

# Snapping Graph Drawings to the Grid Optimally

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  - Fixed-precision representation (e.g. integer coordinates)
  - Geometric similarity (no large vertex movements)
  - Topological similarity (equivalence up to the collapsing of features)

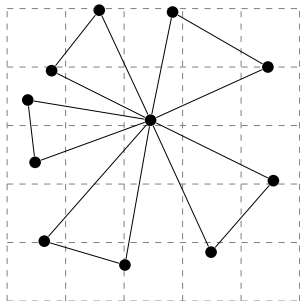
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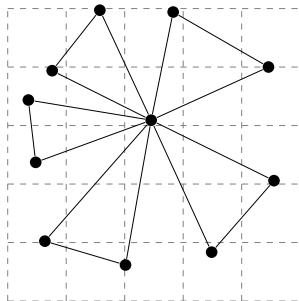
## Our question:

What about topological **equivalence**?

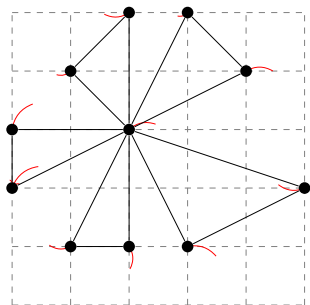
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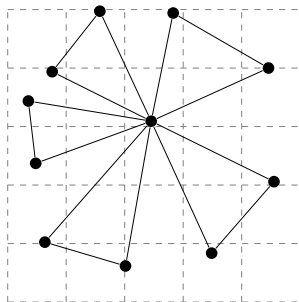
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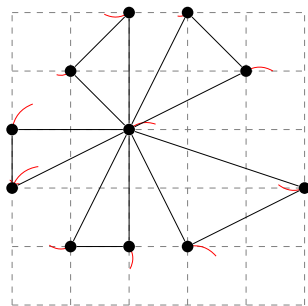
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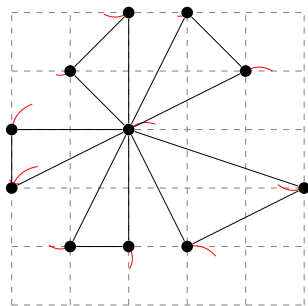
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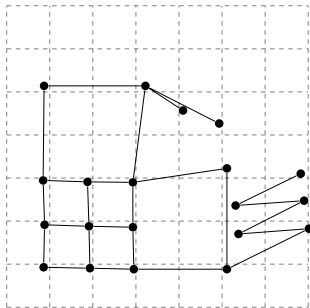
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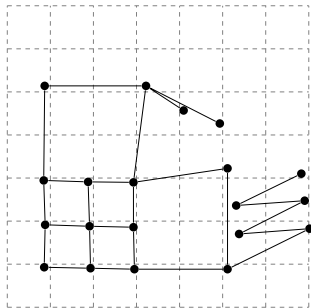
- Snap-rounding already is topologically equivalent.



