

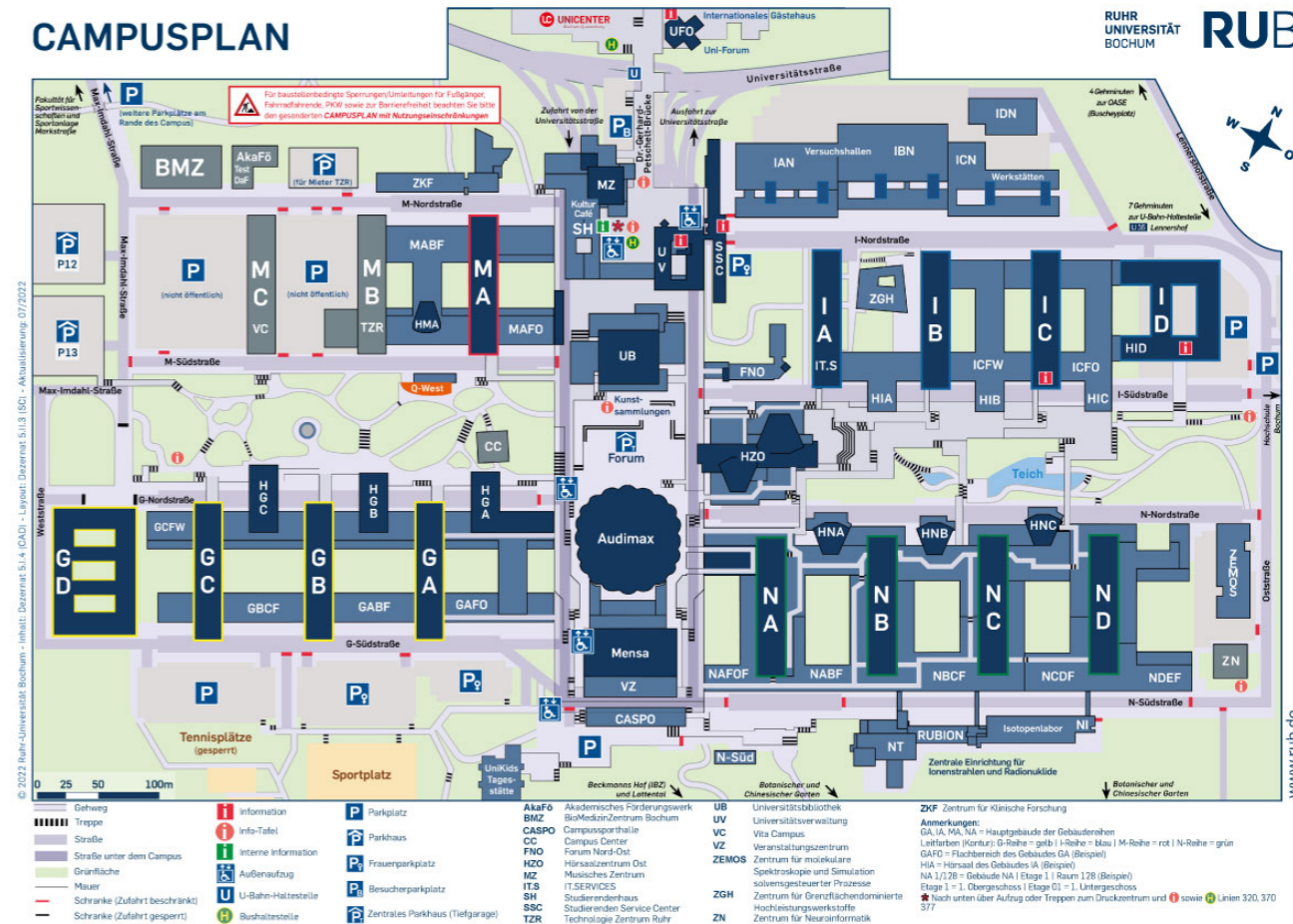
# Hardness and modifications of the weak graph distance

13.03.2024

Maike Buchin, **Wolf Kießler**

# Distance measures for immersed graphs

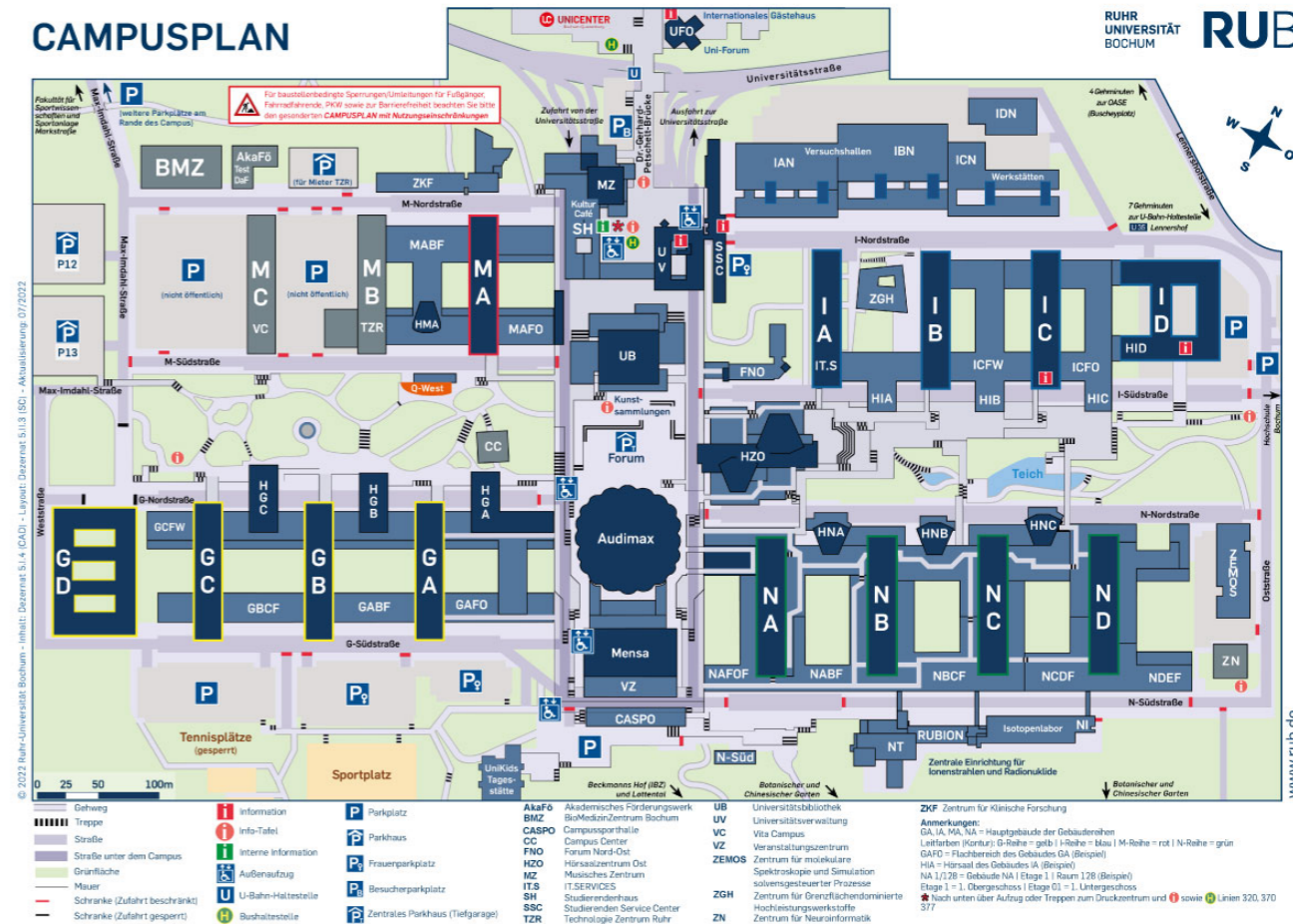
Representations of geometric networks:



# Distance measures for immersed graphs

Representations of geometric networks:

- Embedded graphs: Drawings without crossings
- Immersed graphs: Drawings that may contain crossings
- Plane graphs: Graphs embedded in  $\mathbb{R}^2$

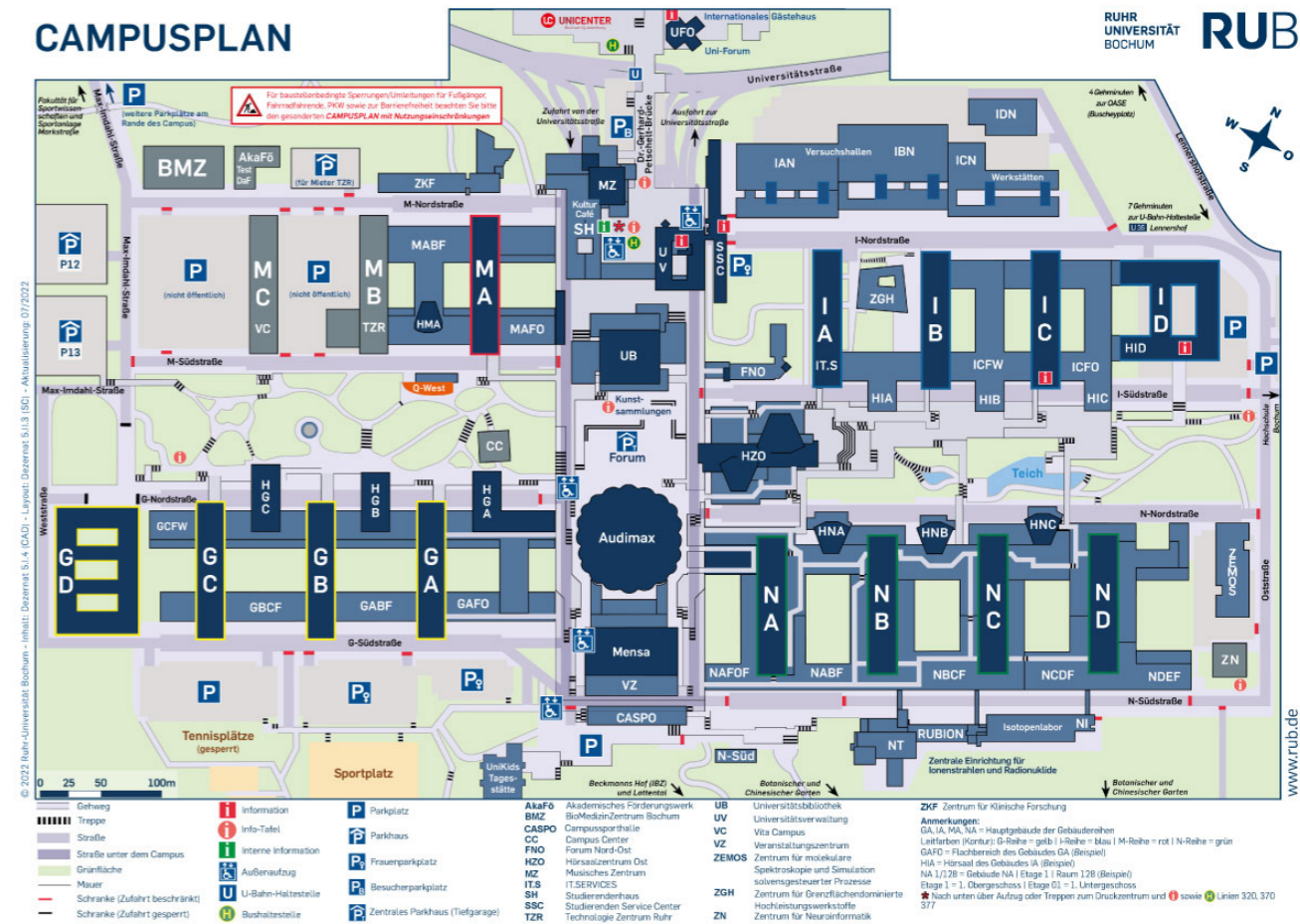


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We consider non-degenerate straight-line immersions







