

# Comparing Temporal Graphs

## with Time Warping

Malte Renken

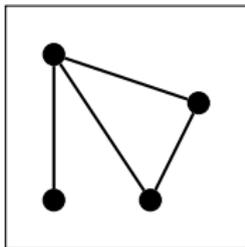


Algorithmics and Computational Complexity,  
TU Berlin, Germany

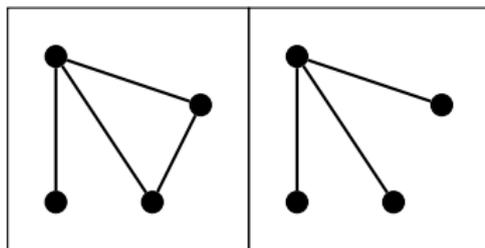
8. July 2019

Joint work with  
Vincent Froese, Brijnesh Jain and Rolf Niedermeier.

# Temporal Graphs



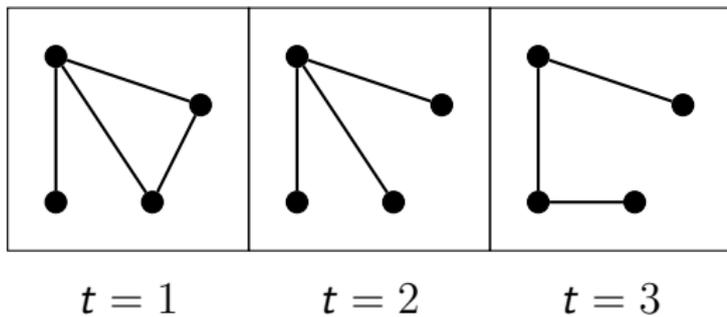
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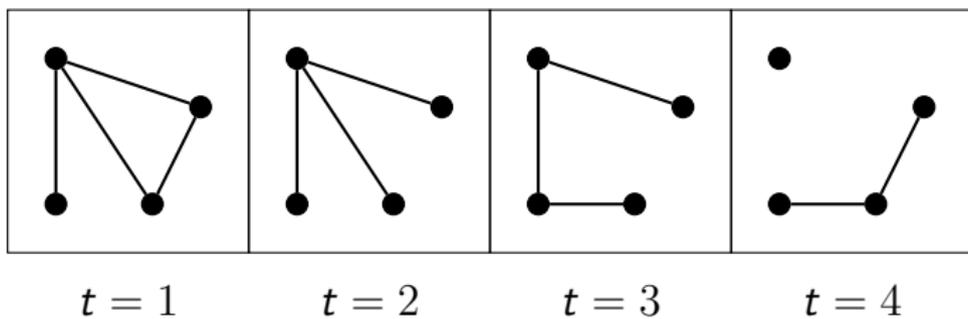
$t = 1$