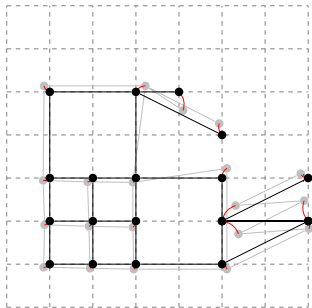
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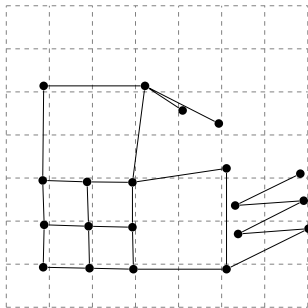
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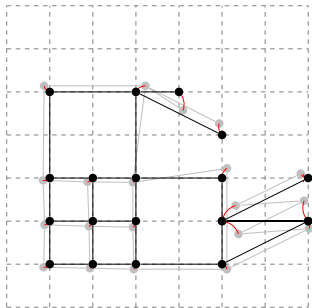
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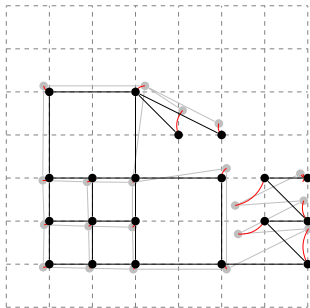
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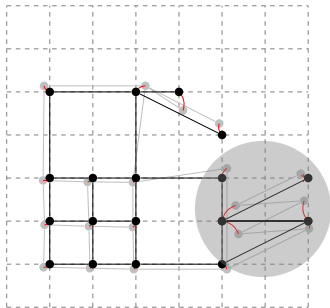
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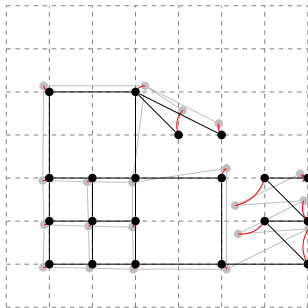
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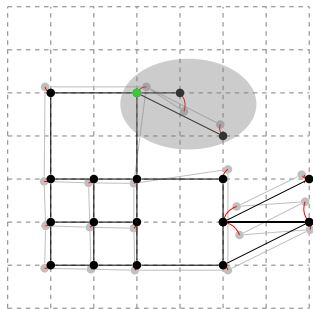


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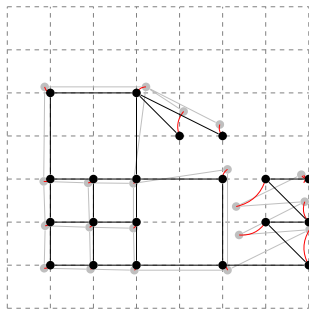


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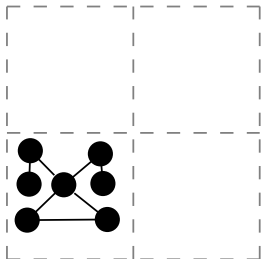


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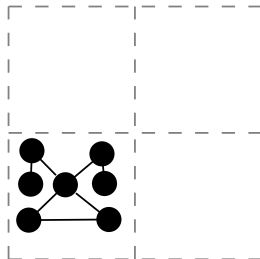


- Snap-rounding alters incidences and forces edges to collapse.
- Rounding to the nearest grid point changes the embedding of the upper-left **vertex**.

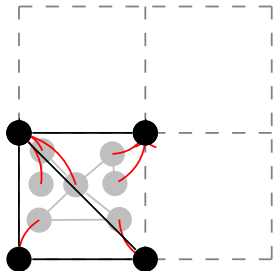
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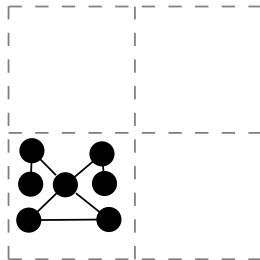
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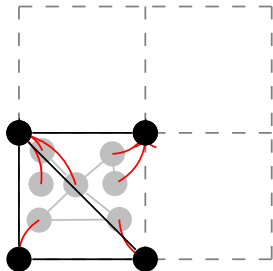
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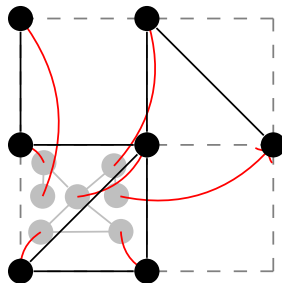
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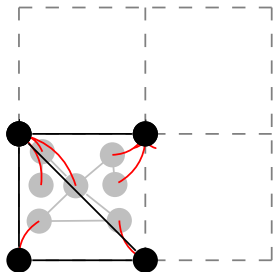


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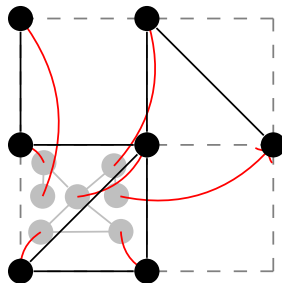