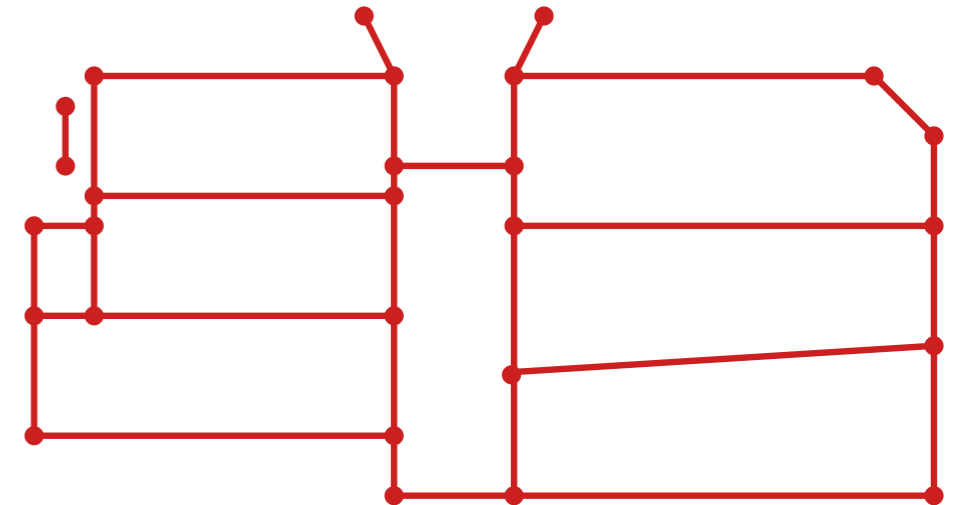
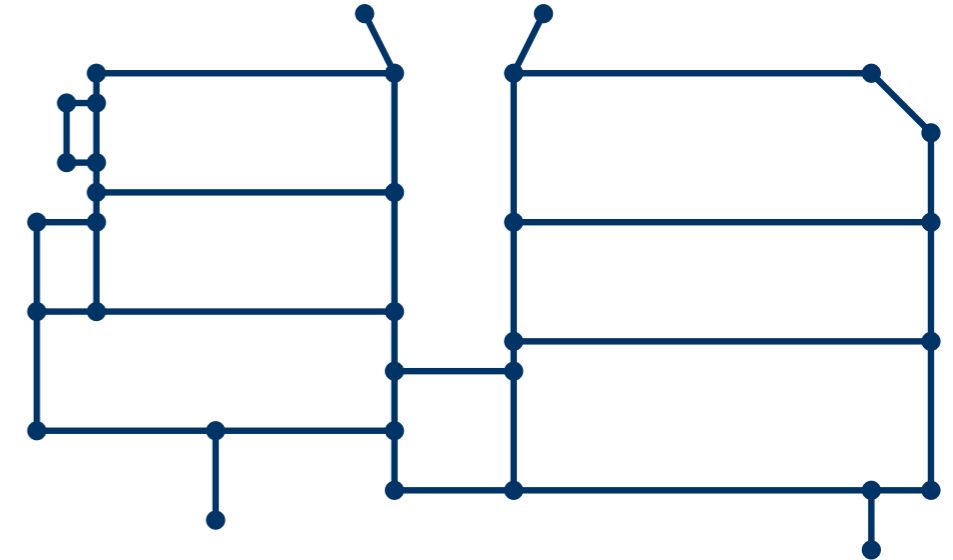
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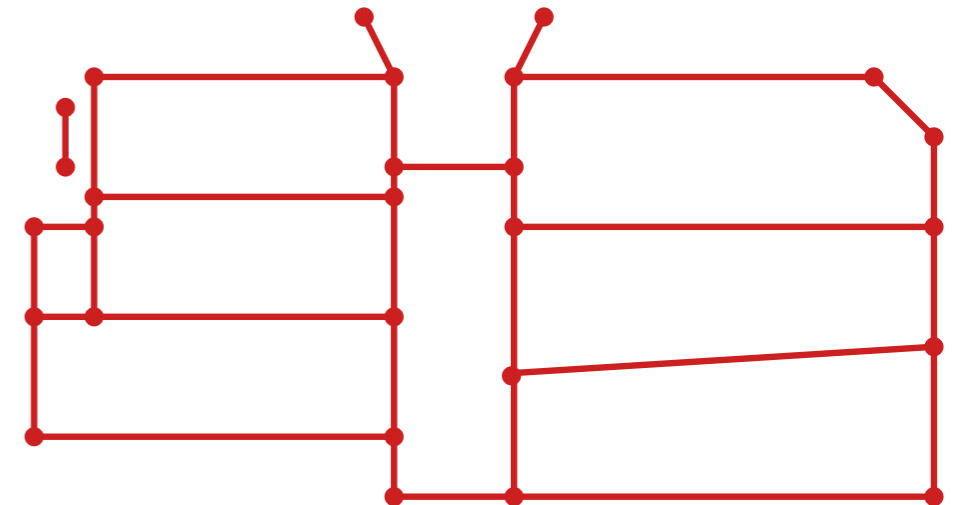
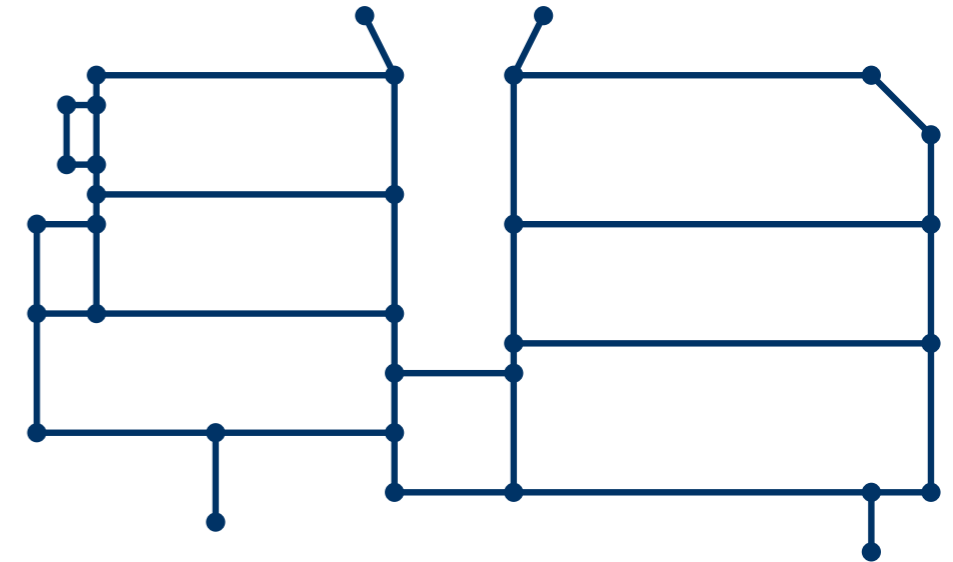
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We consider non-degenerate straight-line immersions

Given two representations of networks: how to compare them?

- Many approaches: edit distances, Fréchet distance, traversal based distances, LPH based distances,...
- Here: Weak Graph Distance due to Akitaya et al.



→ Akitaya et al.: Distance measures for embedded graphs, CGTA 95, 2021.



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1. Introduction
2. Hardness of deciding the weak graph distance
3. Crossing-rigid weak graph distances

## Recap: Weak Fréchet distance

Let  $s_1, s_2: [0, 1] \rightarrow \mathbb{R}^d$  be curves. Their *weak Fréchet distance* is defined by

$$\delta_{wF}(s_1, s_2) = \inf_{\substack{\alpha, \beta: [0, 1] \rightarrow [0, 1] \\ \text{continuous surjection}}} \max_{t \in [0, 1]} d(s_1(\alpha(t)), s_2(\beta(t)))$$

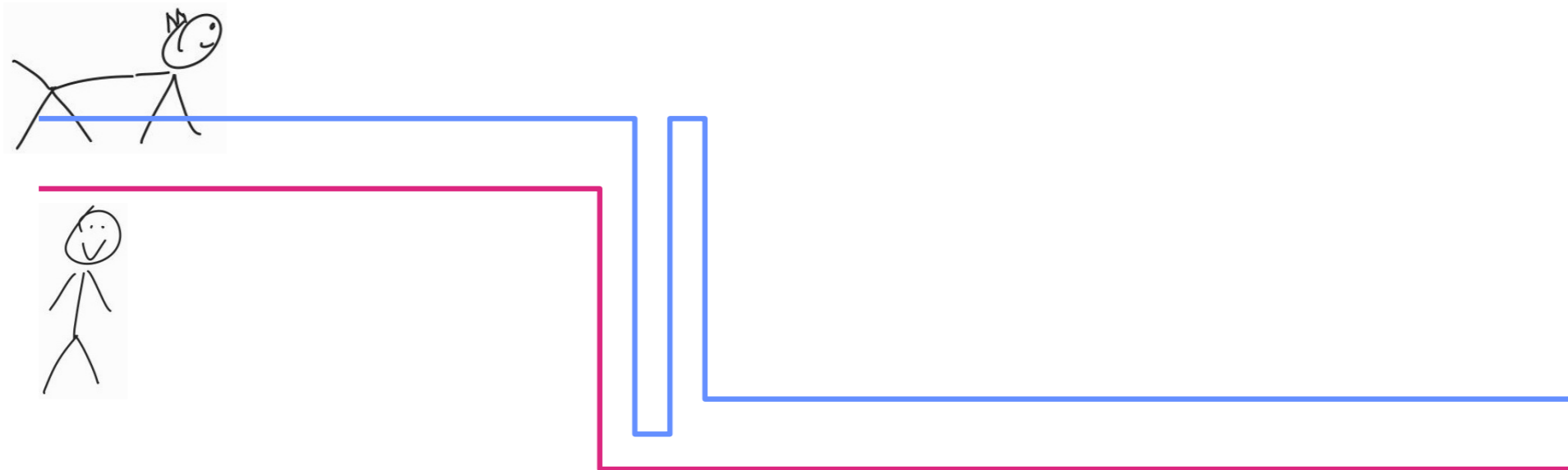
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Intuition: Person and dog:

Weak Fréchet distance equals shortest leash length *when both are allowed to backtrack along their trajectory*



