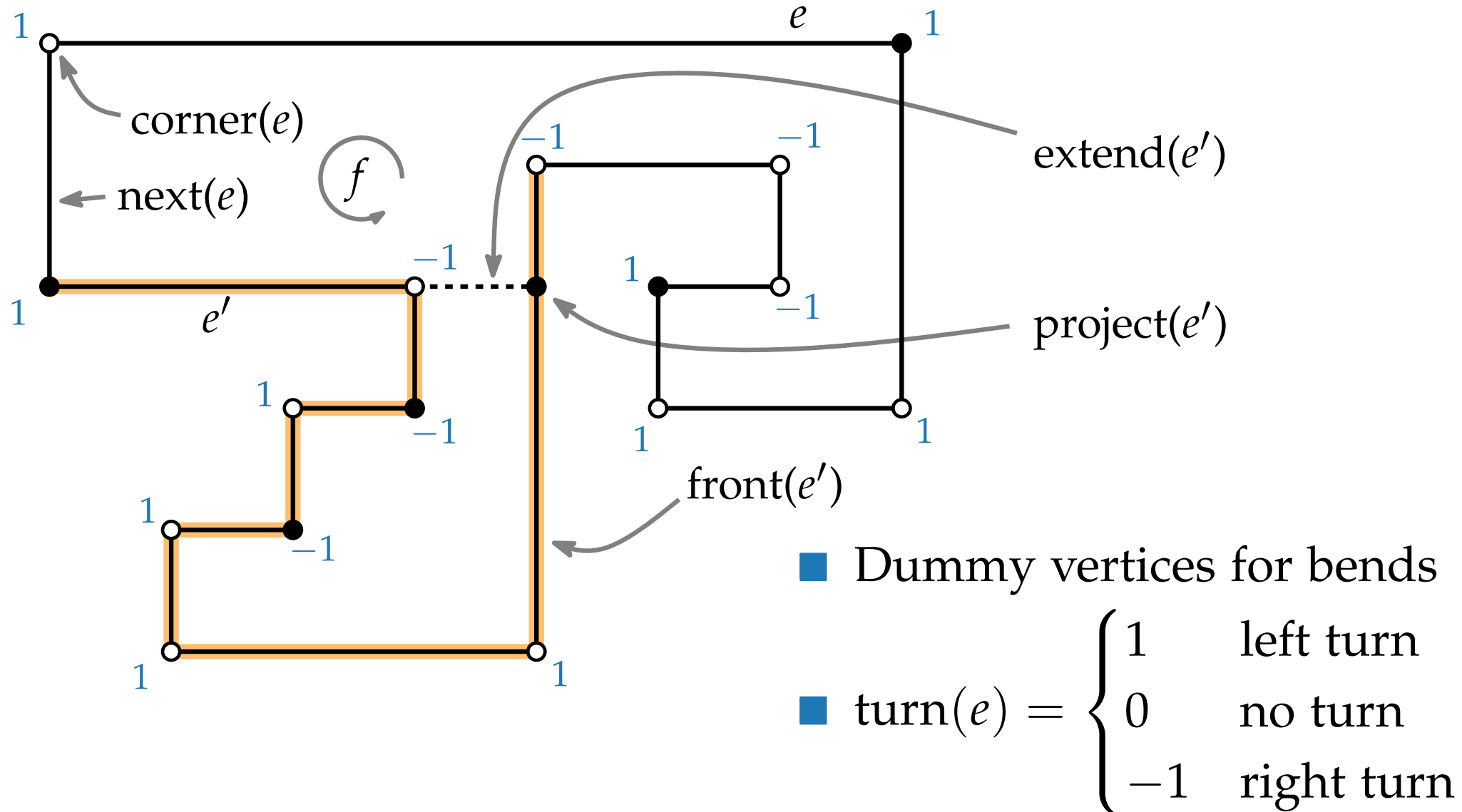
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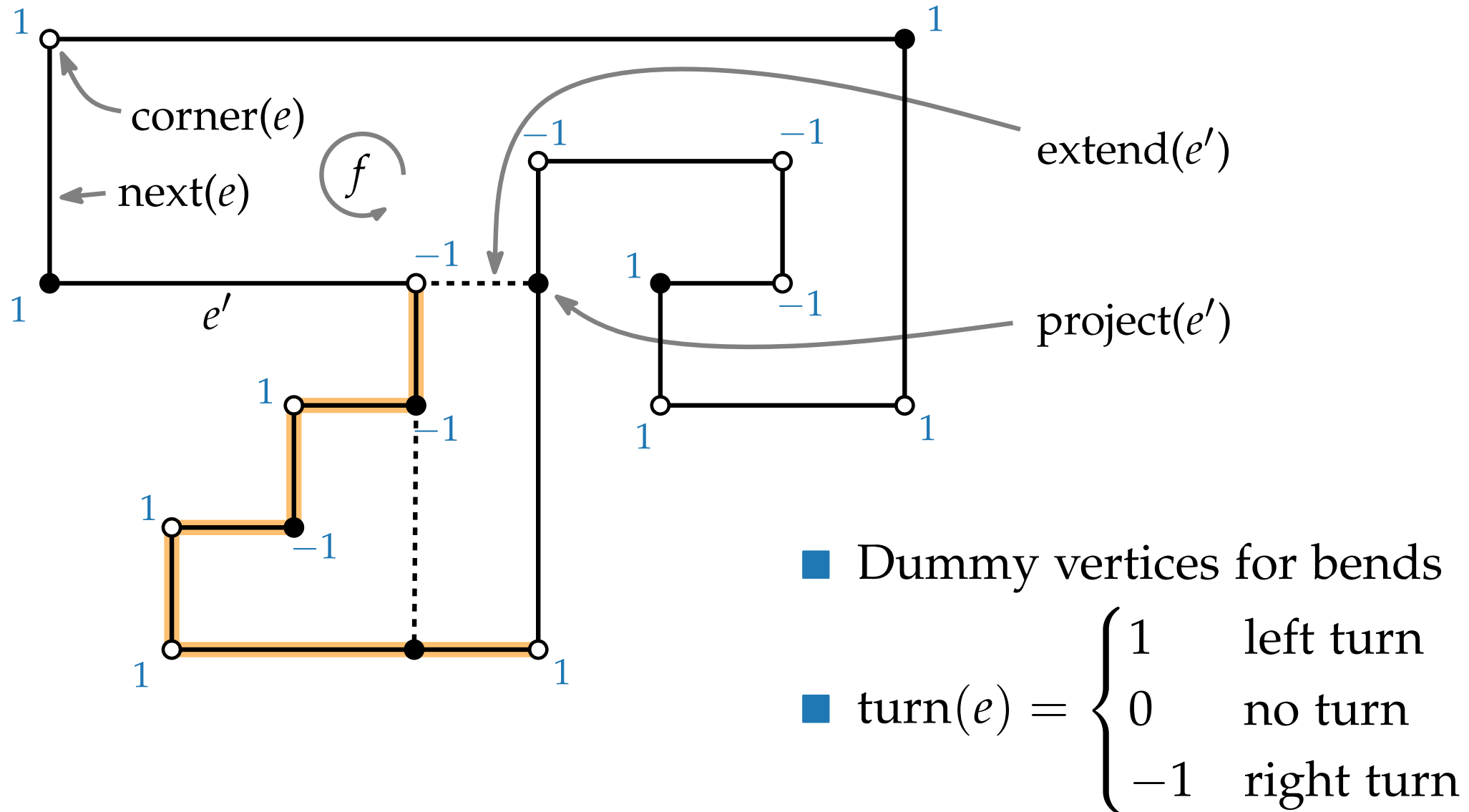
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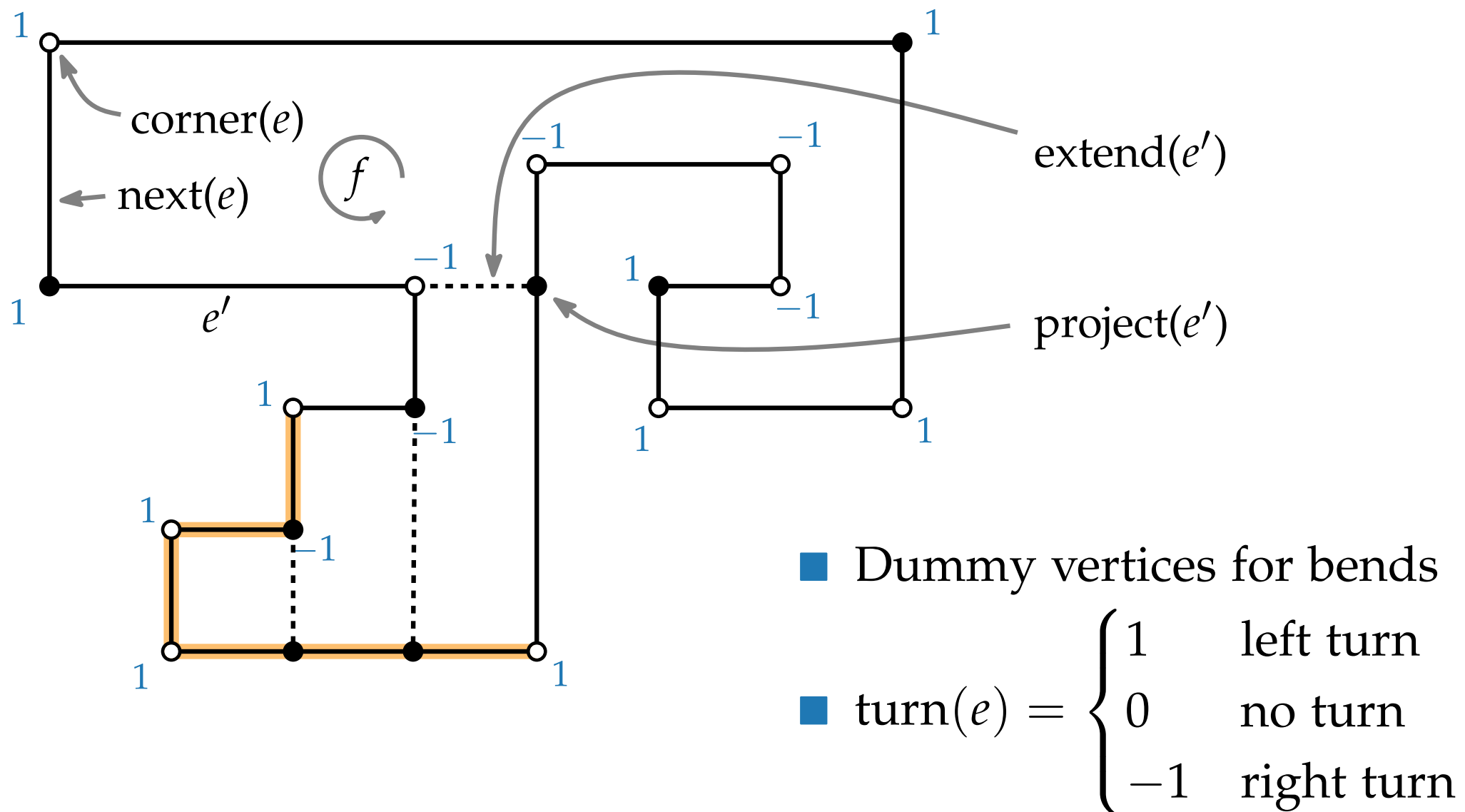


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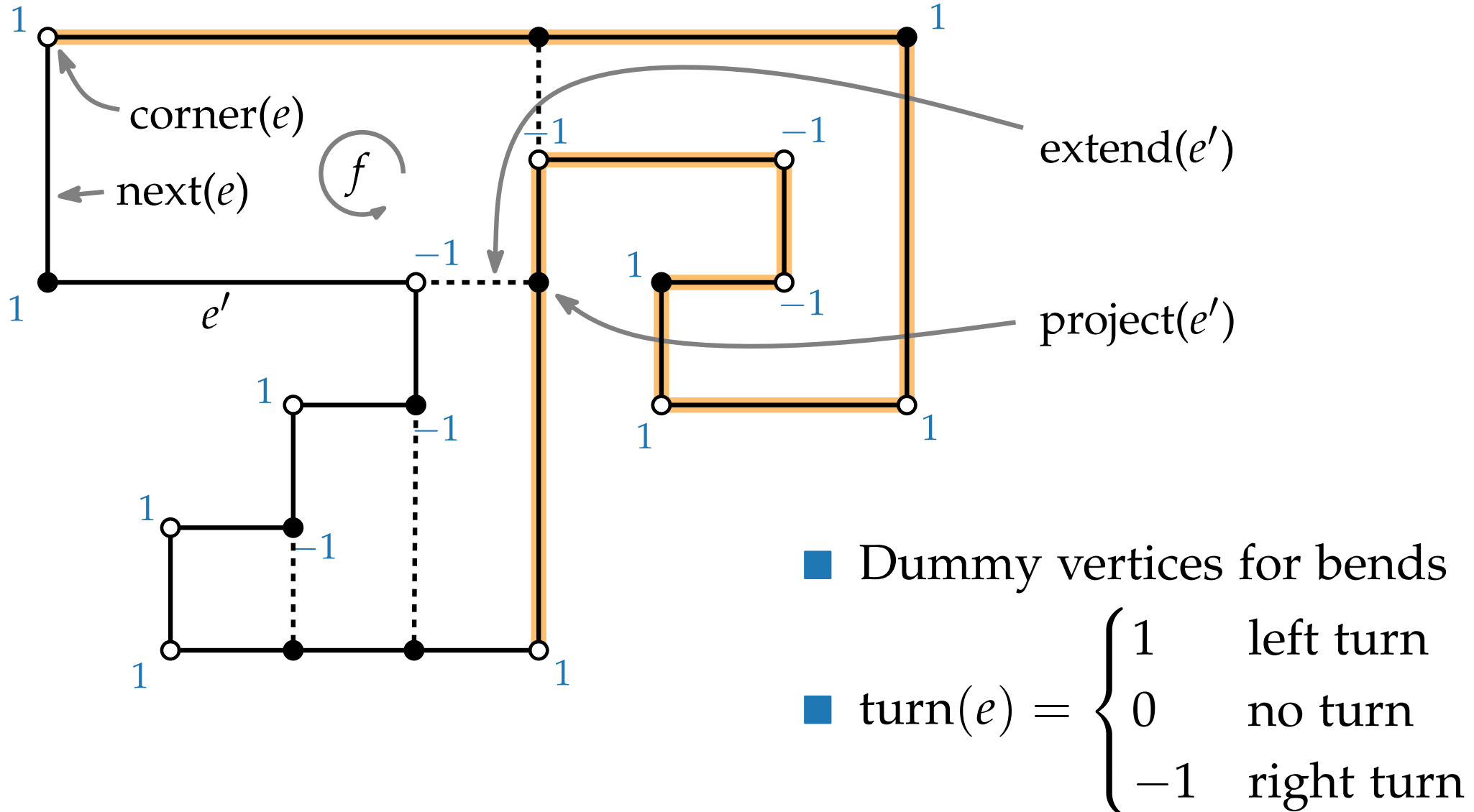