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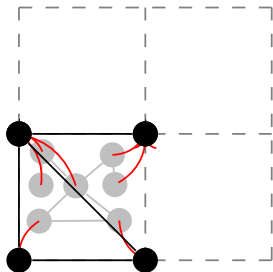


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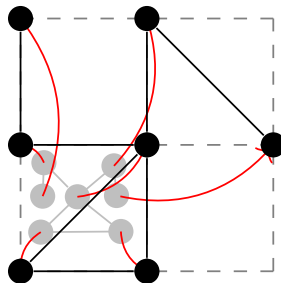


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- “Rounding” dense structures with no features collapsing is closely related to creating minimum-area drawings.

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- Movement is measured in Manhattan-distance.

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- Experimental Evaluation

The  $\mathcal{NP}$ -hardness proof

- We reduce from PLANAR MONOTONE 3SAT.

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- For reduction, consider a decision variant:

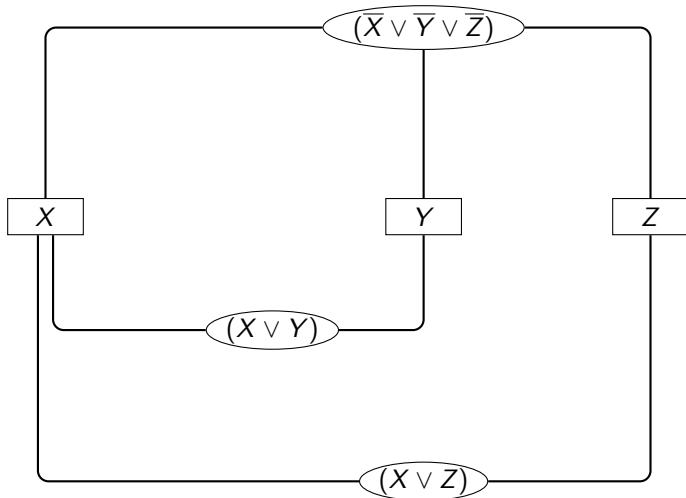
## Problem (Cost-bound TOPOLOGICALLY SAFE SNAPPING)

*Graph  $G = (V, E)$  with given embedding,  
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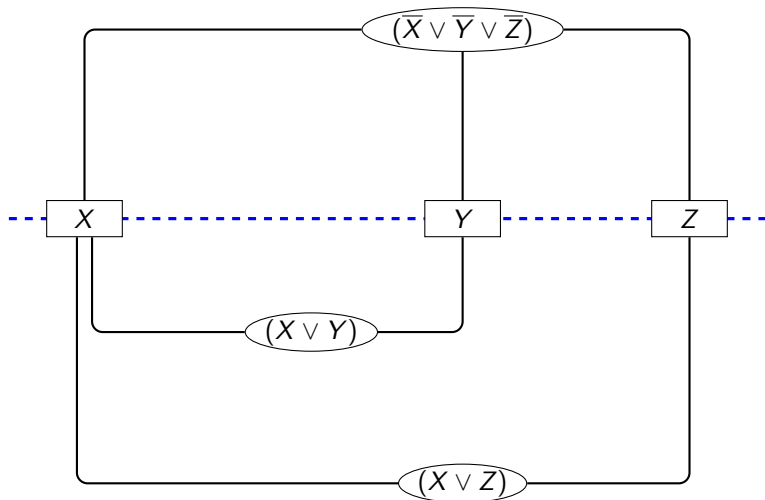
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*Can  $G$  be rounded to integer coordinates within  $B$ , preserving the given embedding with total movement of  $c_{\min}$ ?*