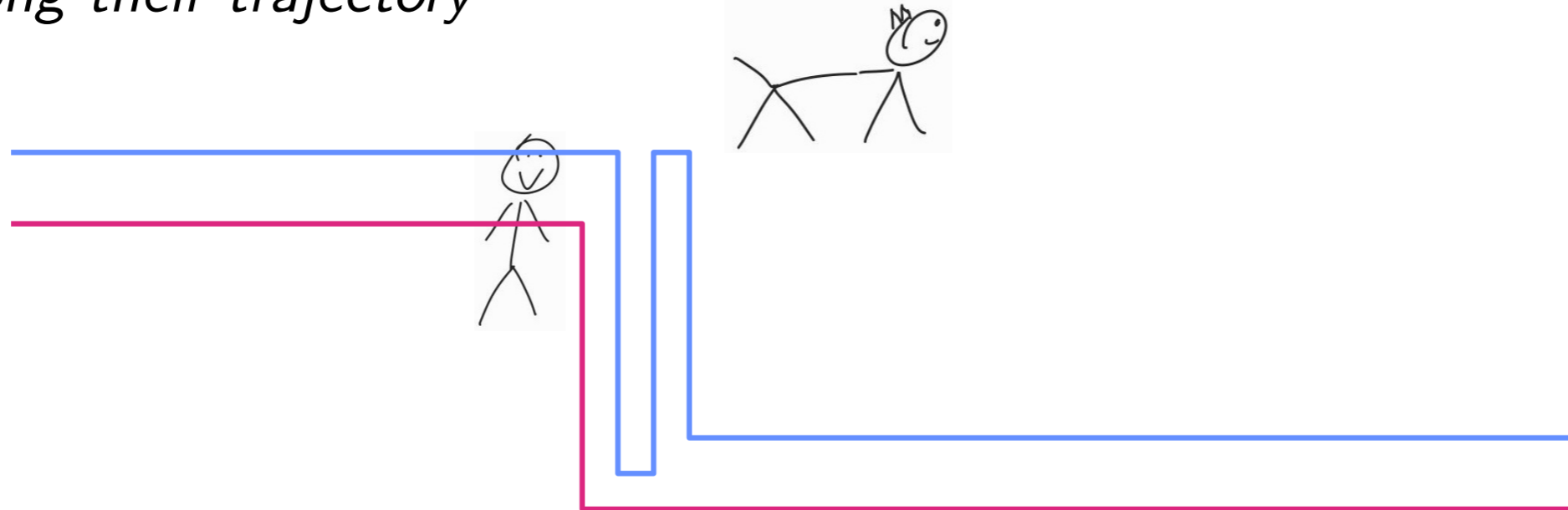
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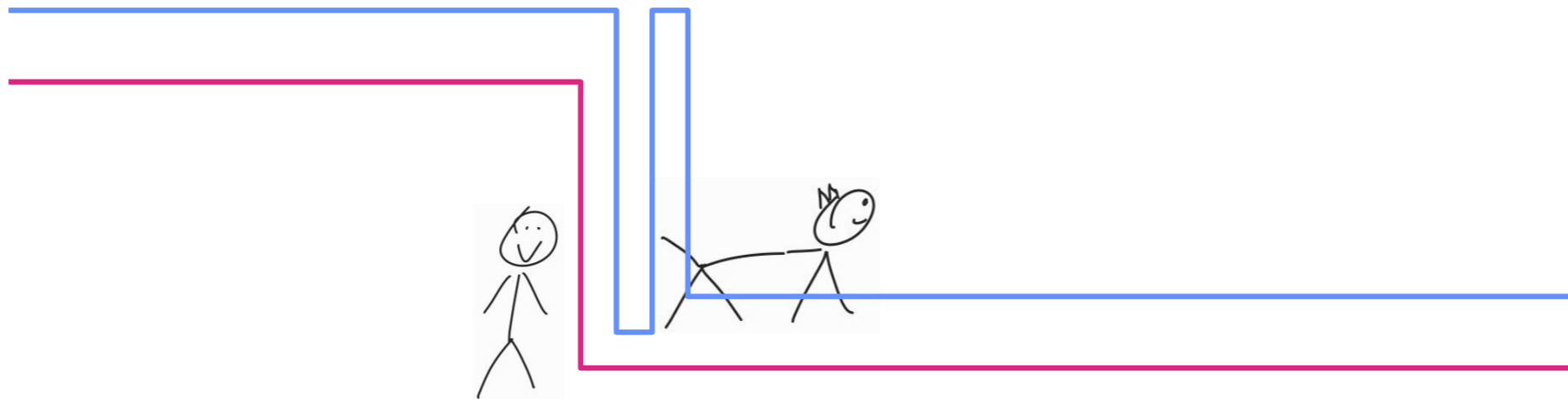
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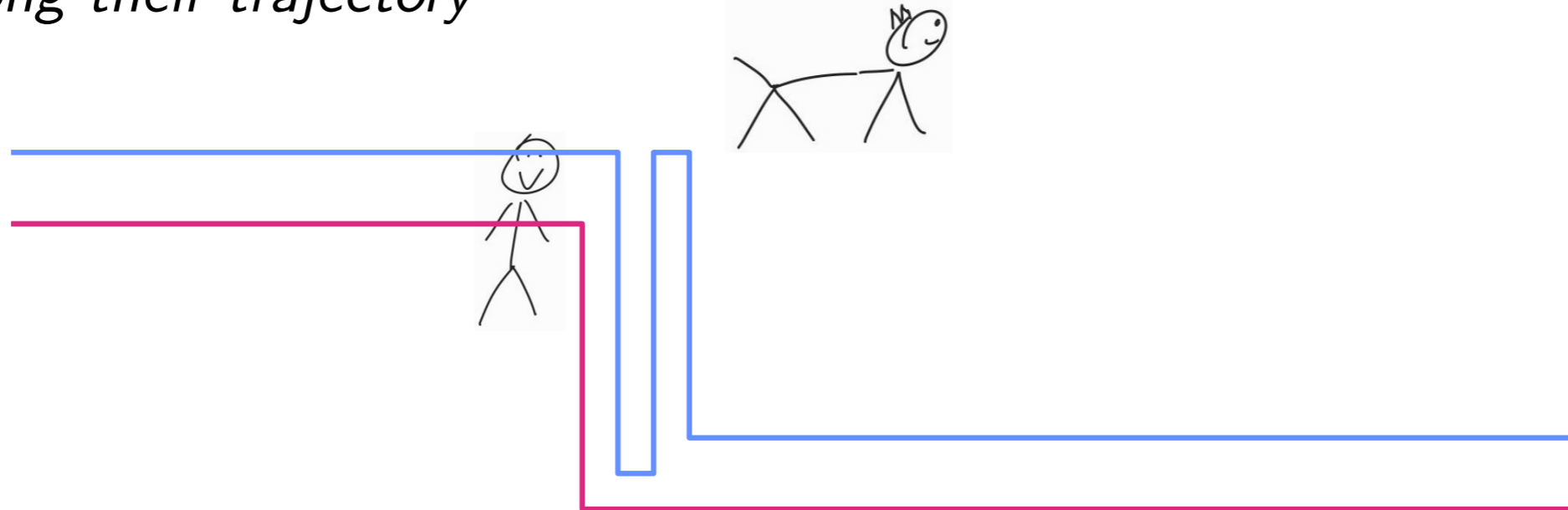
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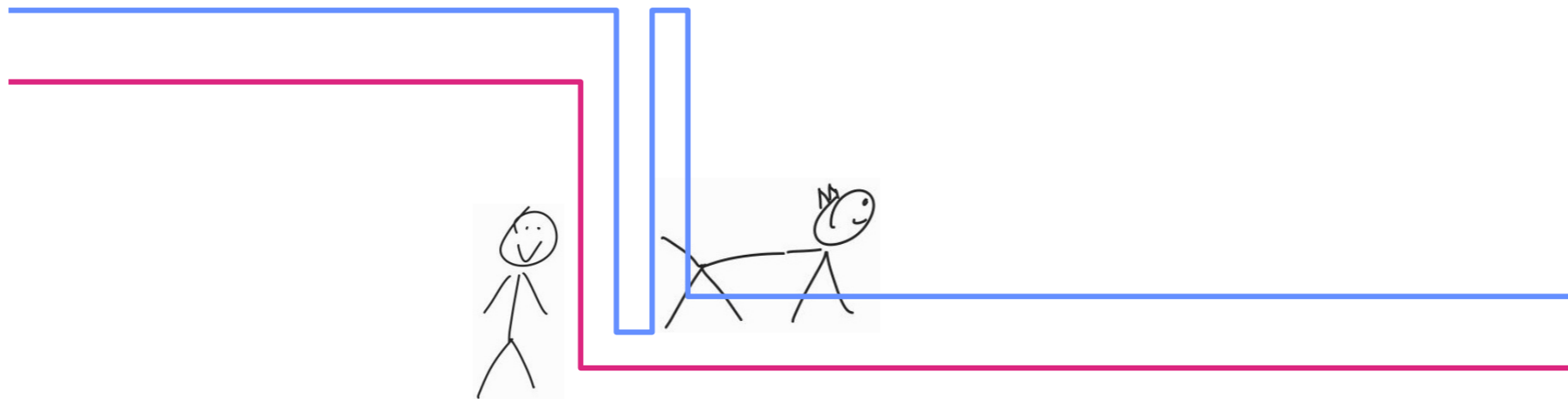
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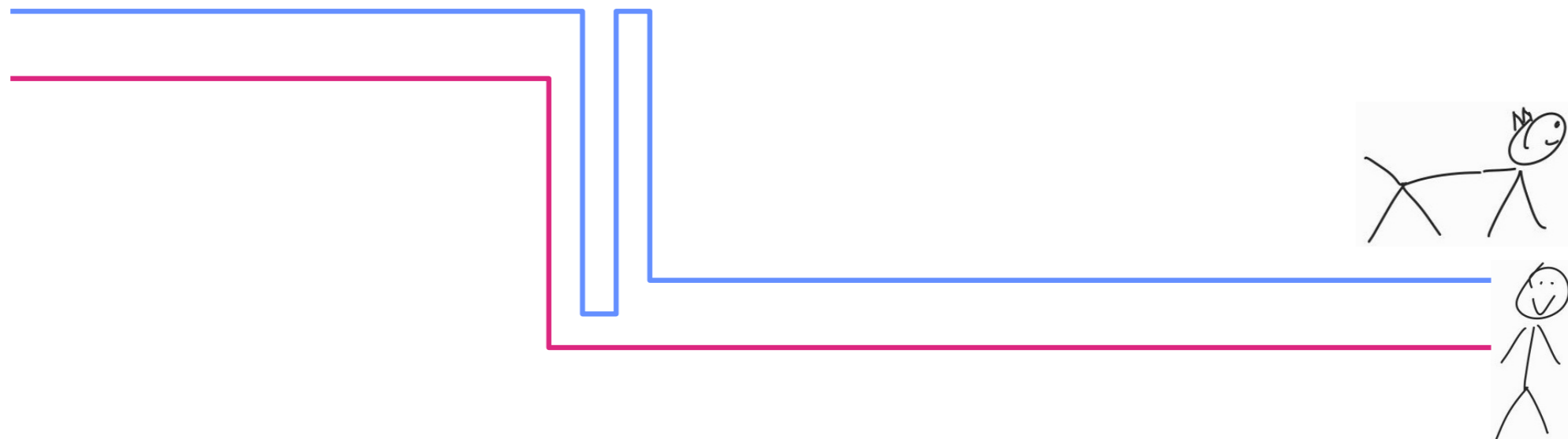
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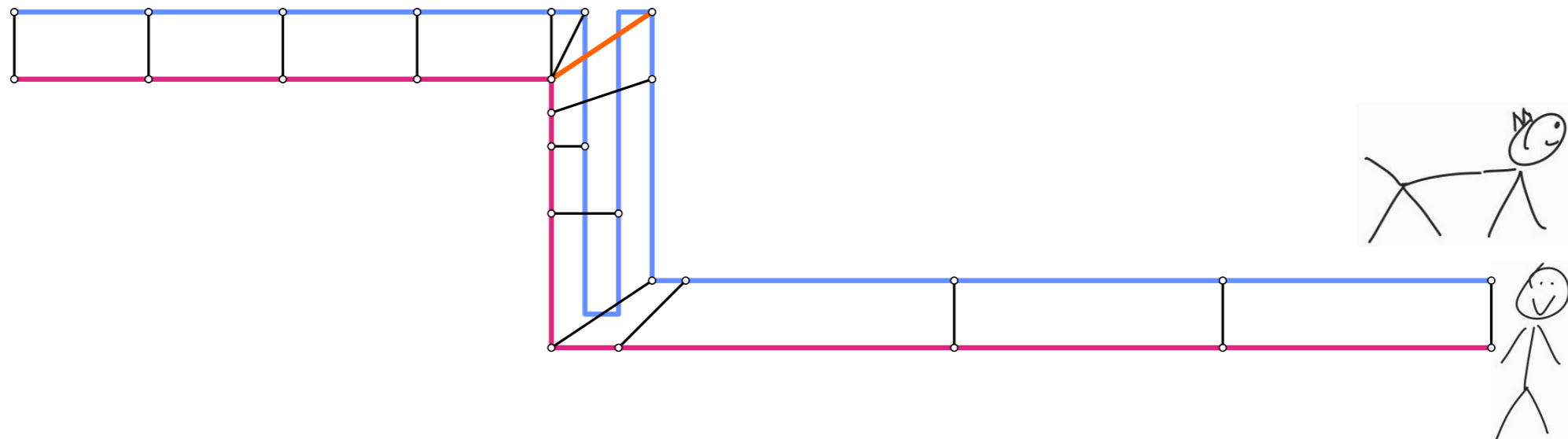
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# Weak graph distance