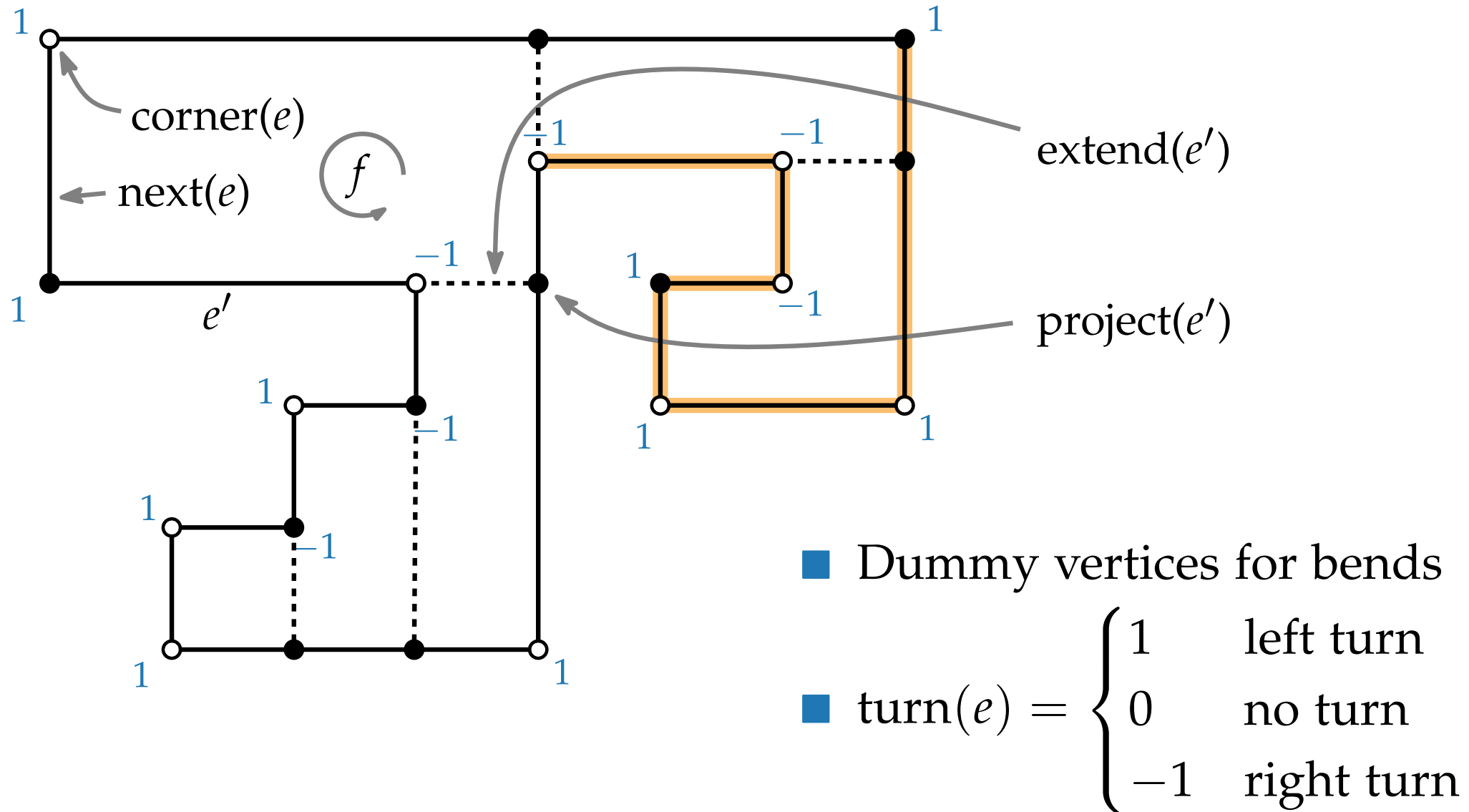
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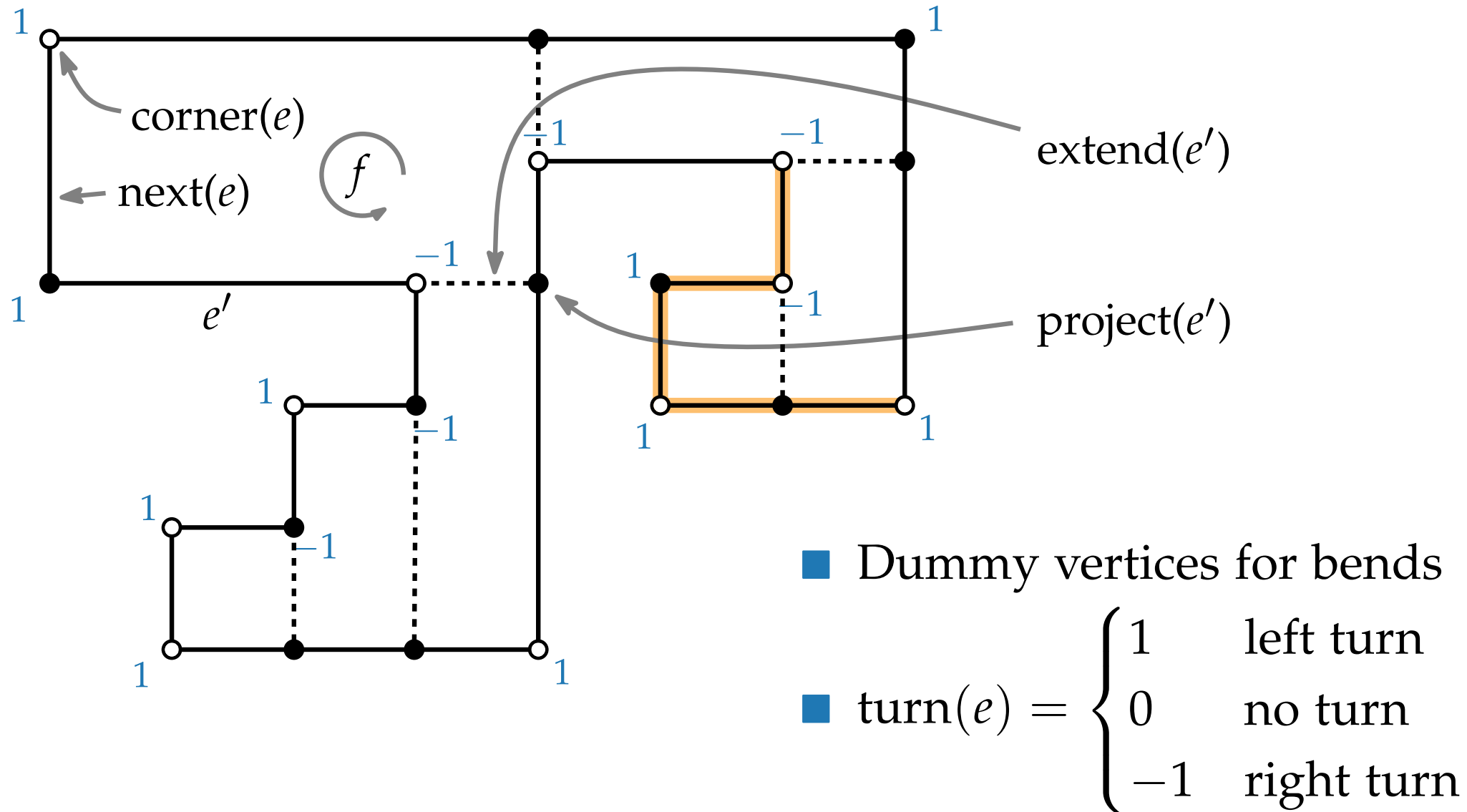
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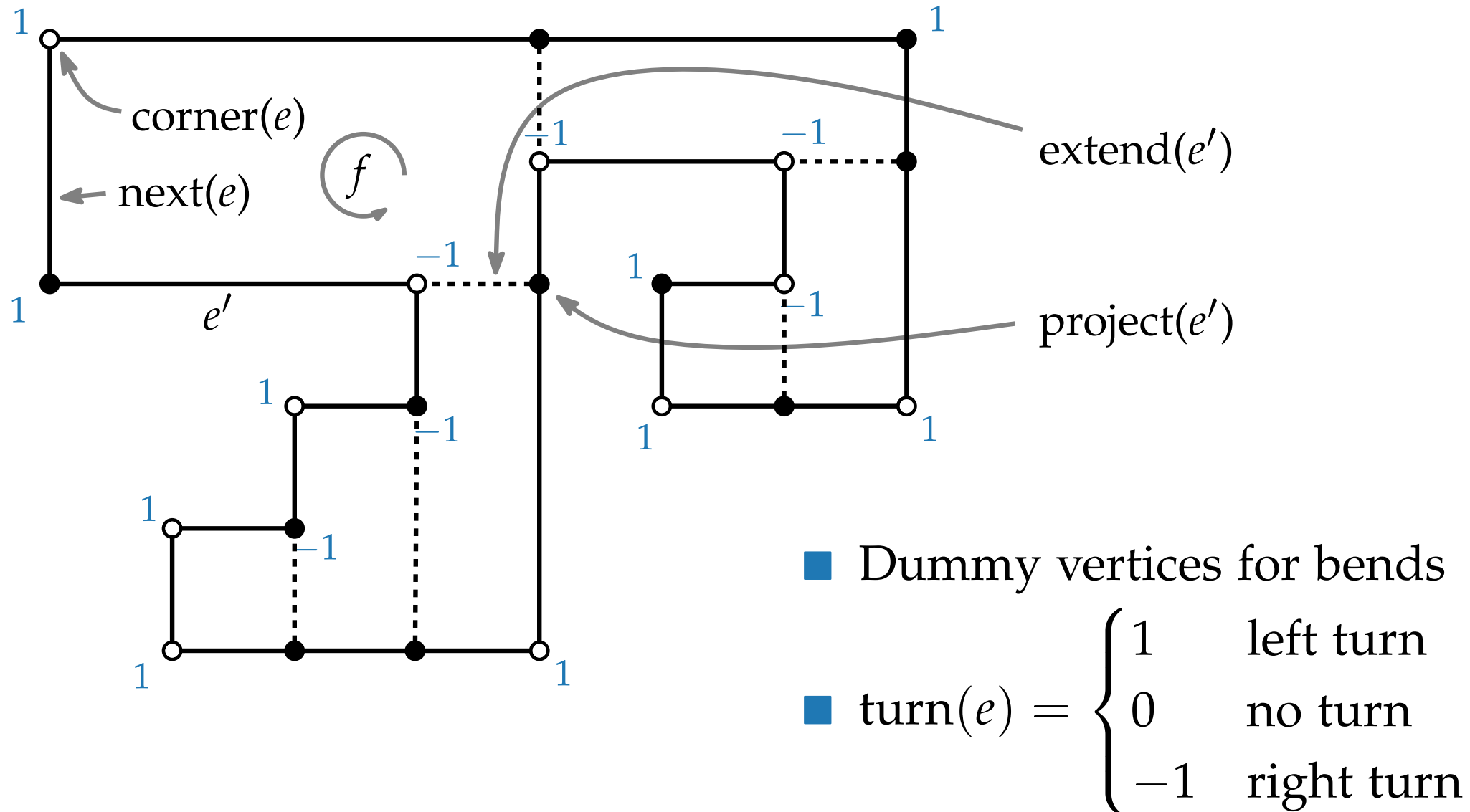
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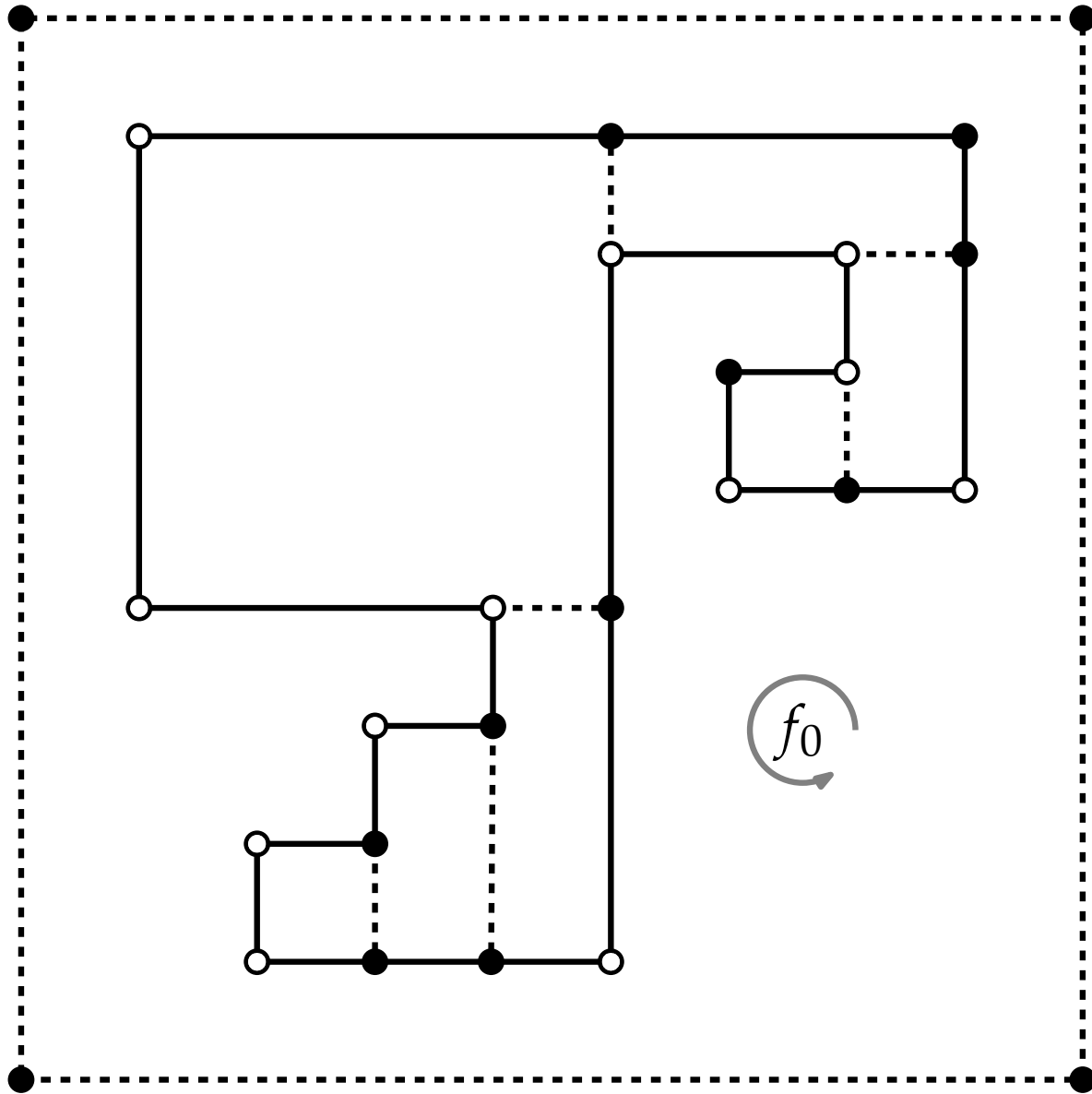