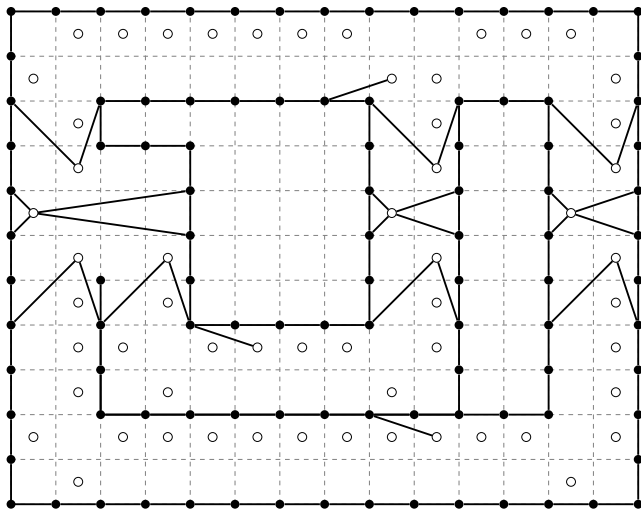
# PM3SAT-formula



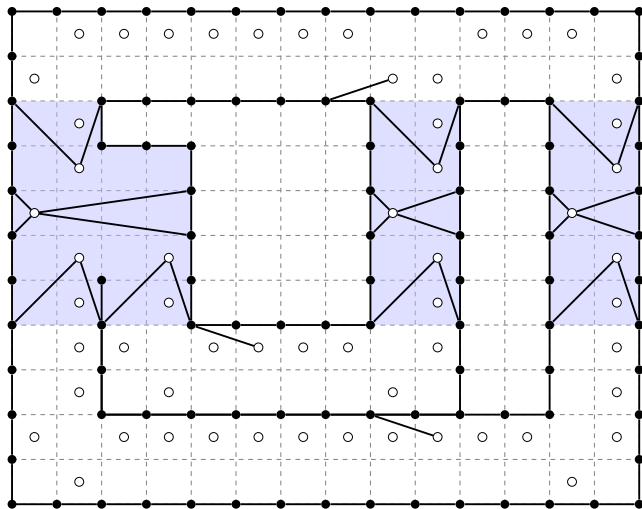
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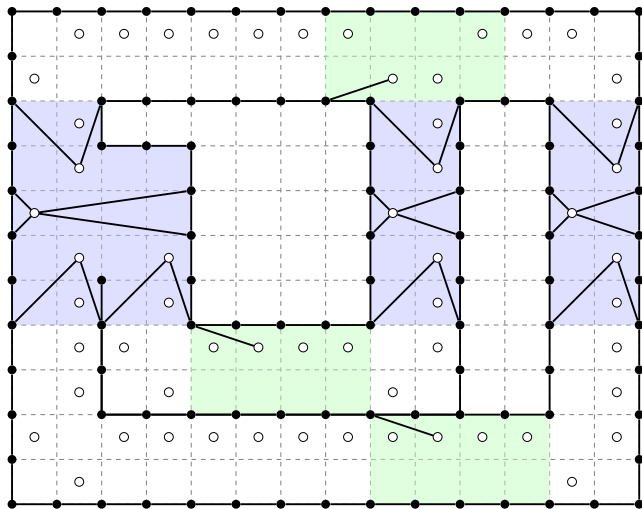


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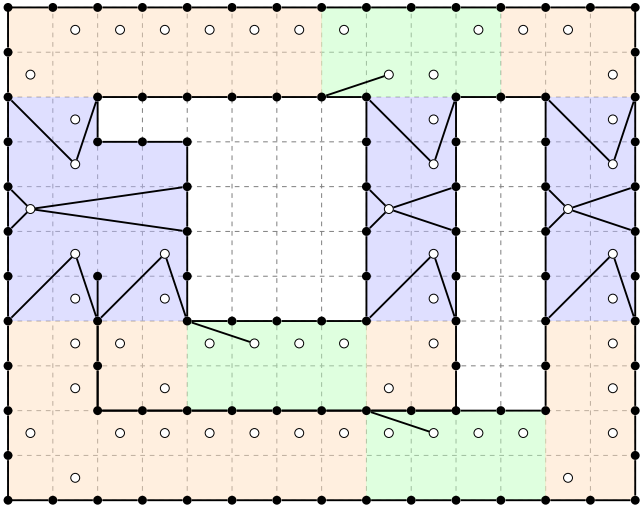