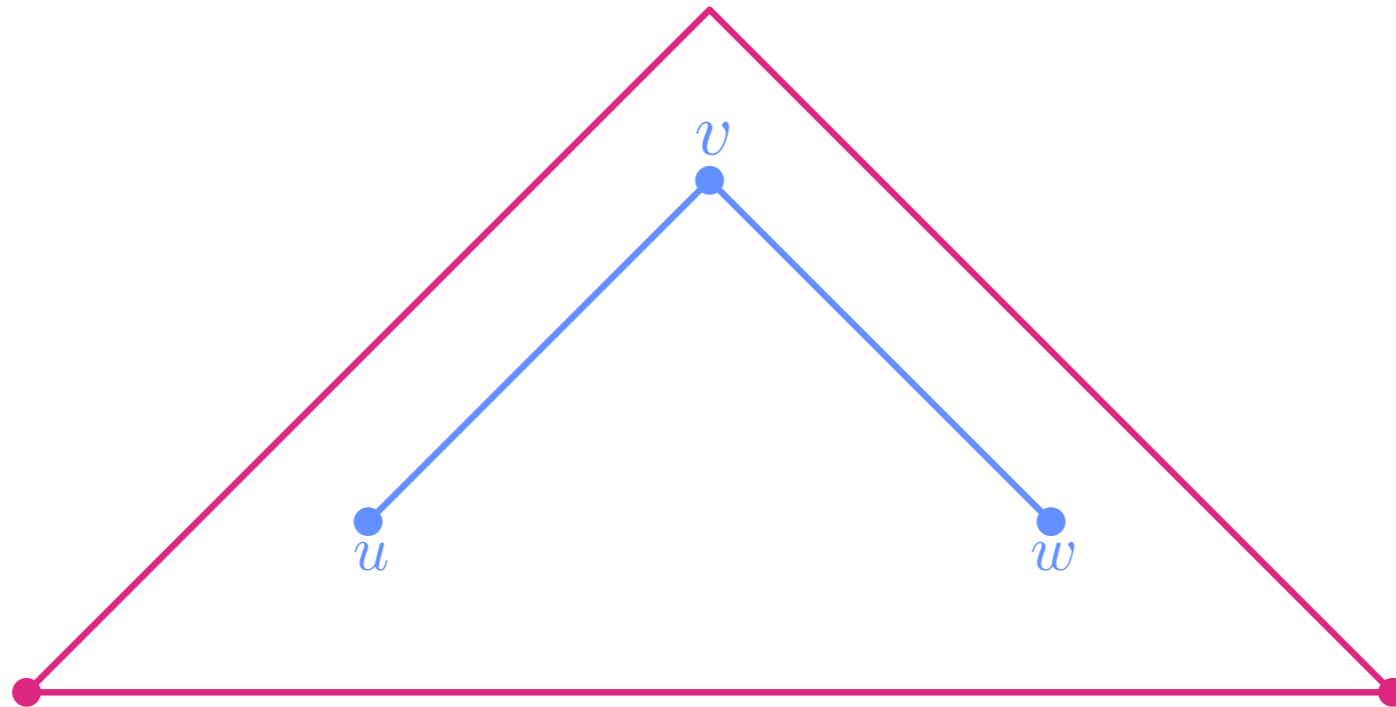
Let  $G_1, G_2$  be immersed graphs. A *graph mapping*  $s: G_1 \rightarrow G_2$  maps

- each vertex  $v$  of  $G_1$  to a point  $s(v)$  on an edge of  $G_2$
- each edge  $\{u, v\}$  of  $G_1$  to a simple path from  $s(u)$  to  $s(v)$  in  $G_2$

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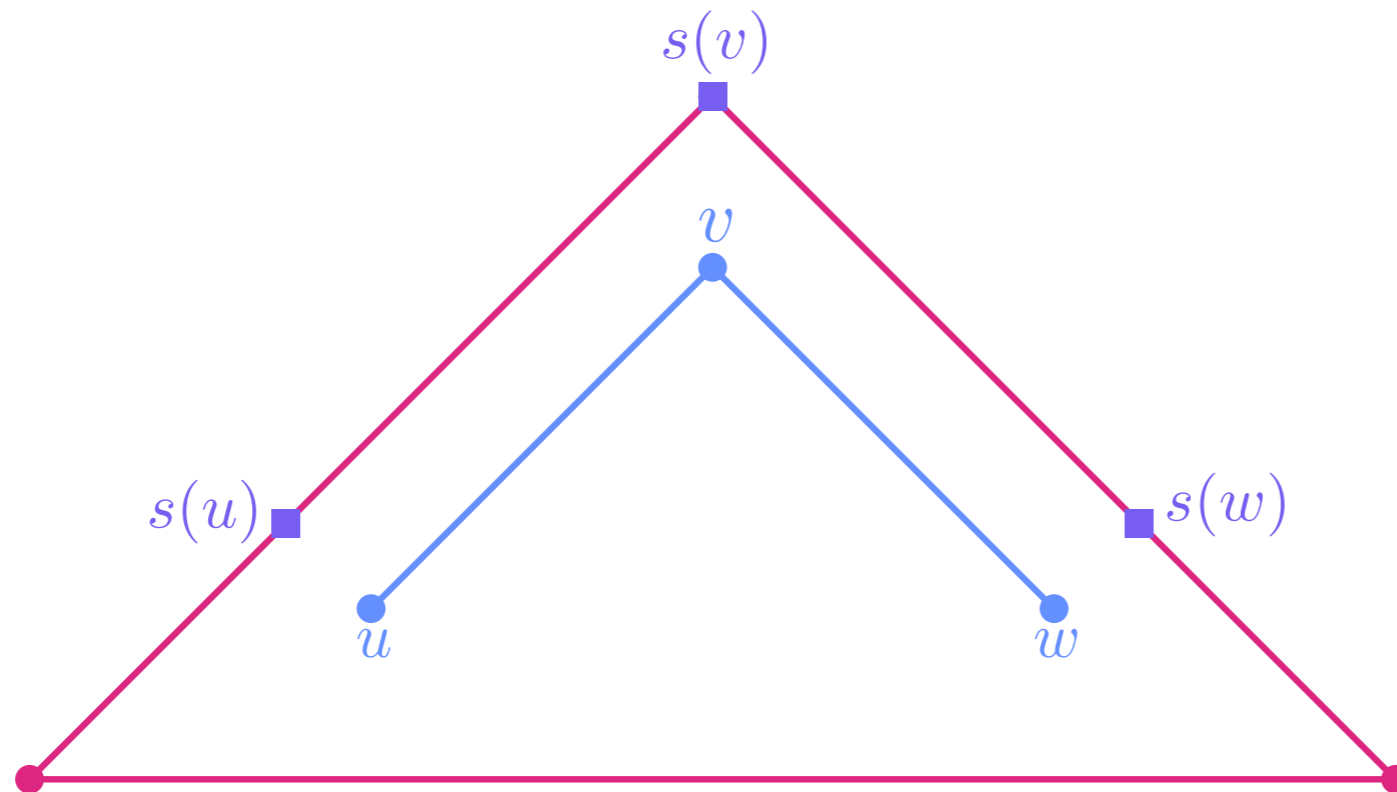
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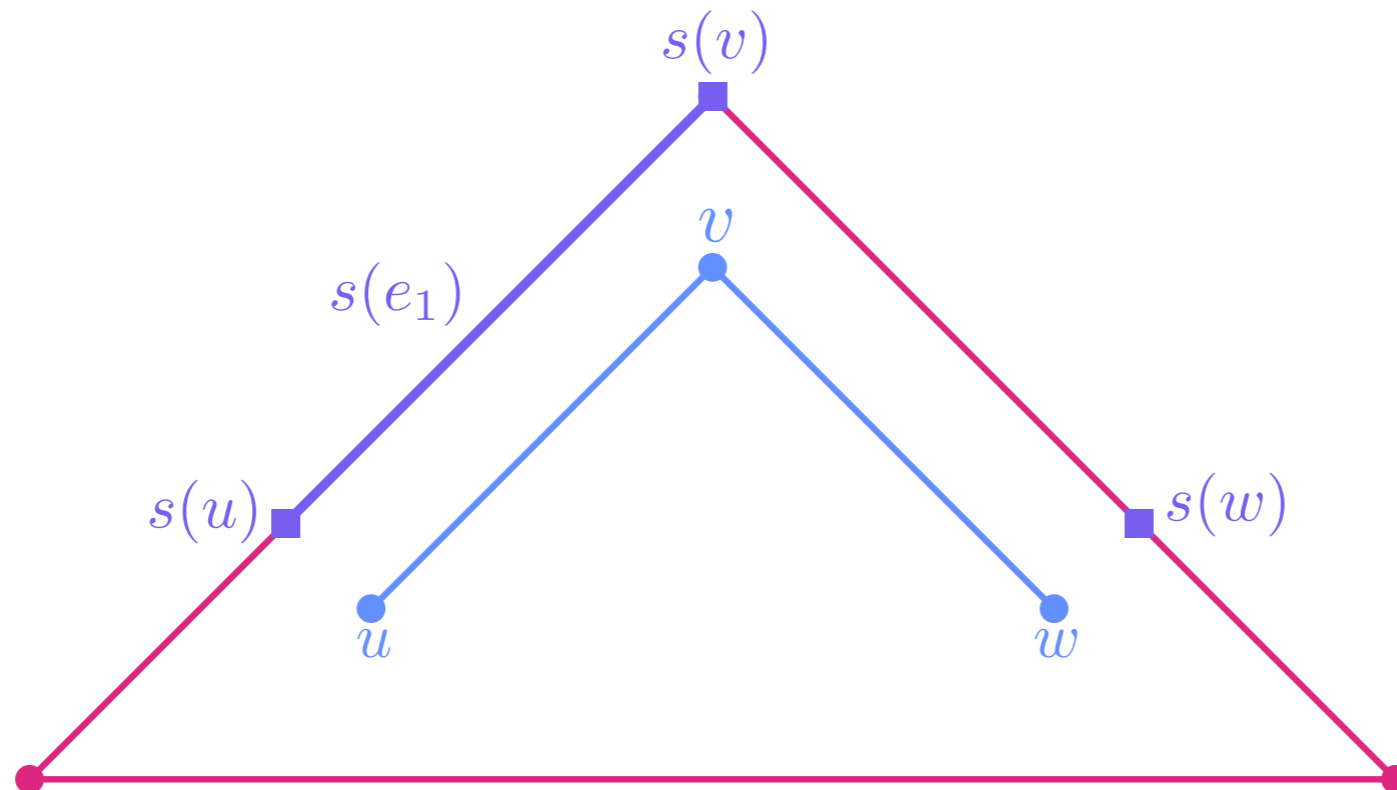
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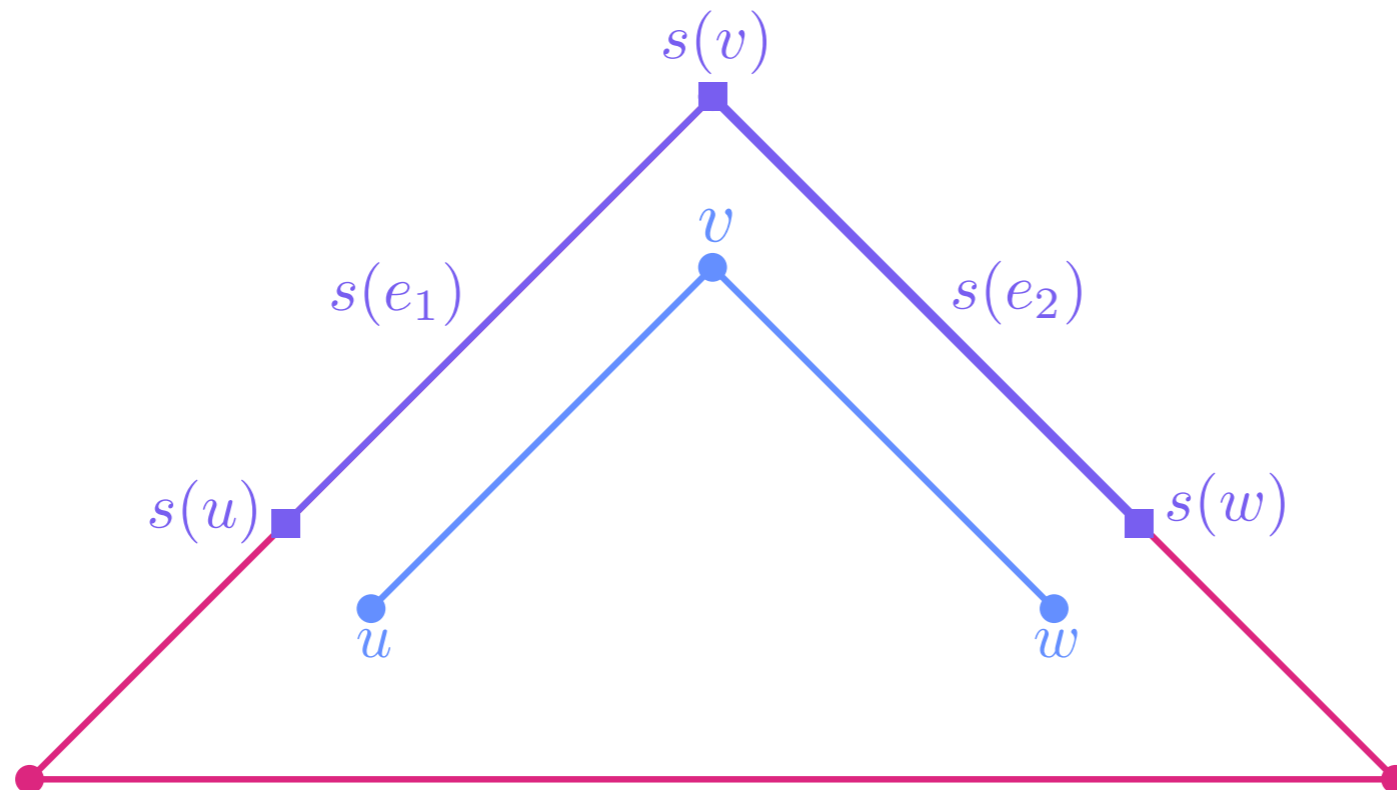
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The *directed weak graph distance* from  $G_1$  to  $G_2$  is defined as