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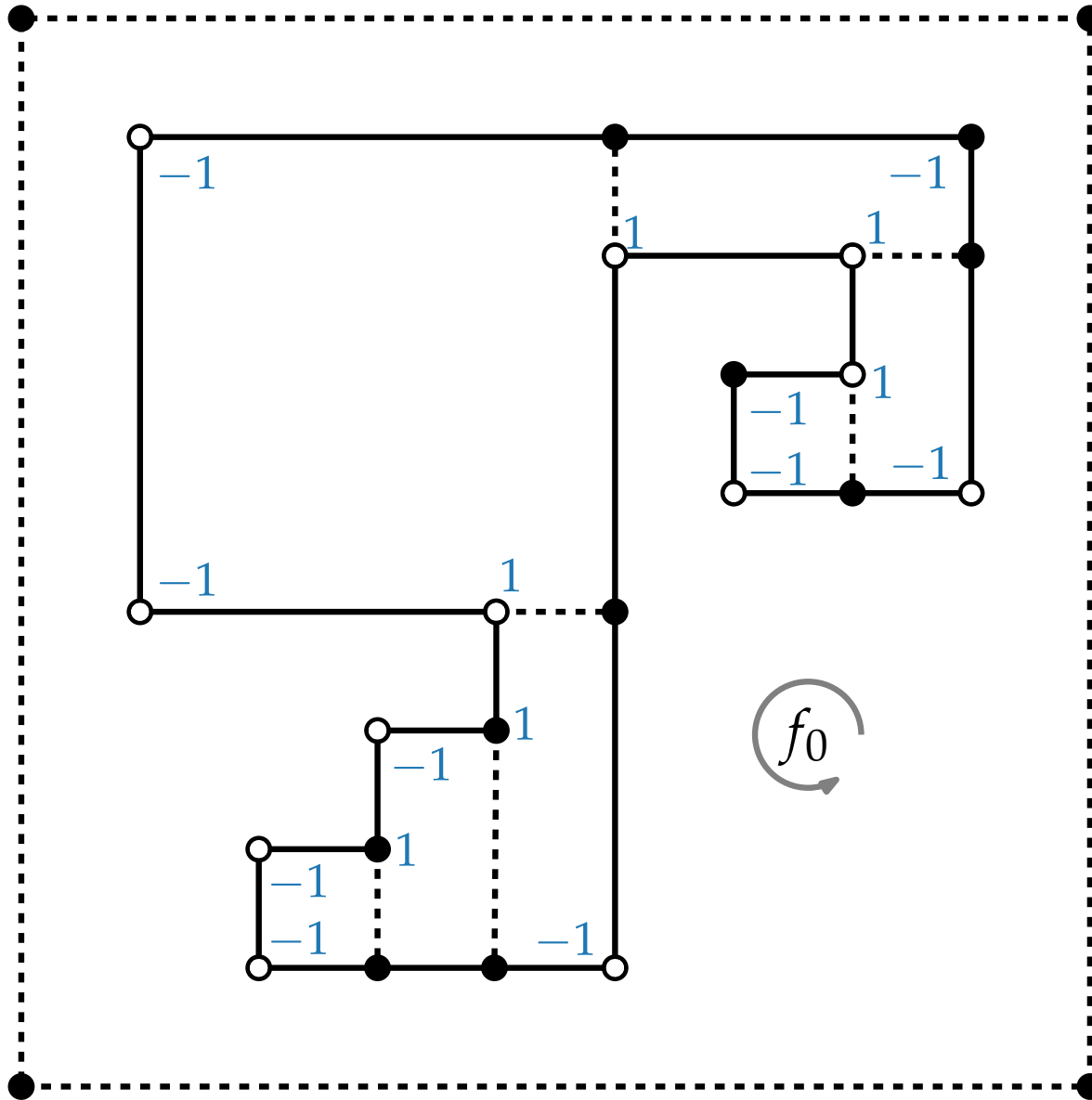
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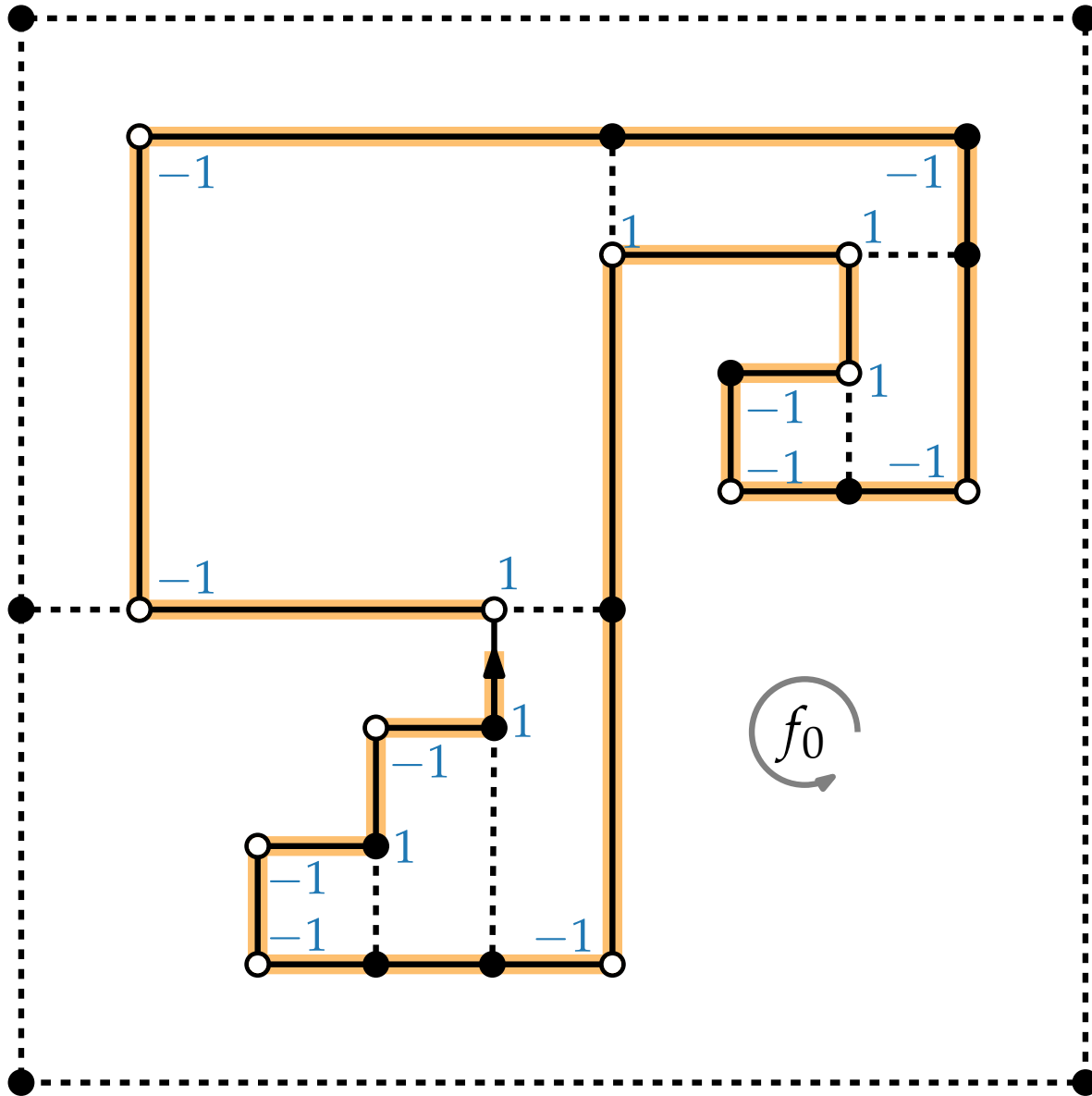
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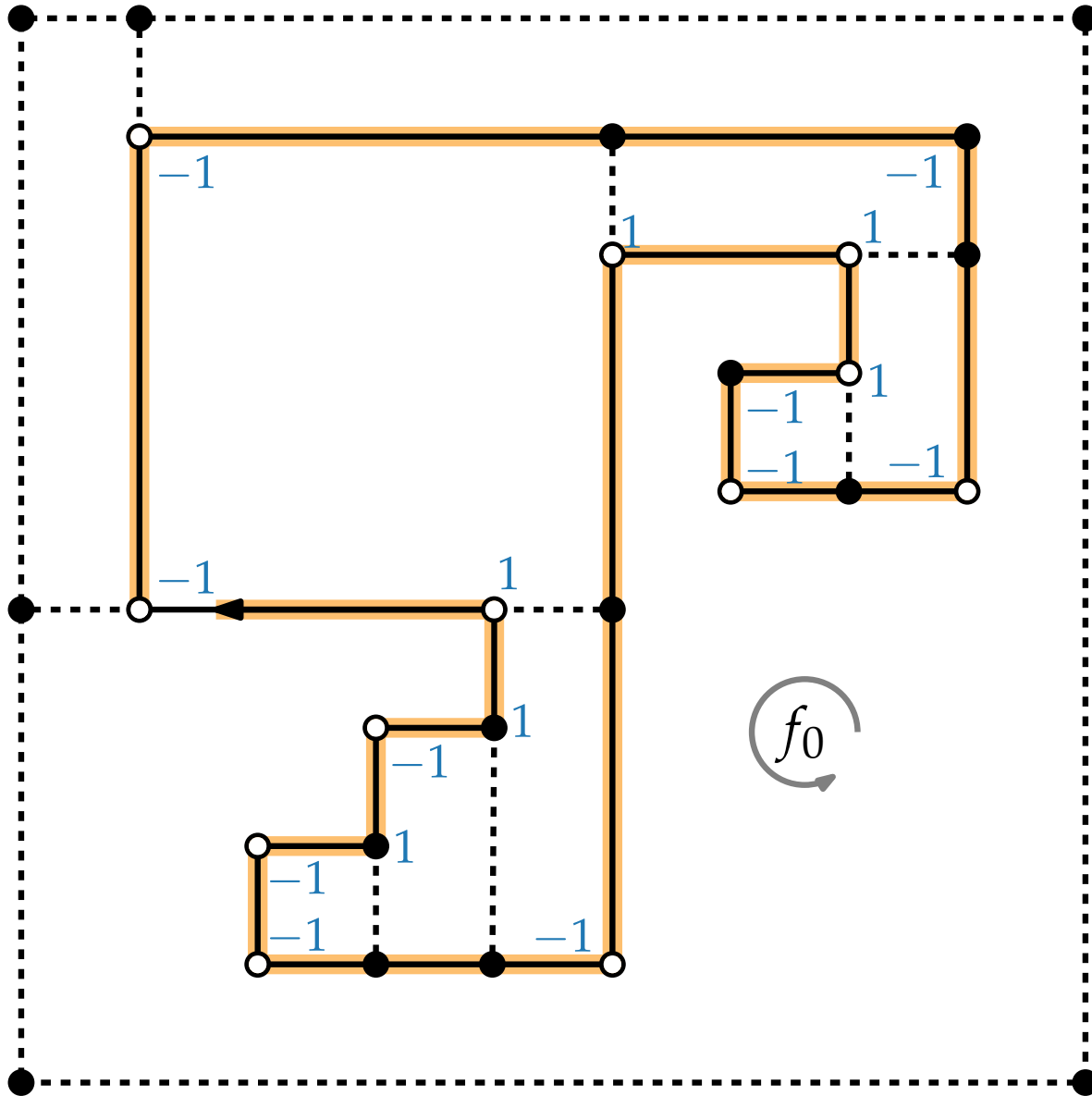