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# Tunnels & Pushes

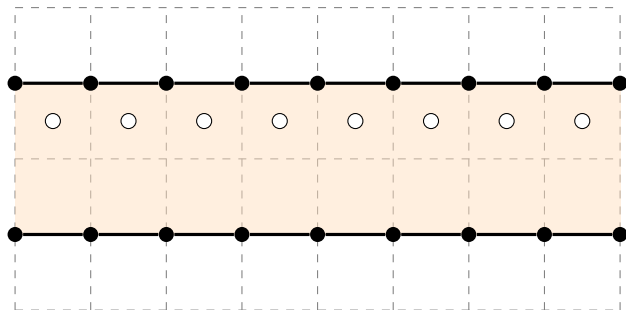
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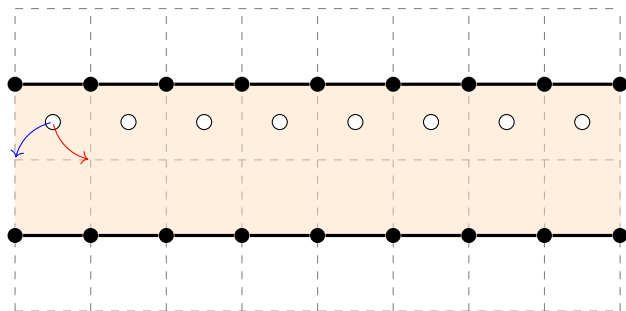
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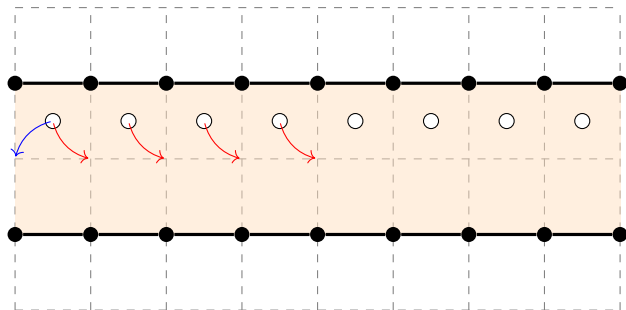
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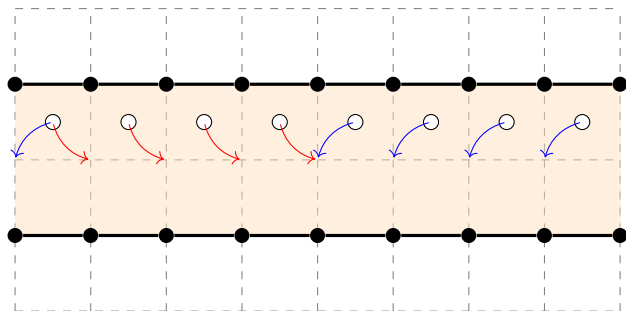
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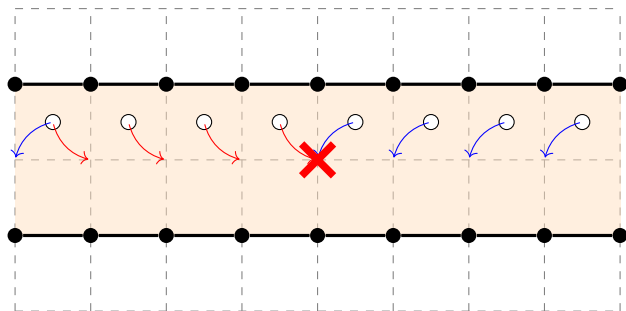
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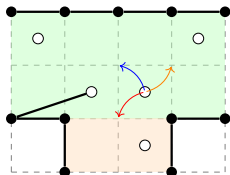




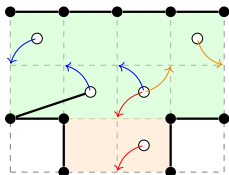
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- Topological safety ensures consistency of transmission.