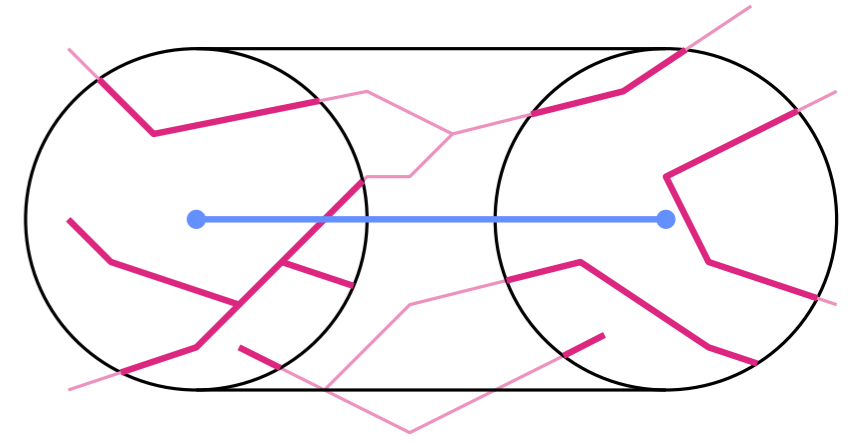
$$\vec{\delta}_{wG} = \min_{\substack{s: G_1 \rightarrow G_2 \\ \text{graph mapping}}} \max_{e \in E(G_1)} \delta_{wF}(e, s(e))$$

graph mapping interpreted as curves

Undirected version:  $\delta_{wG}(G_1, G_2) = \max\{\vec{\delta}_{wG}(G_1, G_2), \vec{\delta}_{wG}(G_2, G_1)\}$

General decision algorithm due to Akitaya et al.

*Vertex placement* of  $v \in V(G_1)$ : connected component of  $G_2 \cap B_\varepsilon(v)$



**RUB**

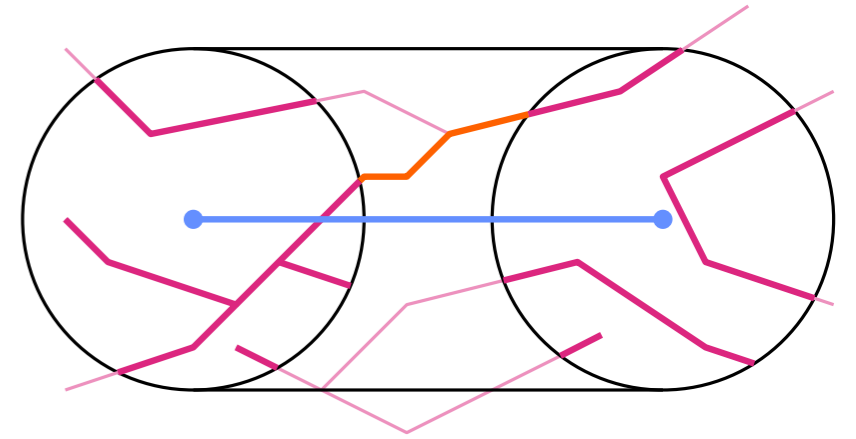


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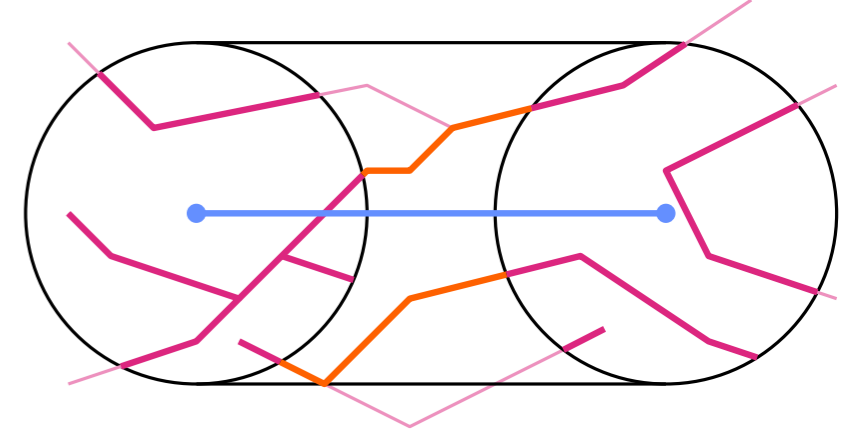


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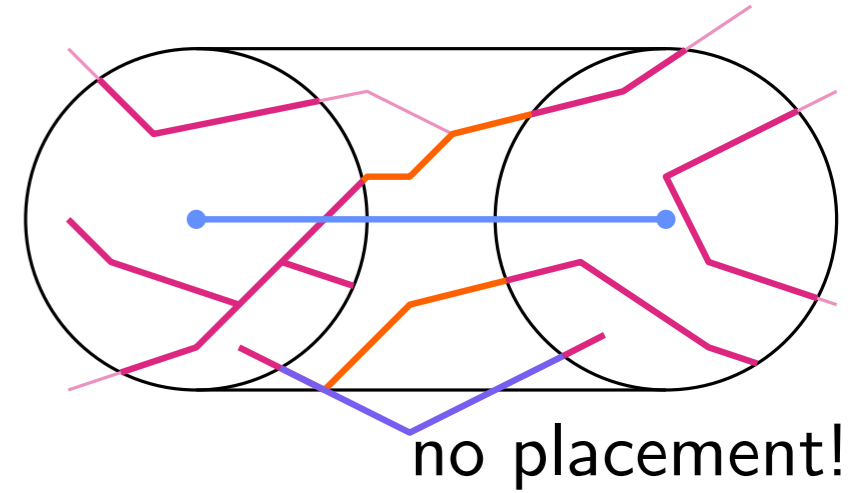
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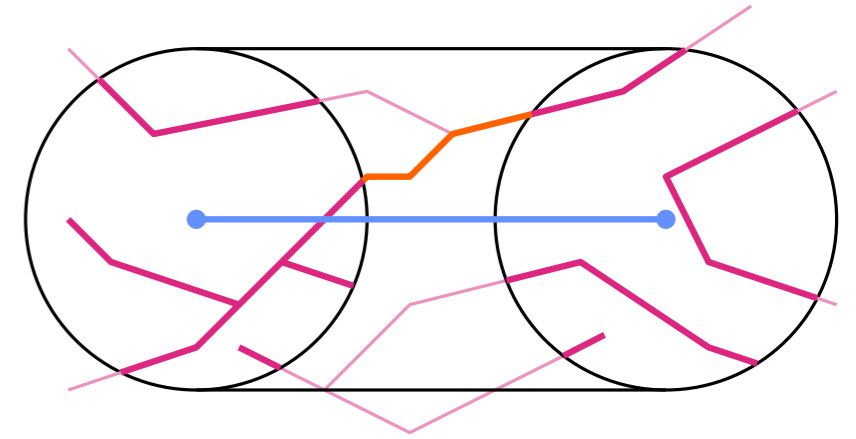
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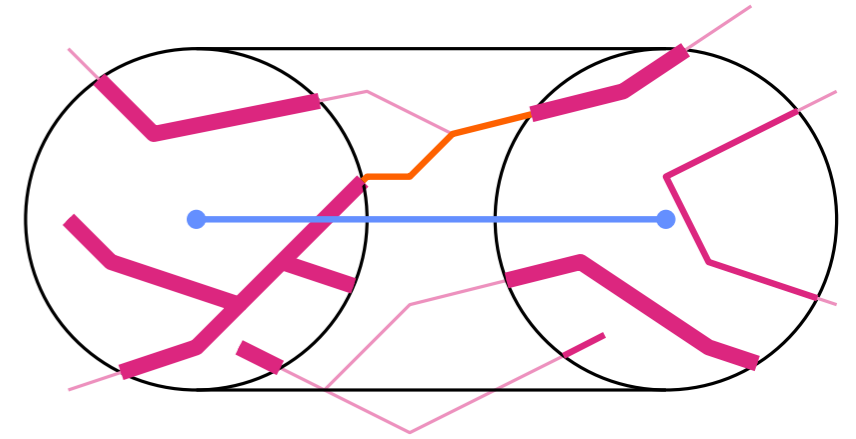
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# General decision algorithm due to Akitaya et al.