$t = 2$

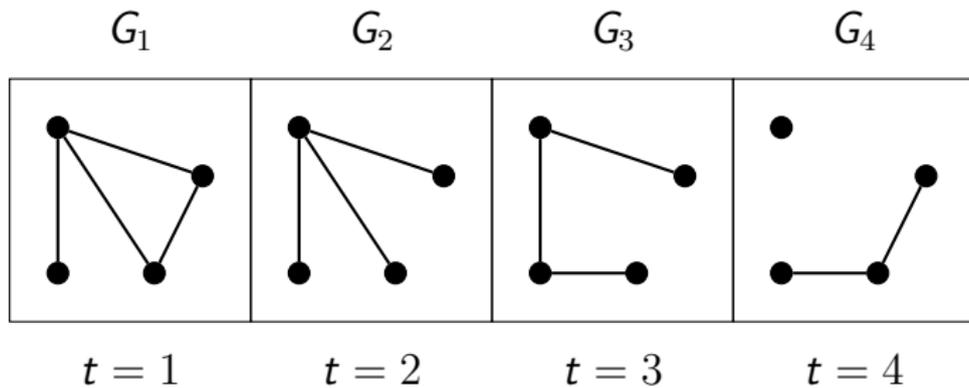
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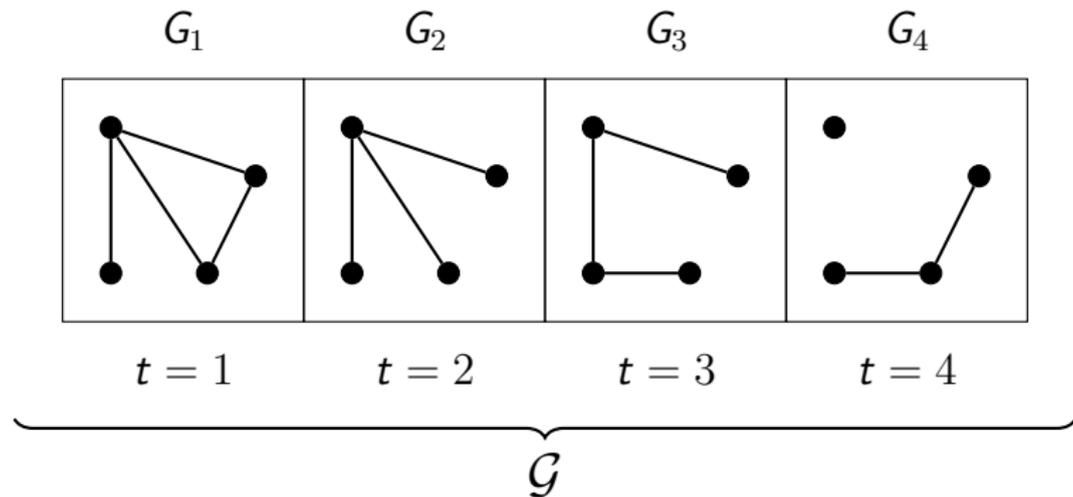
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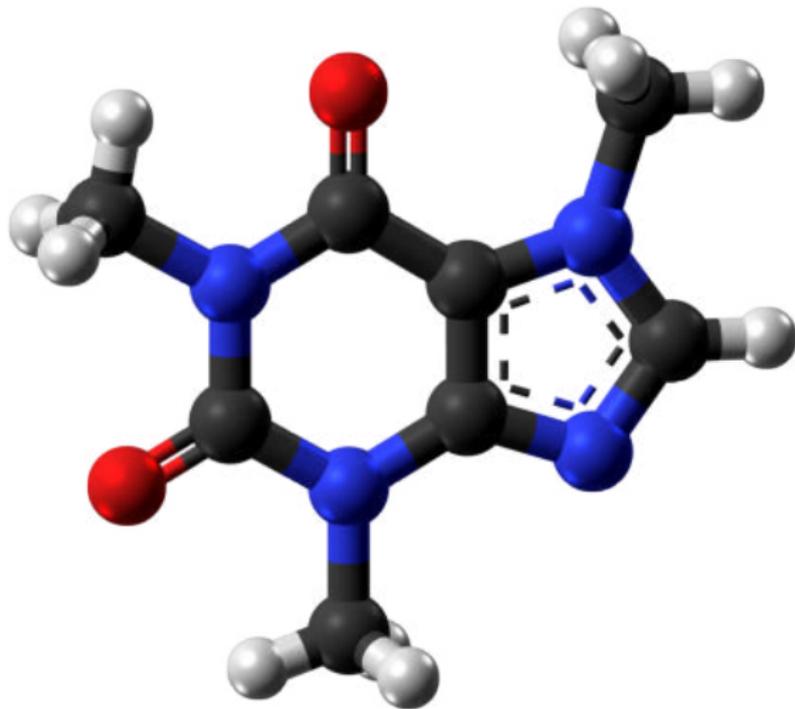