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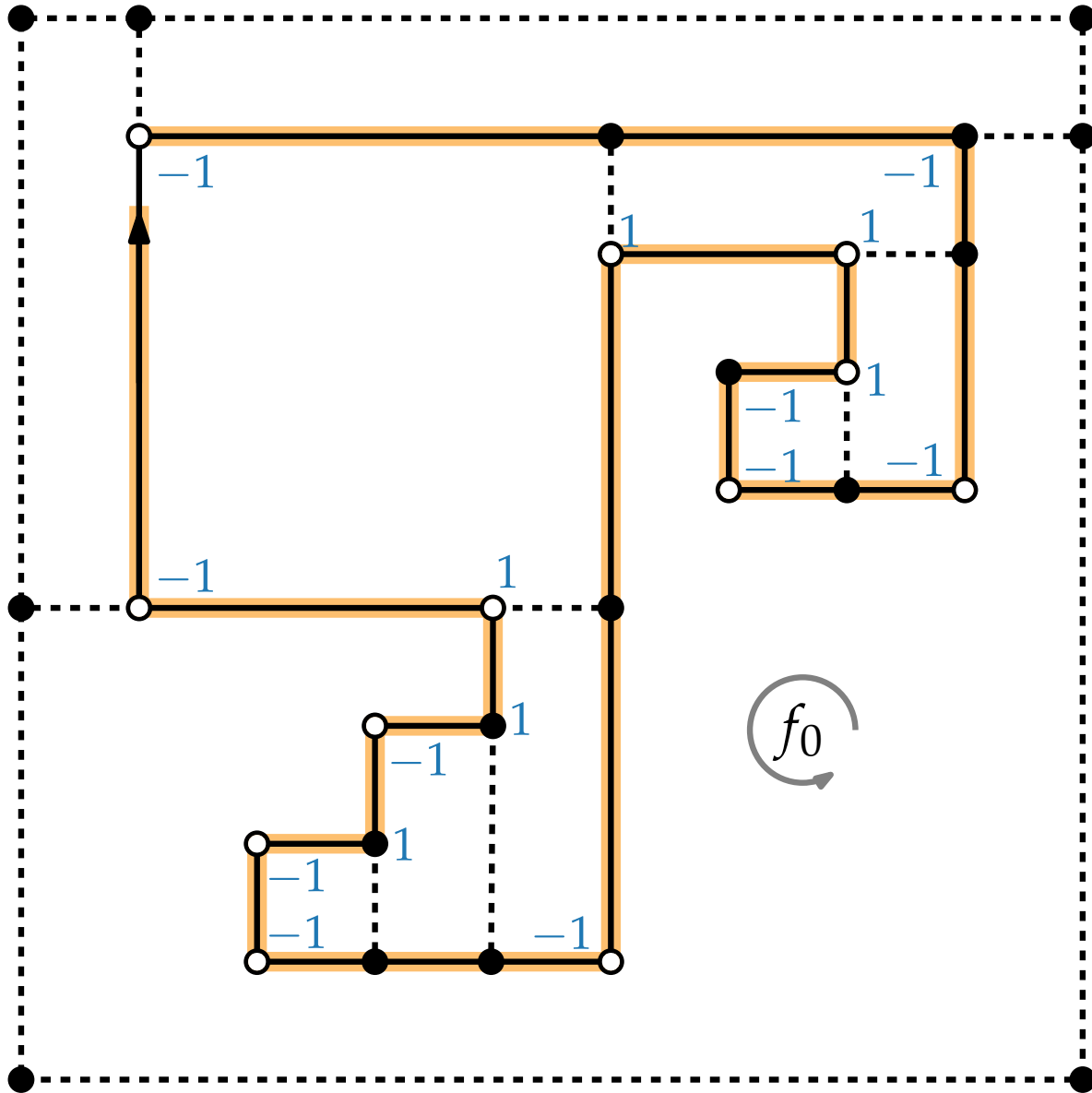




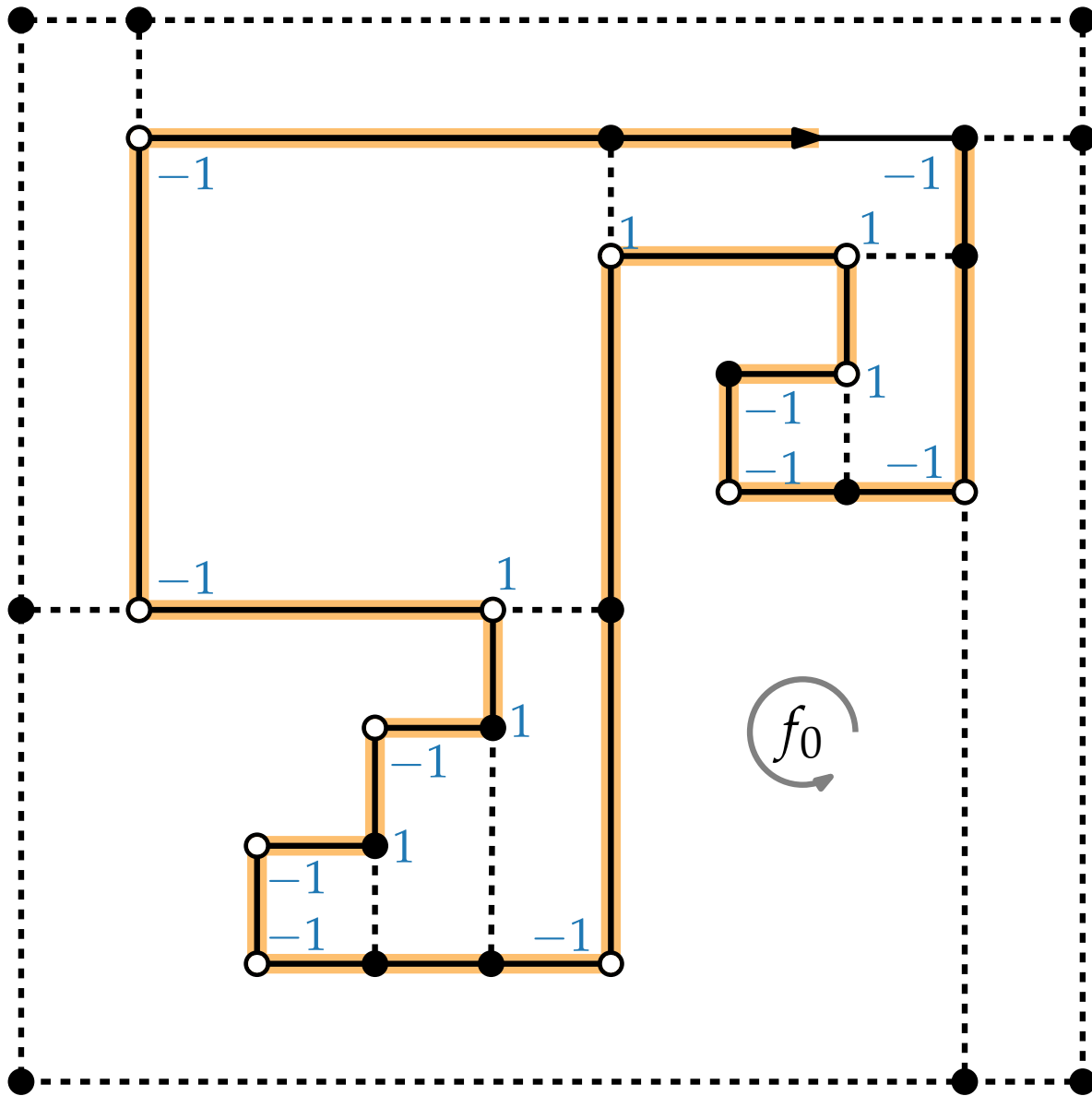
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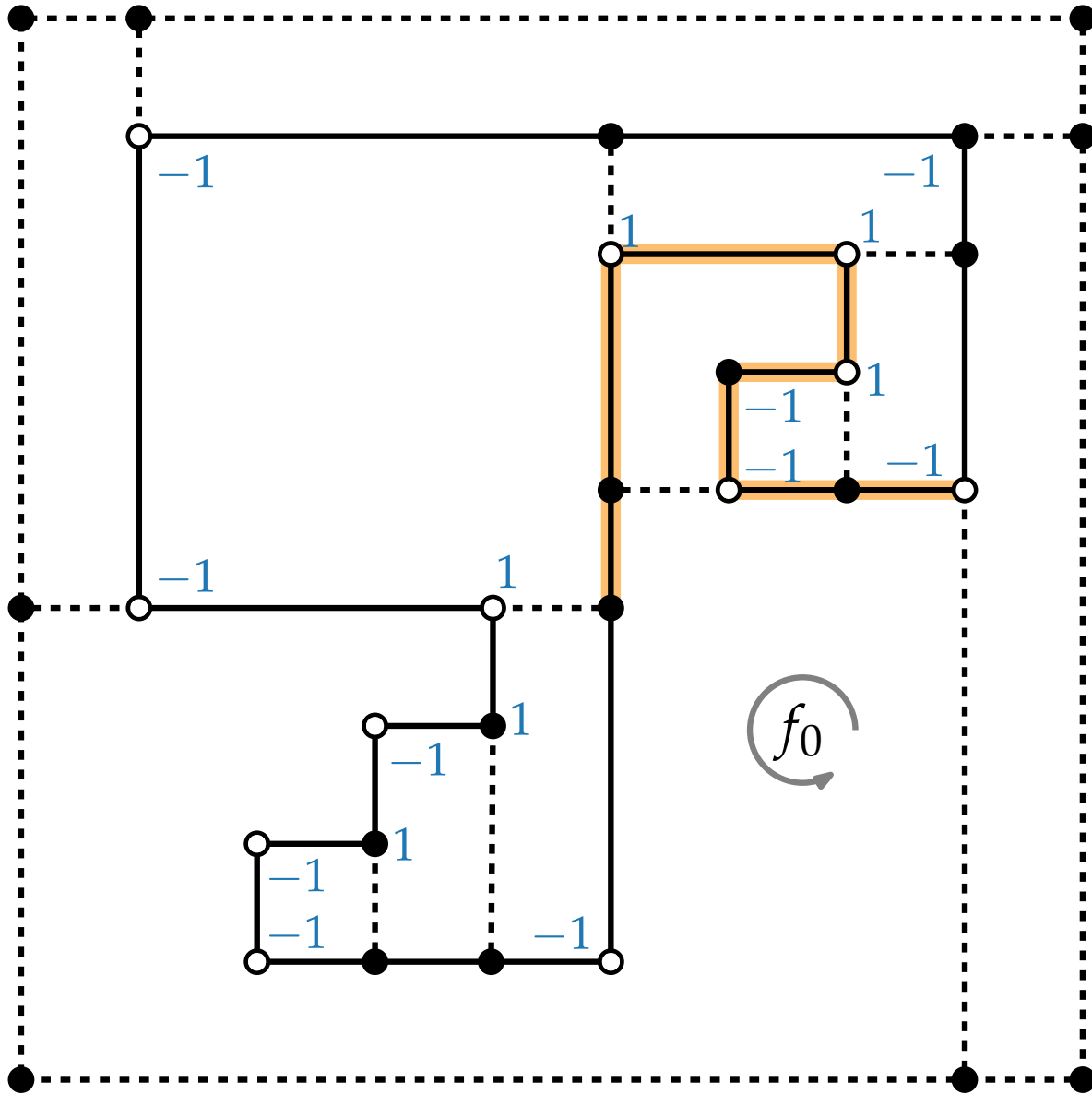
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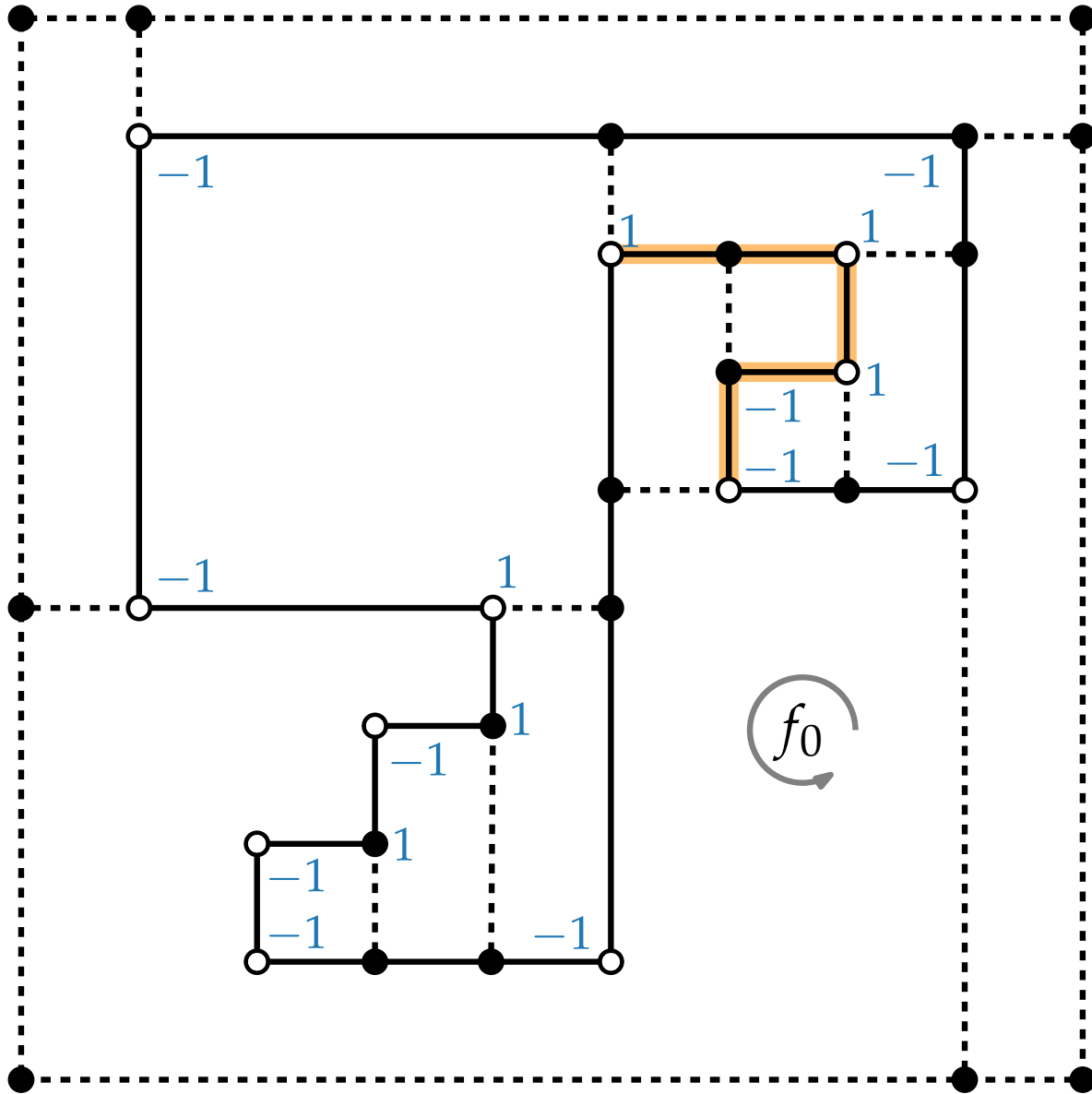