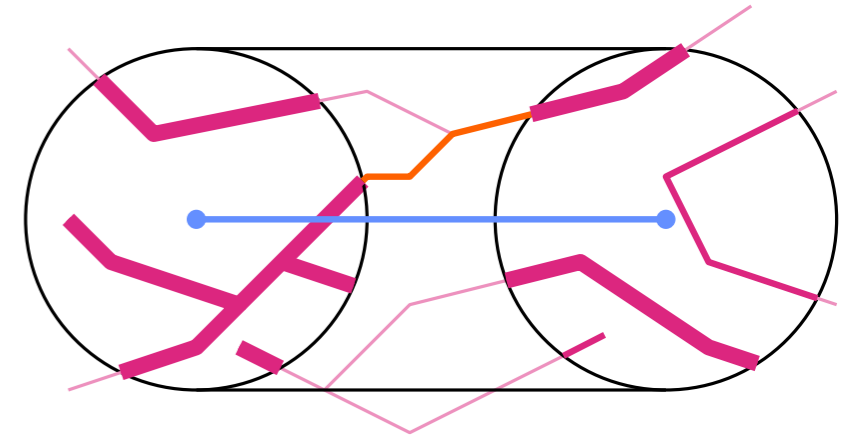
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3. Delete invalid placements.
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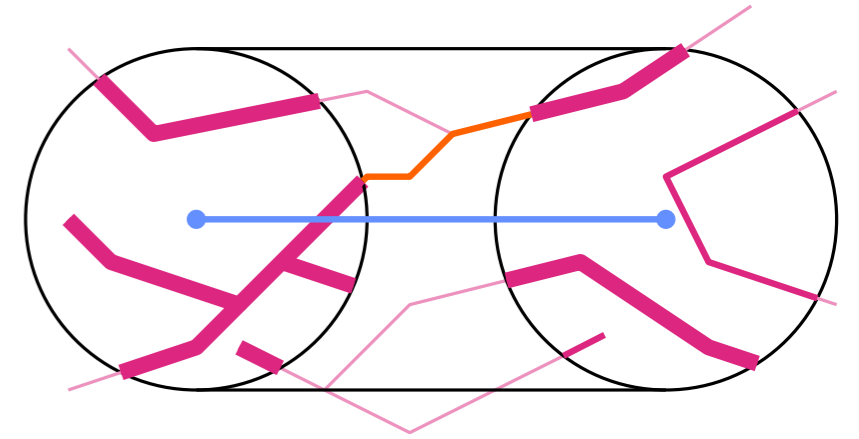
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General decision algorithm:

1. Compute vertex placements.
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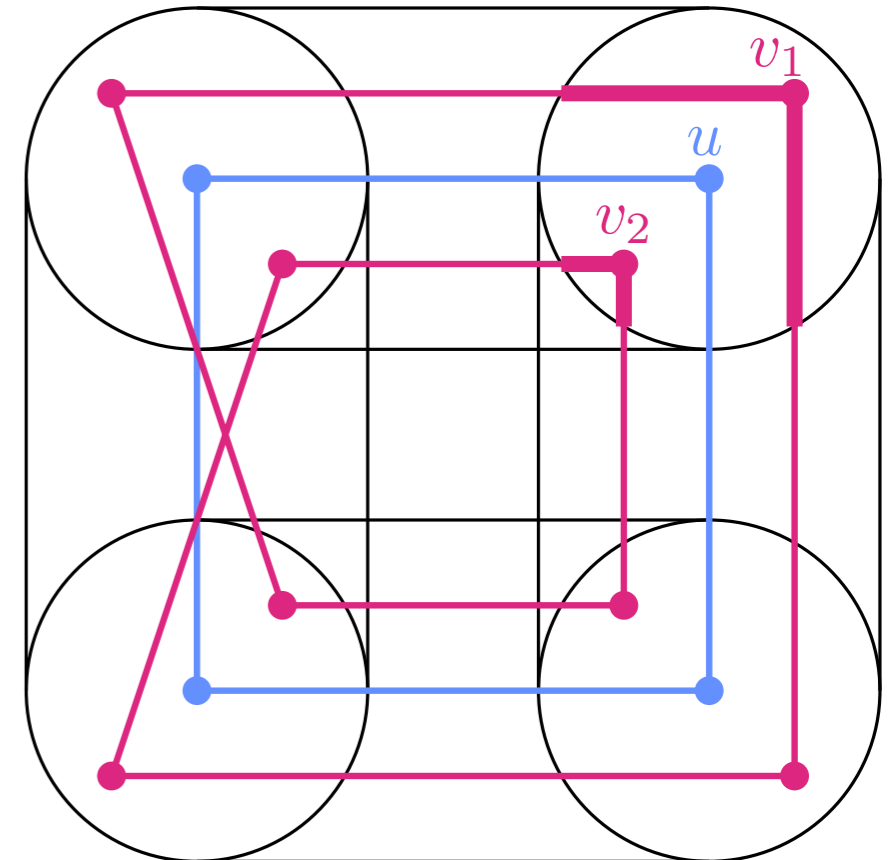
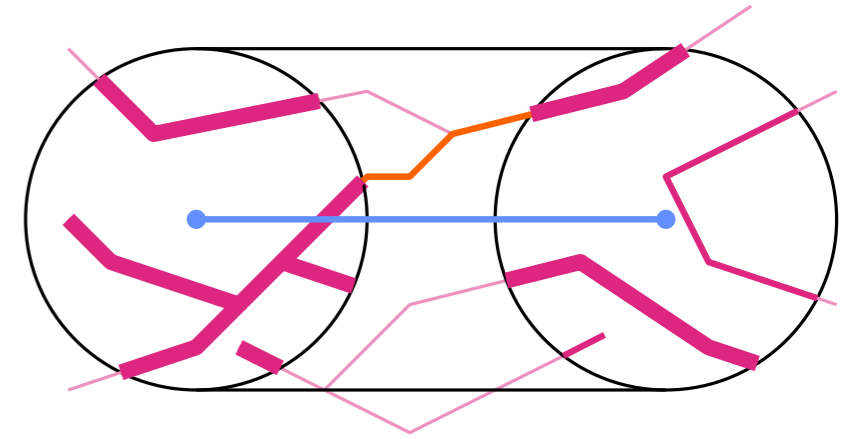
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Hardness of deciding  $\vec{\delta}_{wG}(G_1, G_2) \leq \varepsilon$  if  $G_1$  is plane

**Theorem:** Deciding whether  $\vec{\delta}_{wG}(G_1, G_2) \leq \varepsilon$  is NP-complete even if  $G_1$  is plane and  $G_2$  is immersed in  $\mathbb{R}^2$ .

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*Sketch of the proof by reduction from PLANAR 3COL:*

Let  $G = (V, E)$  be the (planar) input graph

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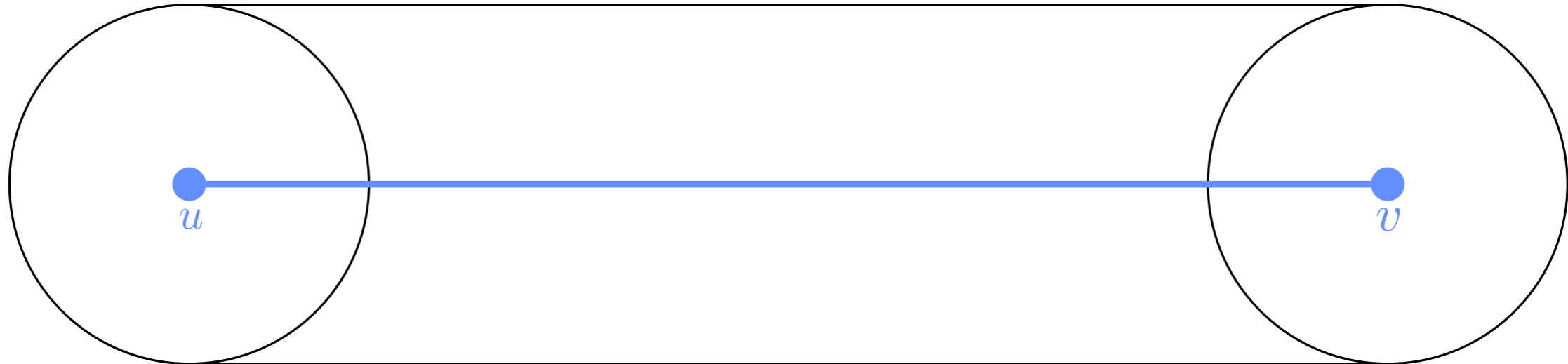
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2. Choose  $\varepsilon$  s.t. all  $\varepsilon$ -balls and tubes are separated



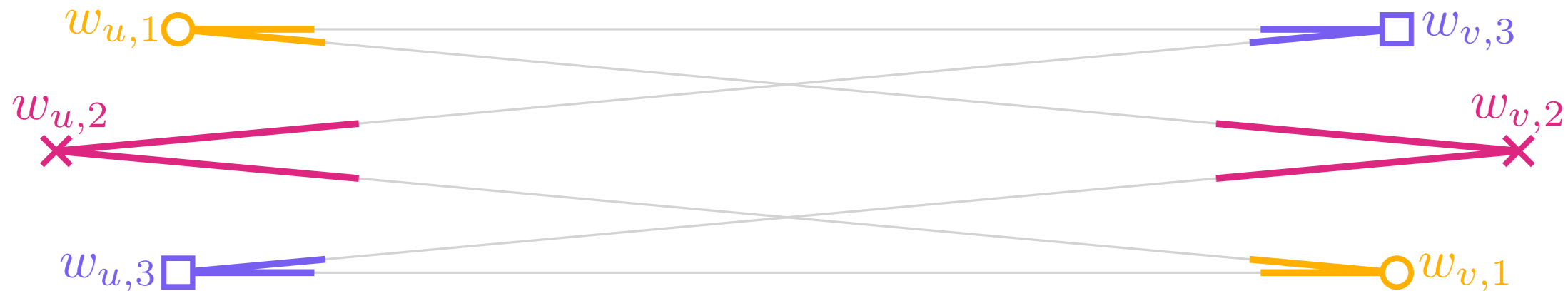
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3. Construct  $G_c$  with
- vertices  $w_{u,i}$  for  $u \in V, i \in [3]$
  - edges  $\{w_{u,i}, w_{v,j}\}$  for  $\{u, v\} \in E, i \neq j$