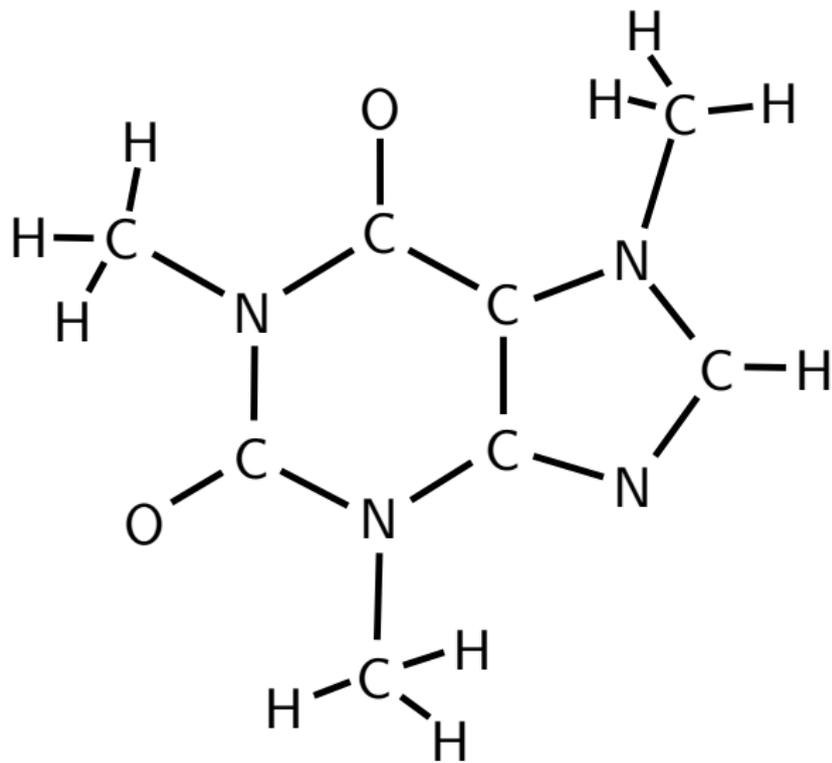
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112.2

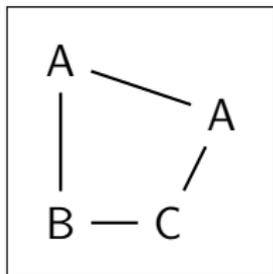


## Main Question

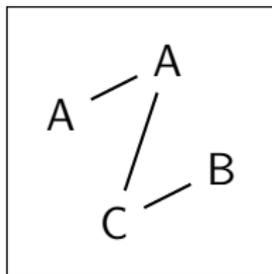
How can we measure the similarity / distance between two temporal graphs  $\mathcal{G}$ ,  $\mathcal{H}$ ?

# Graph distance using vertex signatures

*G*

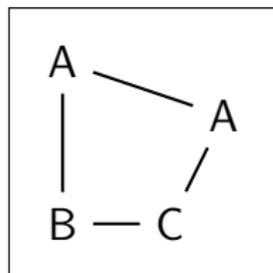


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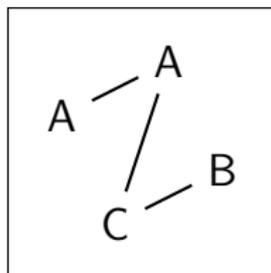


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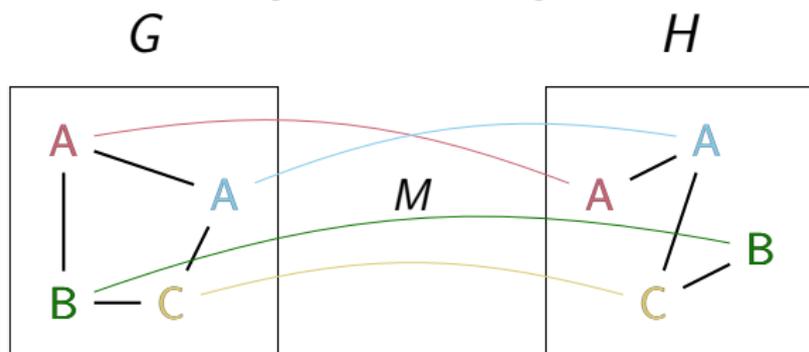
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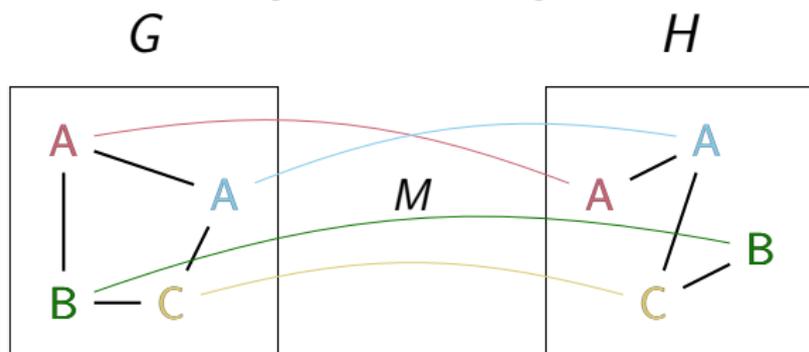
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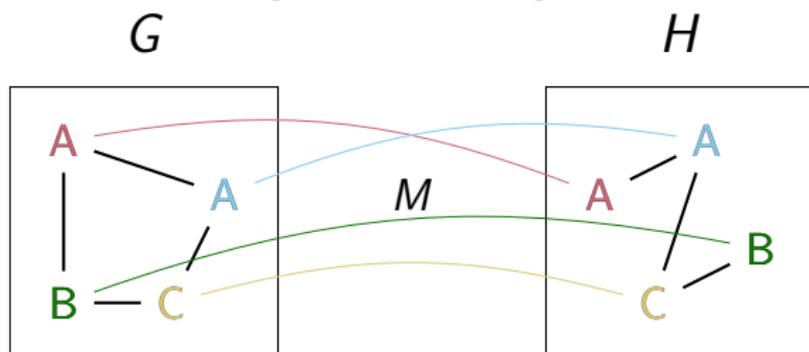