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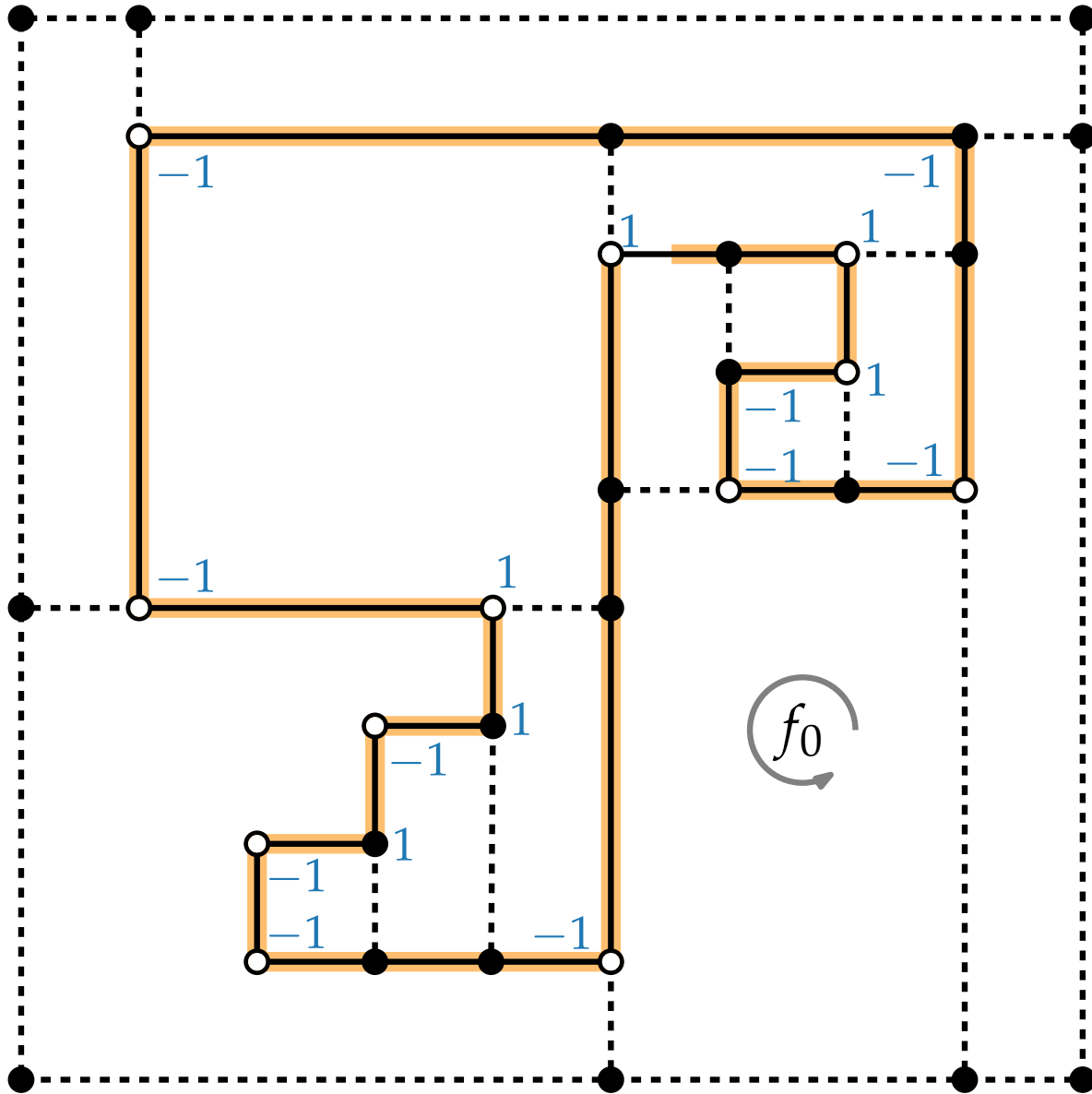
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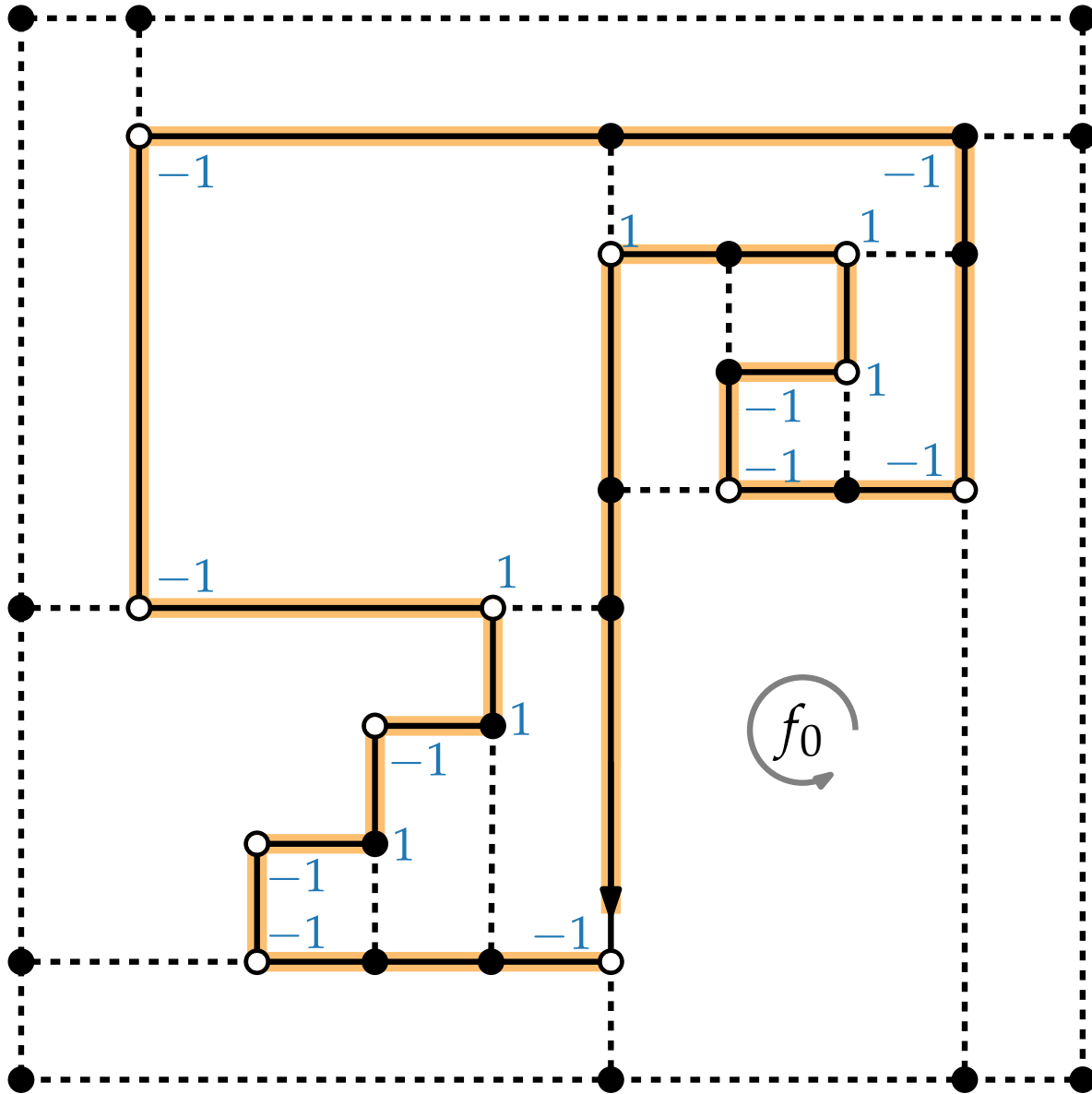
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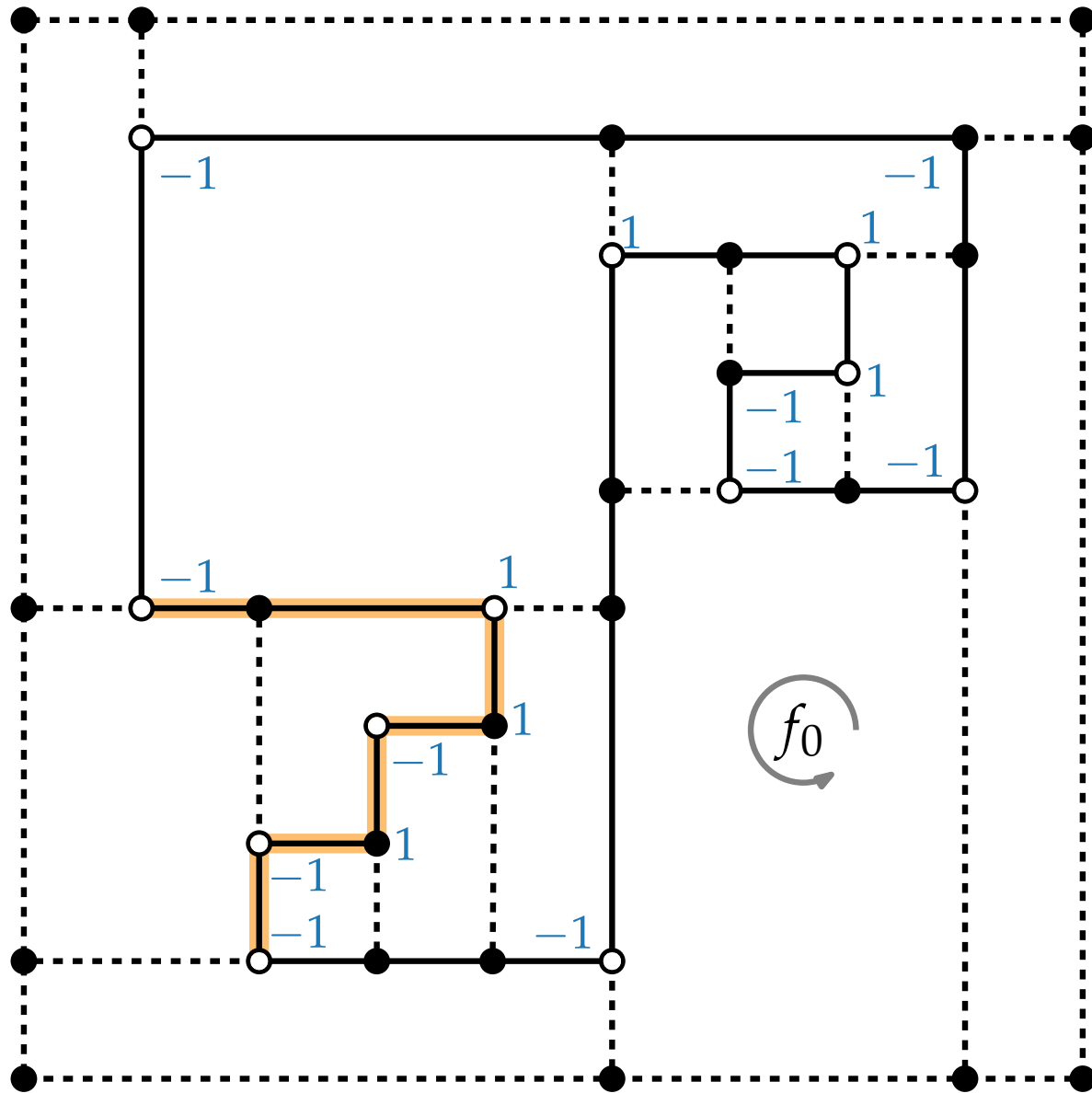