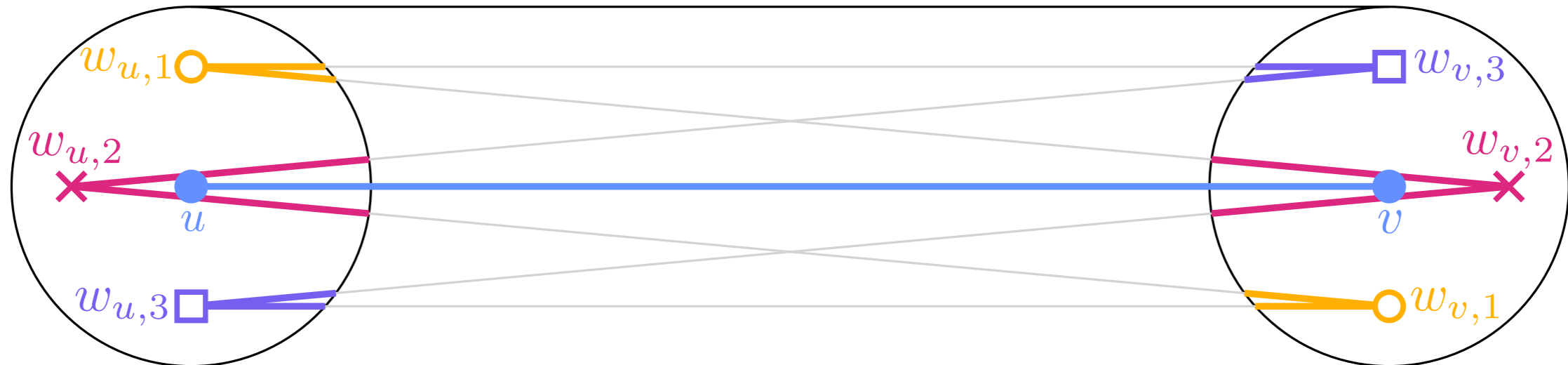
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4. Immerse  $G_c$  s.t.  $w_{u,i}$  lies in  $B_\varepsilon(u)$



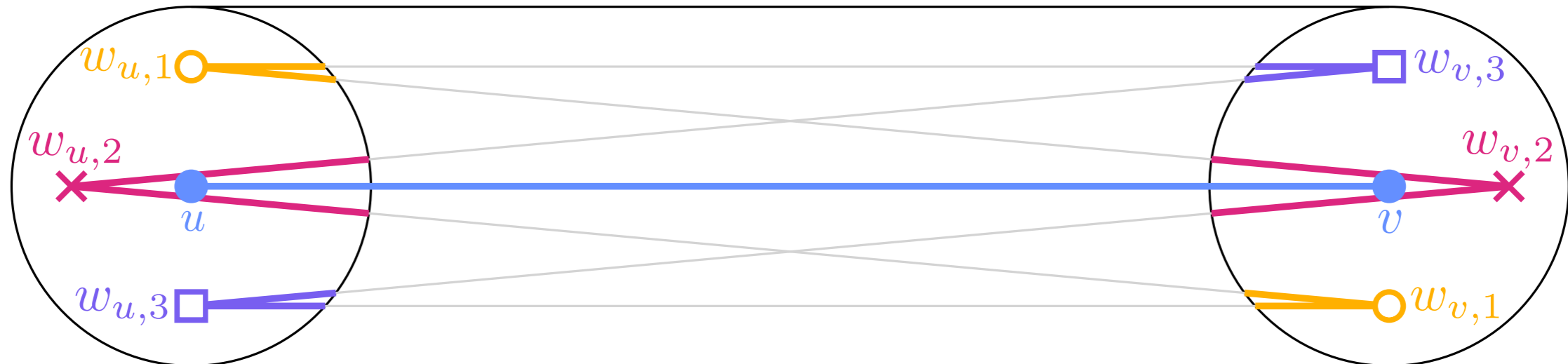
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Idea: Vertex placements  $\leftrightarrow$  colors



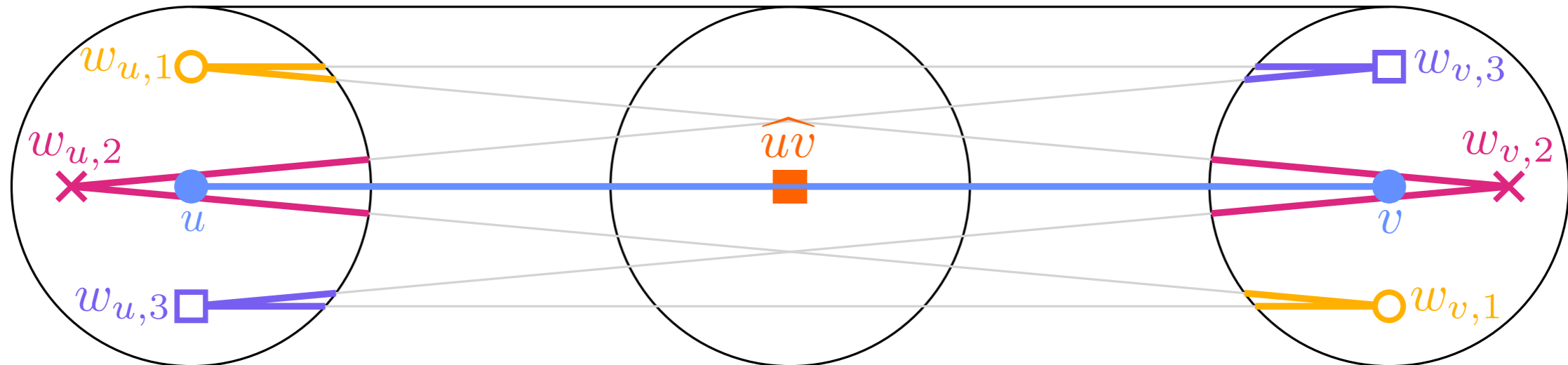
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1. Construct a crossing-free embedding of  $G$  and insert a vertex  $\widehat{uv}$  in the middle of each edge  $\{u, v\} \rightarrow G_p$





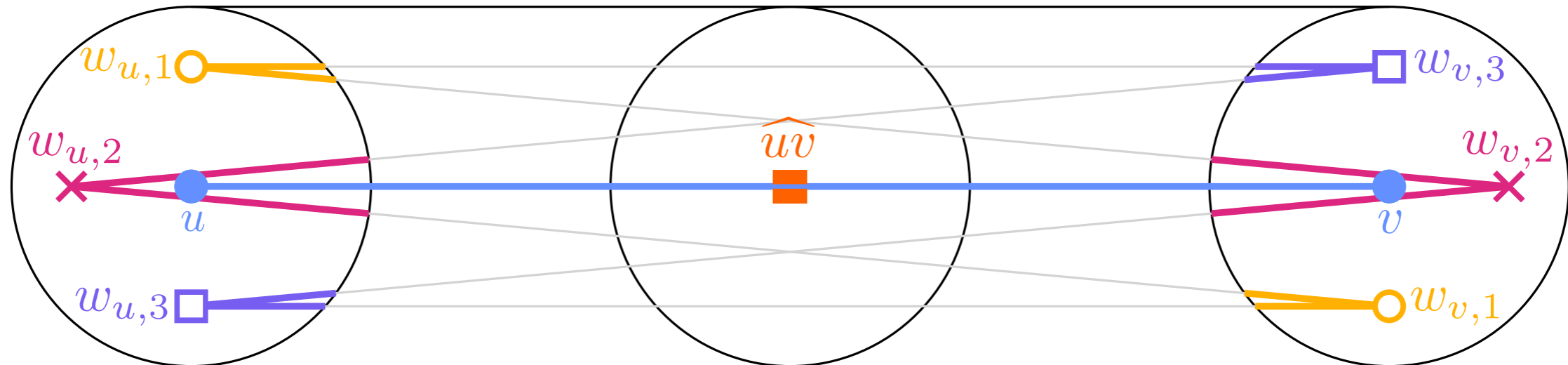
# Hardness of deciding $\vec{\delta}_{wG}(G_1, G_2) \leq \varepsilon$ if $G_1$ is plane

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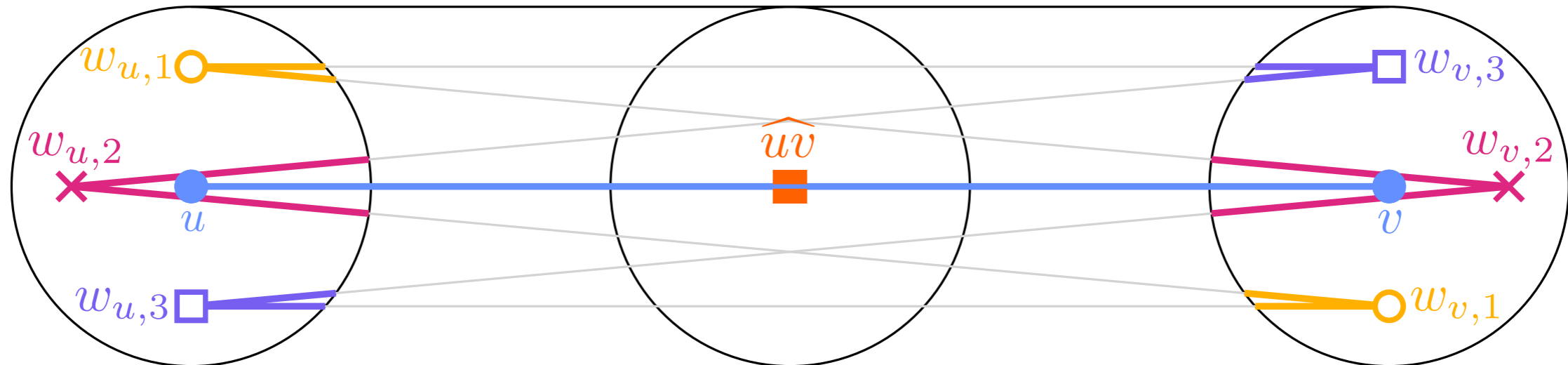
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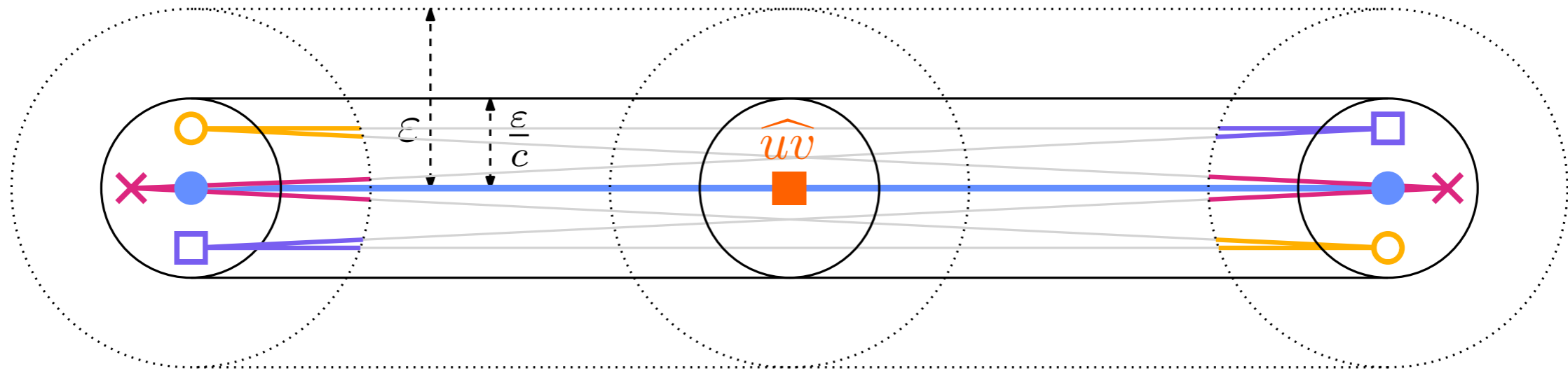
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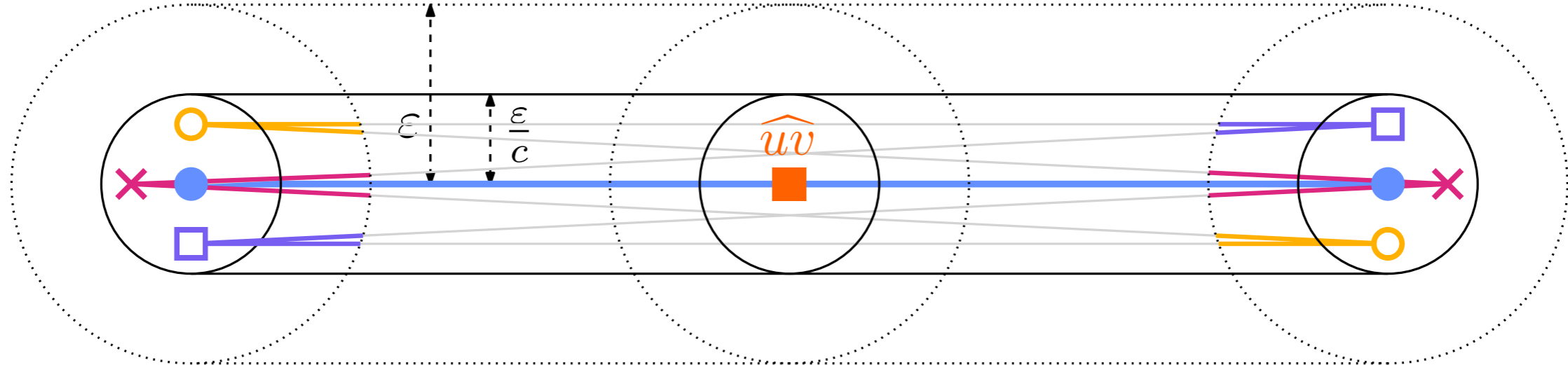
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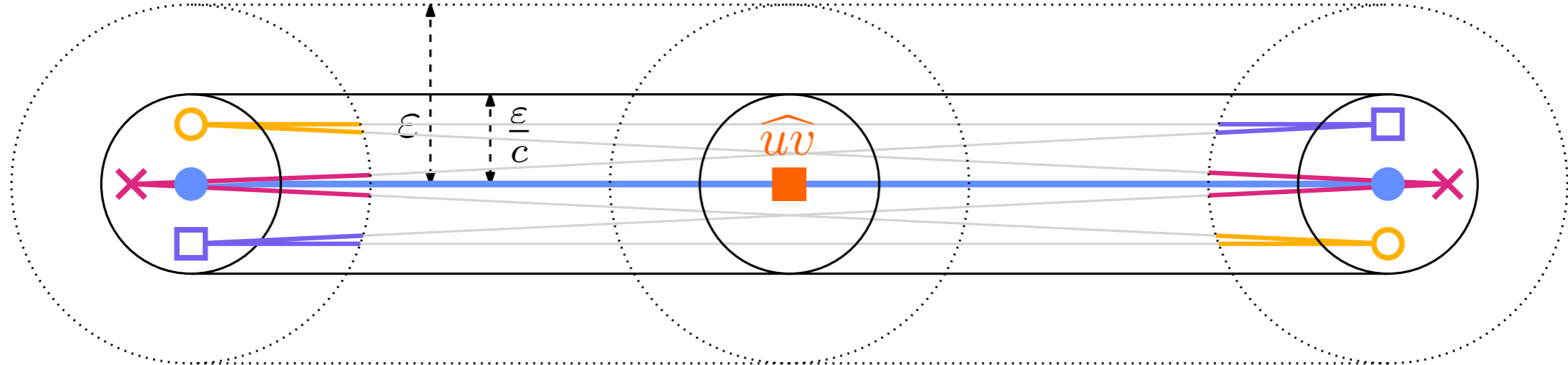
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Negative case: Lower bound of  $\epsilon$  remains intact