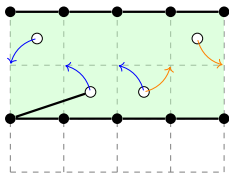




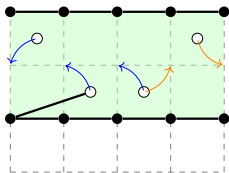
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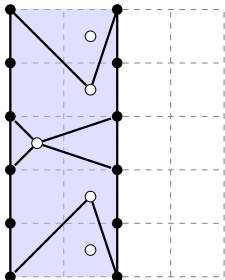
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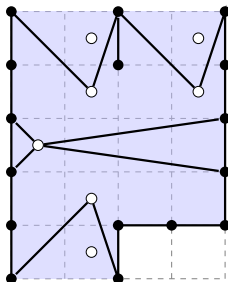


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- All-unnegated gadgets are constructed mirroring at a horizontal line.



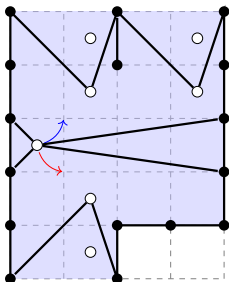
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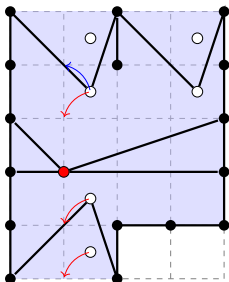
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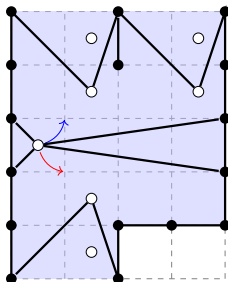


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- Moving the assignment vertex up equals a **TRUE**-assignment, **FALSE** otherwise.

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- If the formula is unsatisfiable, at least one black vertex has to be moved  $\Rightarrow c_{\min}$  is exceeded.

### Corollary

TOPOLOGIALLY SAFE SNAPPING *is also*  $\mathcal{NP}$ -hard *when using*  
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**Euclidean** TOPOLOGIALLY SAFE SNAPPING with the objective to minimize **maximum** movement is  $\mathcal{APX}$ -hard.

# Integer Linear Program

**Things to handle:**

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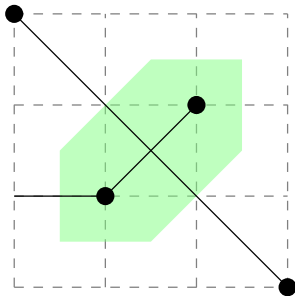
- Constraint: distinct vertex coordinates.

# Planarity

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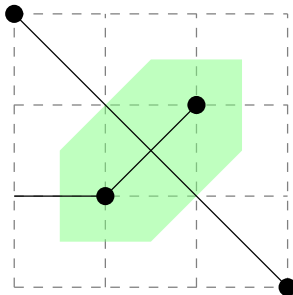
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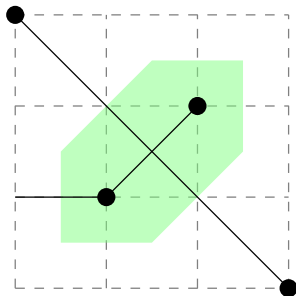


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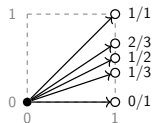
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- According to bounding box size:

$$D_{\min} = \frac{1}{\max\{X_{\max}, Y_{\max}\} + 1}$$

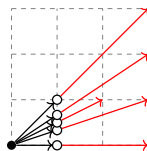
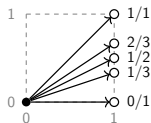


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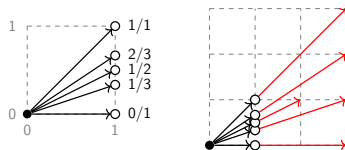


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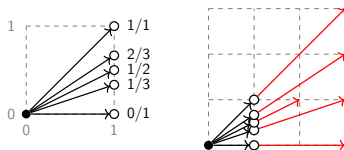
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- Inside  $[-k, k] \times [-k, k]$  area, there are  $\Theta(k^2)$  directions to consider.
- Consider them to be ordered counter-clockwise.