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all maximal matchings between the two vertex sets

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- ▶ Computation in cubic time using Jonker-Volgenant (or Hungarian).

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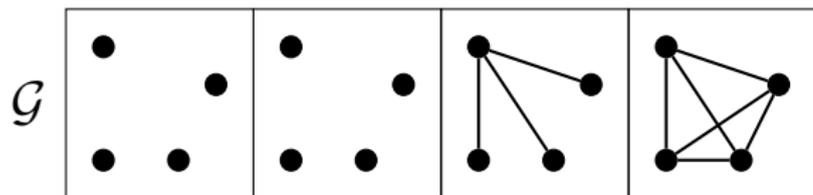
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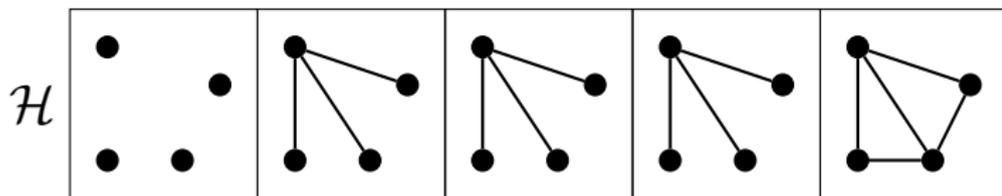
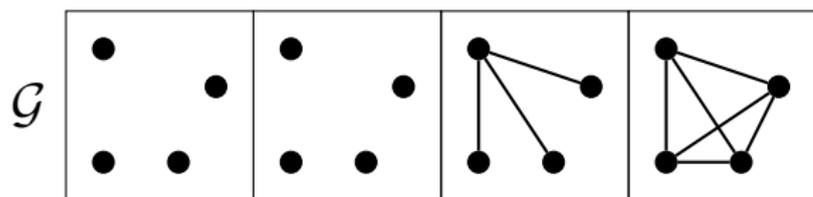
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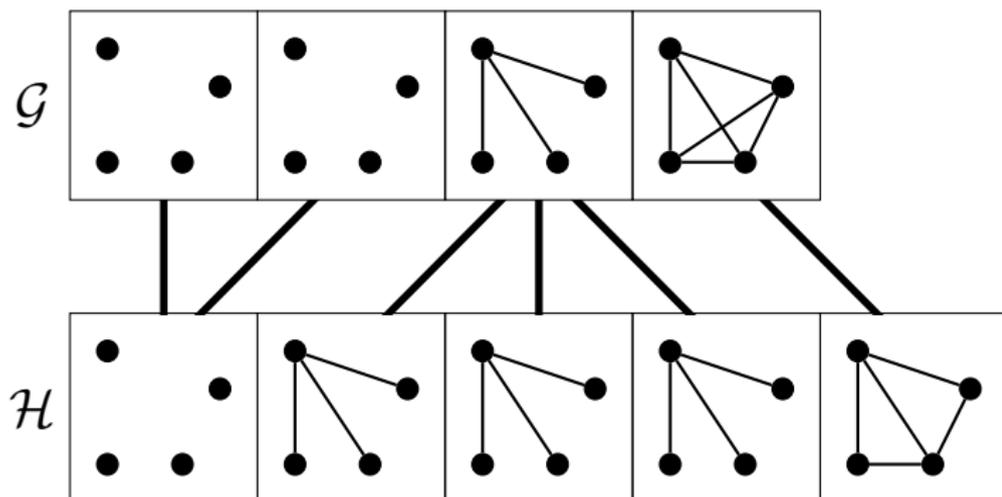
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**Solution:** **Time warping** — assign each layer to the other one it resembles most (no crossings allowed!).