## Hardness of the embedded case in $\mathbb{R}^d$ , $d \geq 3$

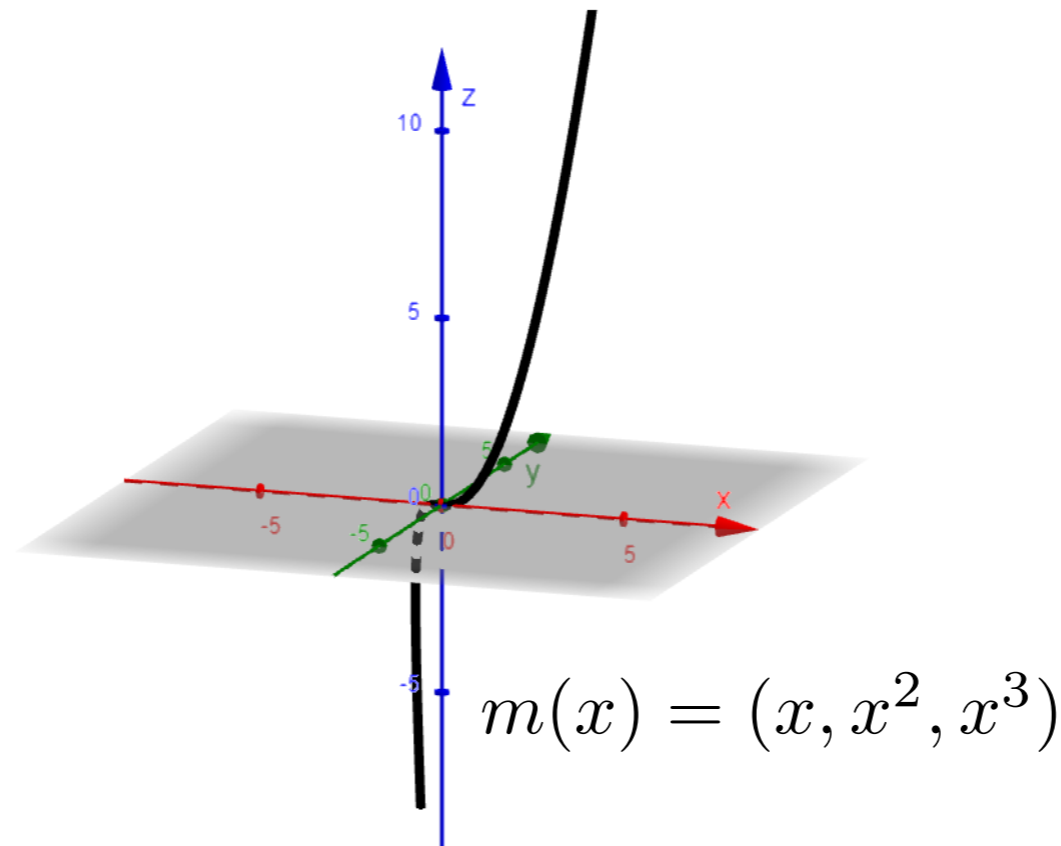
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(Up to details,) embed on the 3-dim moment curve instead





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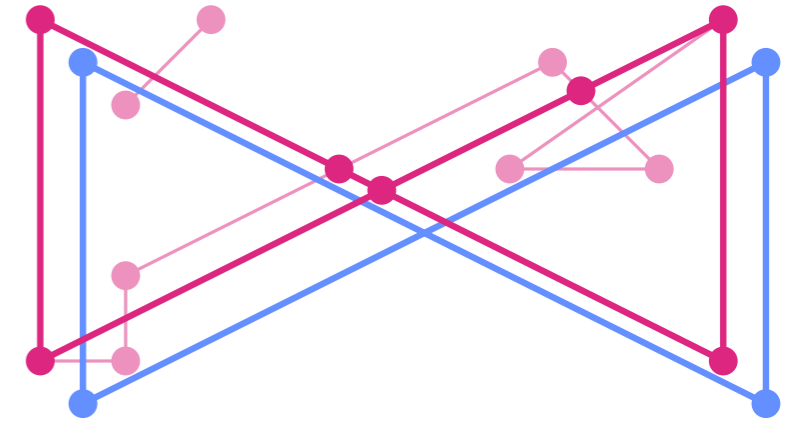
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*(Informal) Definition:* A graph mapping that maps each edge with  $n$  crossings to a path containing

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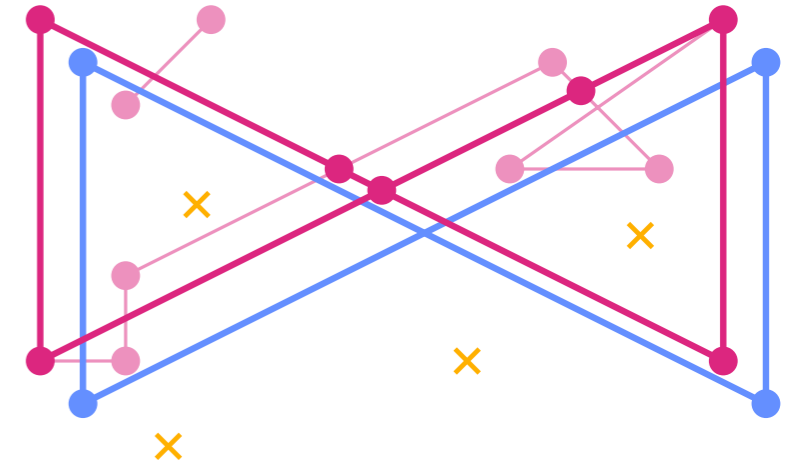


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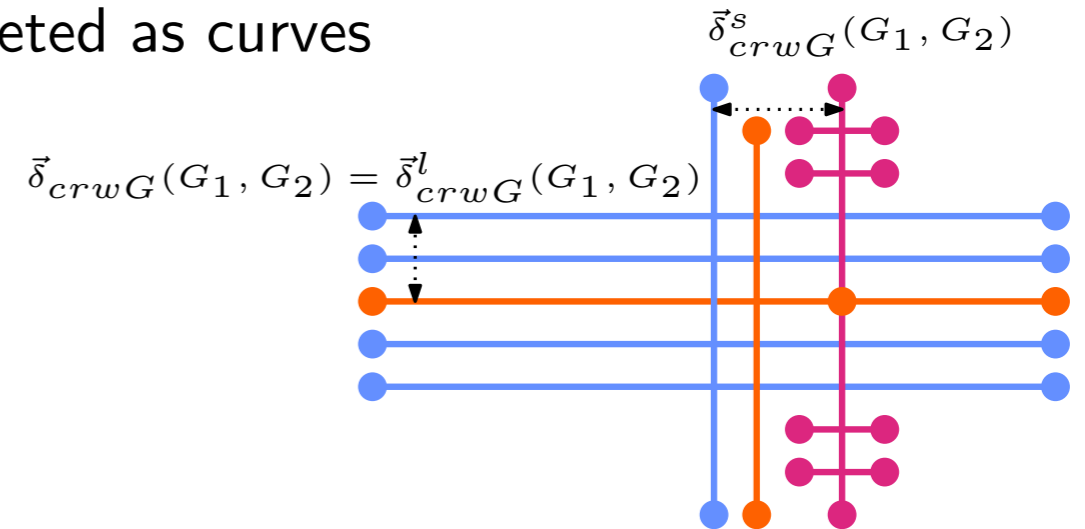


The *directed crossing-rigid weak graph distance* is defined as

$$\vec{\delta}_{crwG}(G_1, G_2) = \inf_{\substack{s:G_1 \rightarrow G_2 \\ \text{crossing-rigid}}} \max_{e \in E} \delta_{wF}(e, s(e))$$

interpreted as curves

Analogous:  $\vec{\delta}_{crwG}^l, \vec{\delta}_{crwG}^s$



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*Caveat:* since placements are no longer compact, there might be no  $\varepsilon$ -placement but still  $\delta_{crwG}^{(l)/(r)}(G_1, G_2) = \varepsilon$

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