- Circular order of neighbors around any vertex must not change.

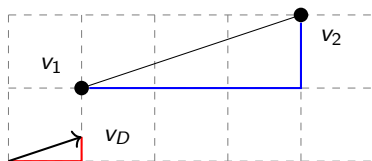
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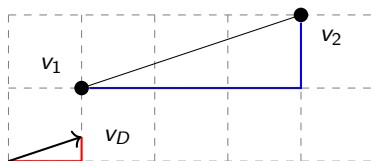
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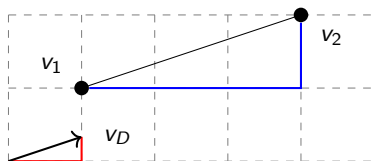
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- Map edges to directions and compare the ordering of those directions to the given embedding.

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- In practice, our model easily becomes too large to solve (in reasonable time).
- We use **delayed constraint generation** to iteratively improve our model.
- We generate most constraints on demand: first iteration is simple rounding (with unique coordinates).



## Experimental Evaluation

- Using the JAVA bindings for IBM CPLEX.

# The Setup

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- Test system: Linux server with 16 cores (2666 MHz, 4 MB cache), 16 GB main memory.

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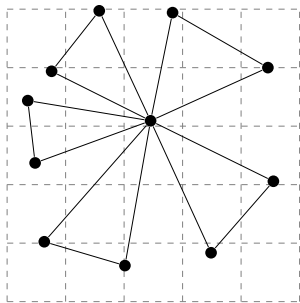
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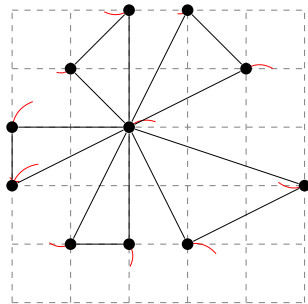
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- For delayed constraint generation, time is accumulated total.

# The Good



	Full	Delayed
rows		
cols		
time		

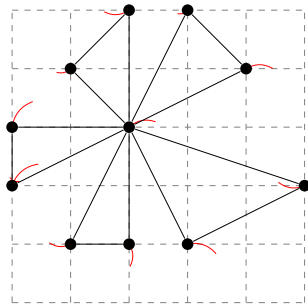
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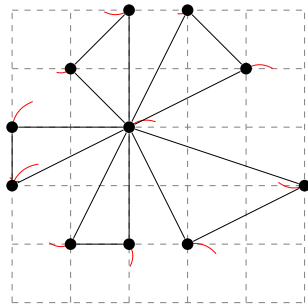
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- Even small examples take several seconds to solve.

	Full	Delayed
rows	42 699	
cols	11 300	
time	10.6 s	

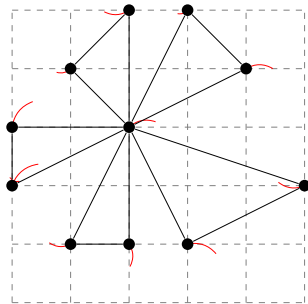
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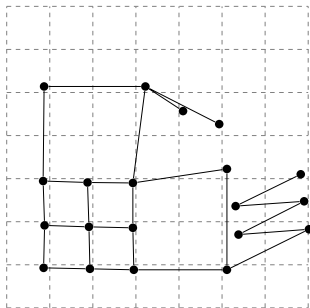
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- Delayed constraint generation gives speed-up.