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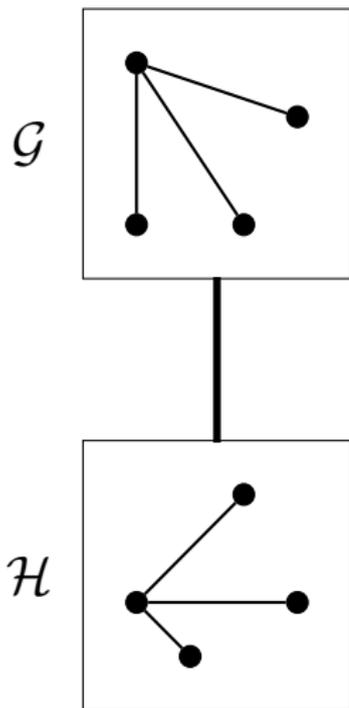
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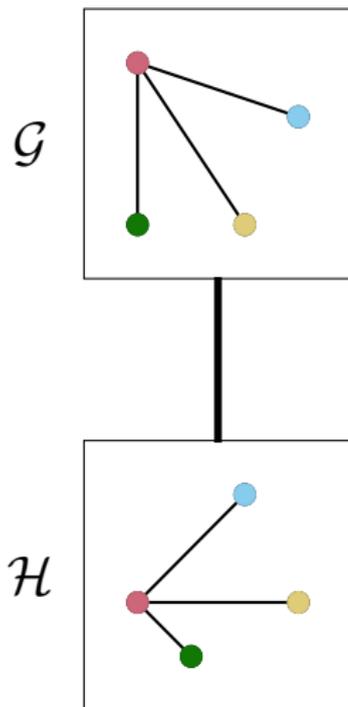
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**Bad news:** ... if all pairwise distances are known in advance.

# Pairwise distances

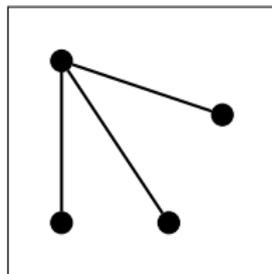


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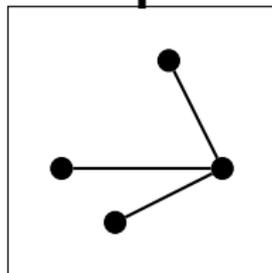


# Pairwise distances

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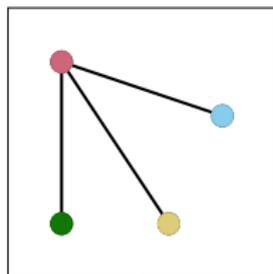


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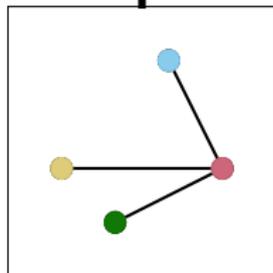