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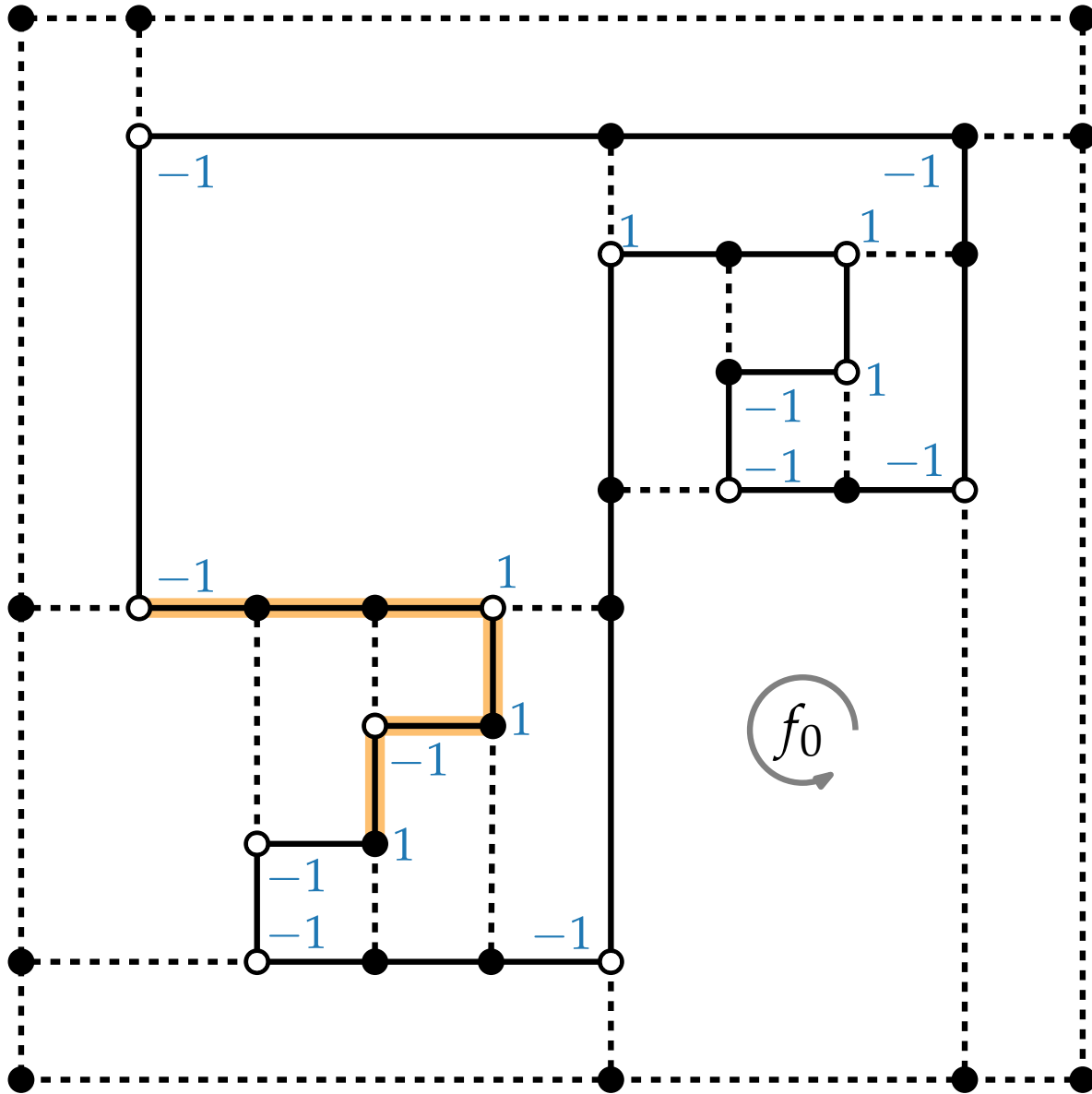


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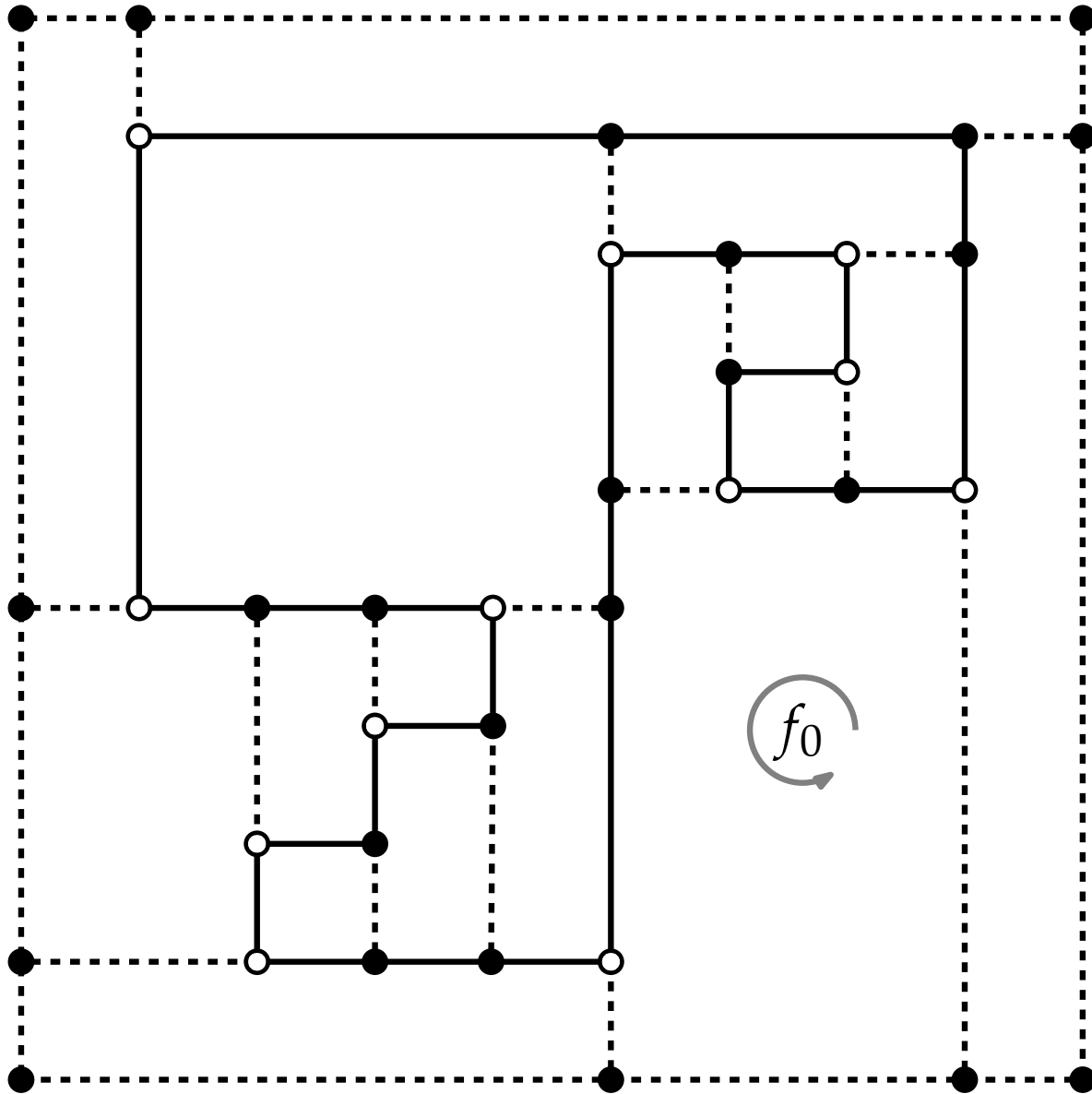




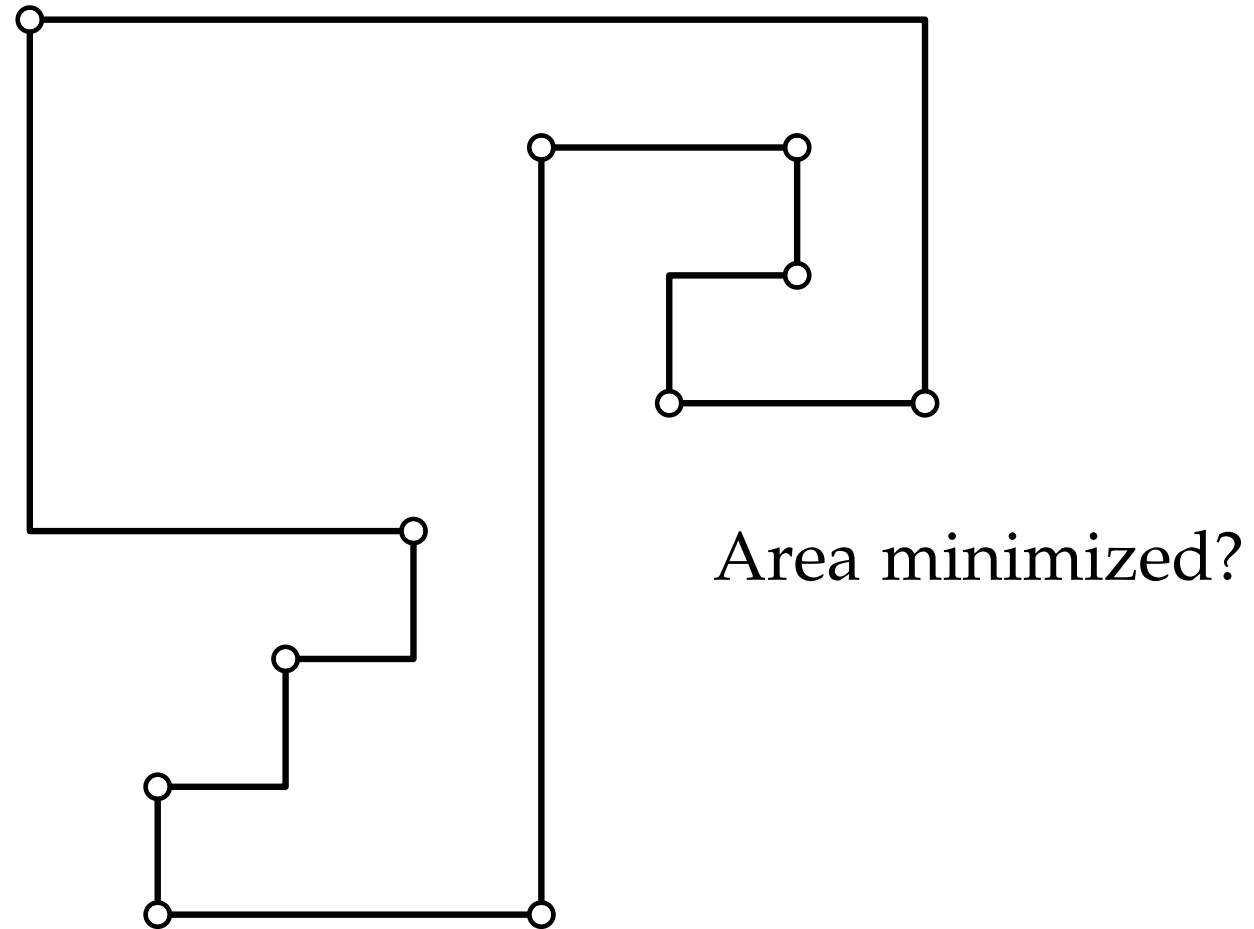
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