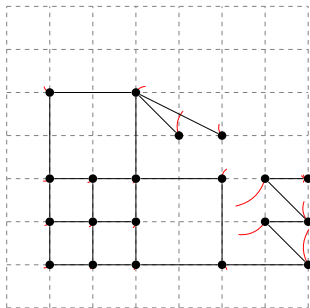
	Full	Delayed
rows	42 699	88
cols	11 300	110
time	10.6 s	0.5 s

# The Bad



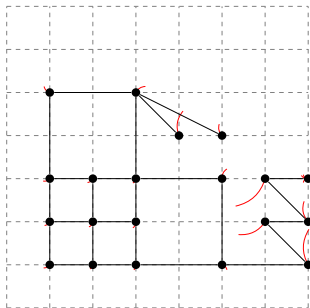
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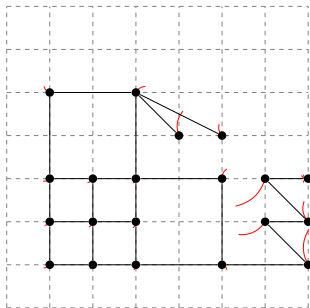
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- We have canceled this computation after 10 minutes using the full model.

	Full	Delayed
rows	323	441
cols	82	816
time		†

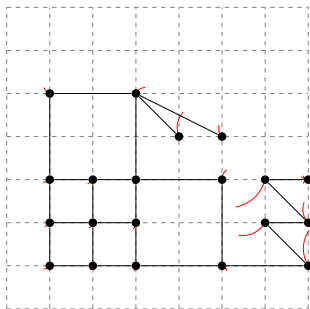
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cols	82 816	4 044
time	†	

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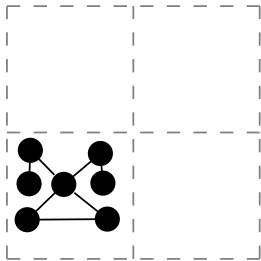


- We have canceled this computation after 10 minutes using the full model.
- Delayed constraint generation did cut a lot of “trivial” constraints, but...
- ...waiting more than 3 minutes is too long for a graph on 20 vertices!

	Full	Delayed
rows	323 441	15 161
cols	82 816	4 044
time	†	211.6 s

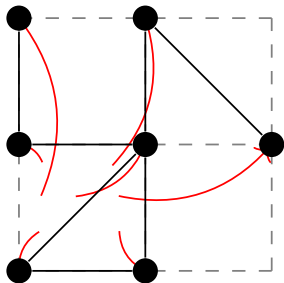


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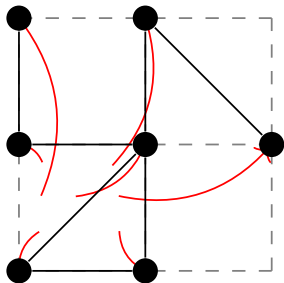
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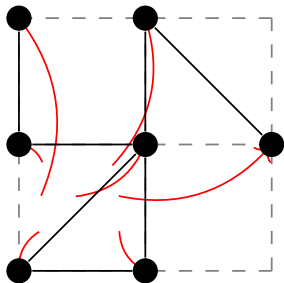
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- Graph and bounding box are small, thus the model is small.

	Full	Delayed
rows	2 603	
cols	916	
time	4.8 s	

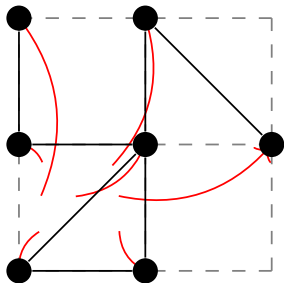
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	Full	Delayed
rows	2 603	2 271
cols	916	816
time	4.8 s	20.2 s

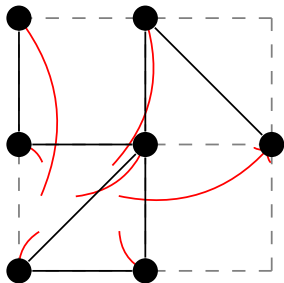
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- Questions about approximability remain open.