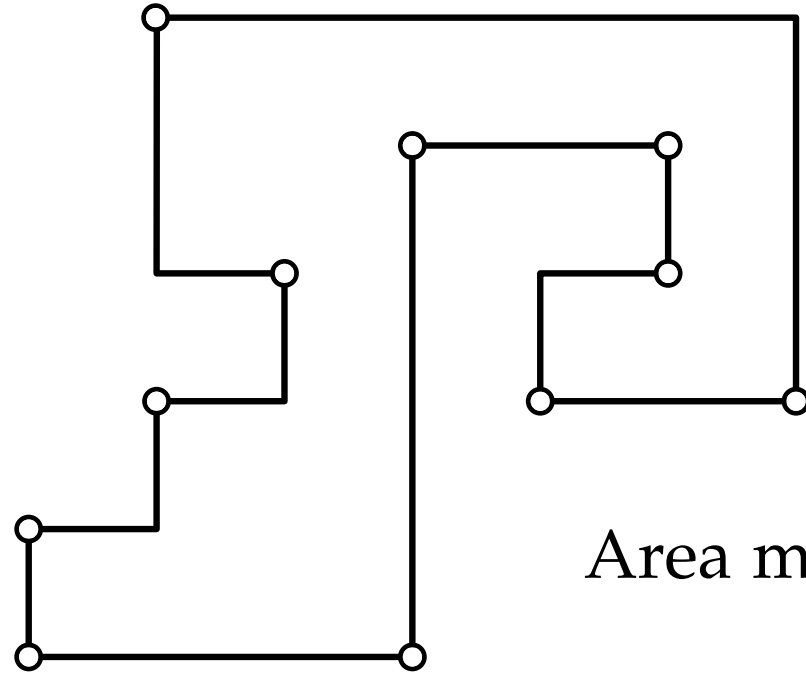
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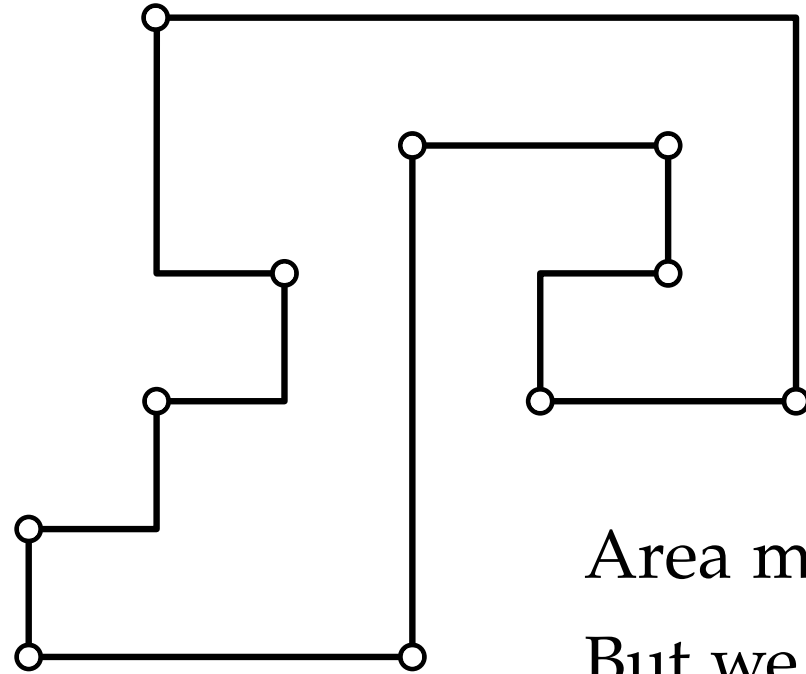


# Refinement of $(G, H)$ – Outer Face



Area minimized? **No!**

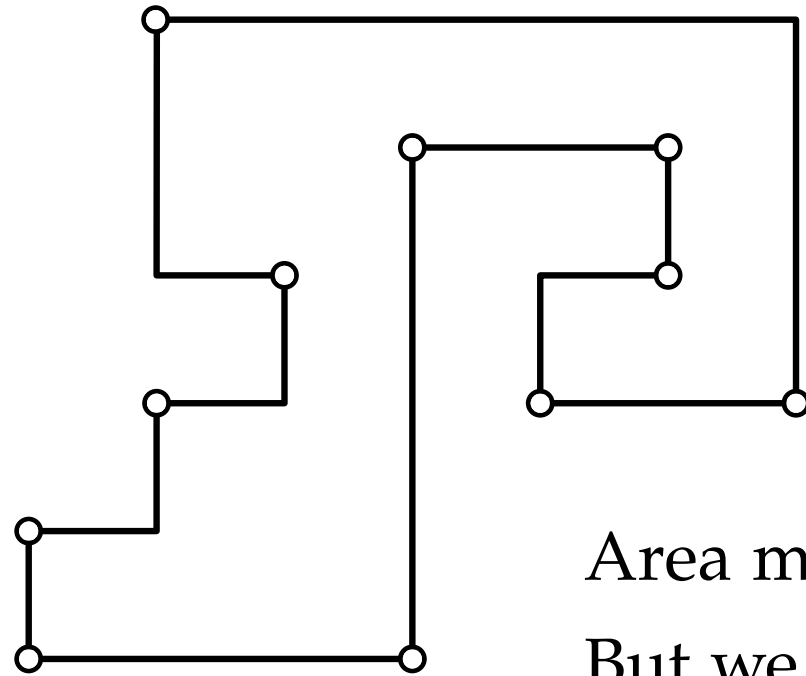
# Refinement of $(G, H)$ – Outer Face



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But we get bound  $O((n + b)^2)$  on the area.

# Refinement of $(G, H)$ – Outer Face



Area minimized? **No!**

But we get bound  $O((n + b)^2)$  on the area.

**Theorem.** [Patrignani 2001]  
Compaction for given orthogonal representation is in general NP-hard.