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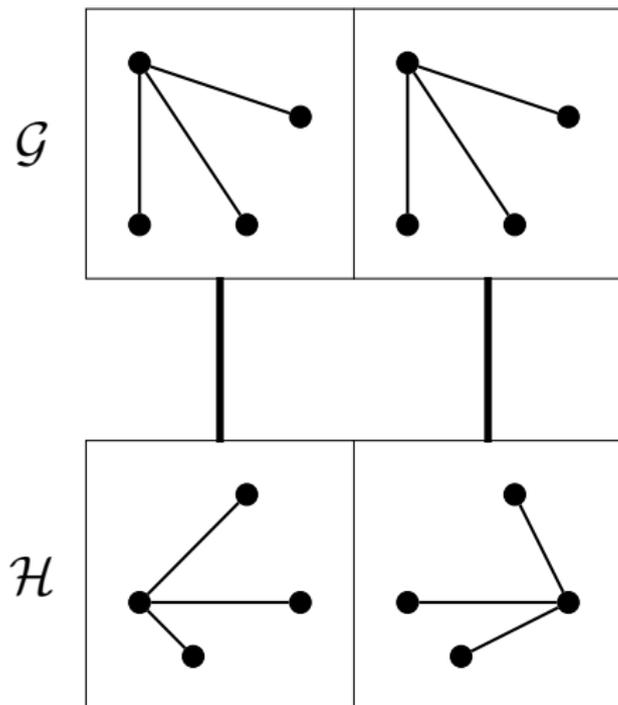
$\mathcal{G}$



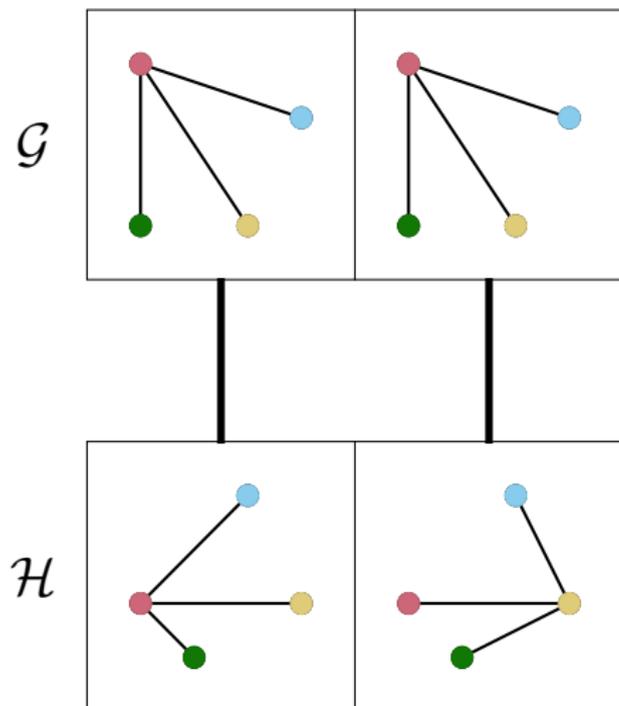
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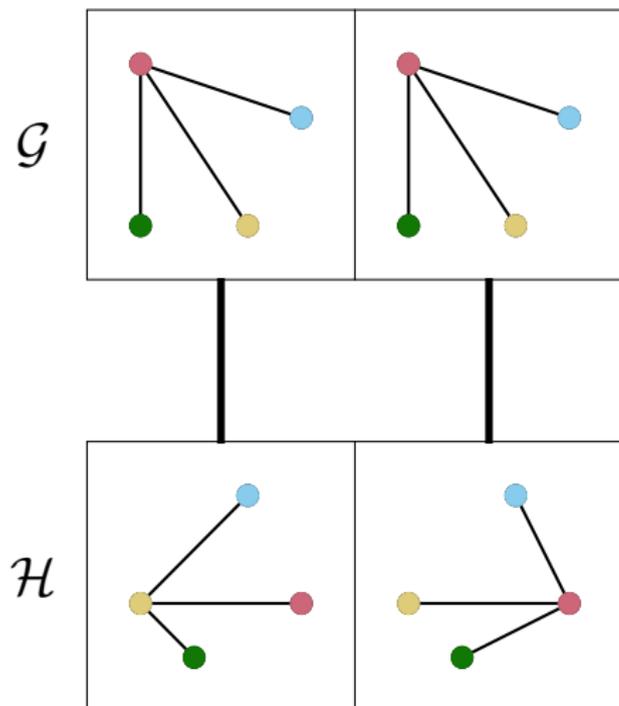
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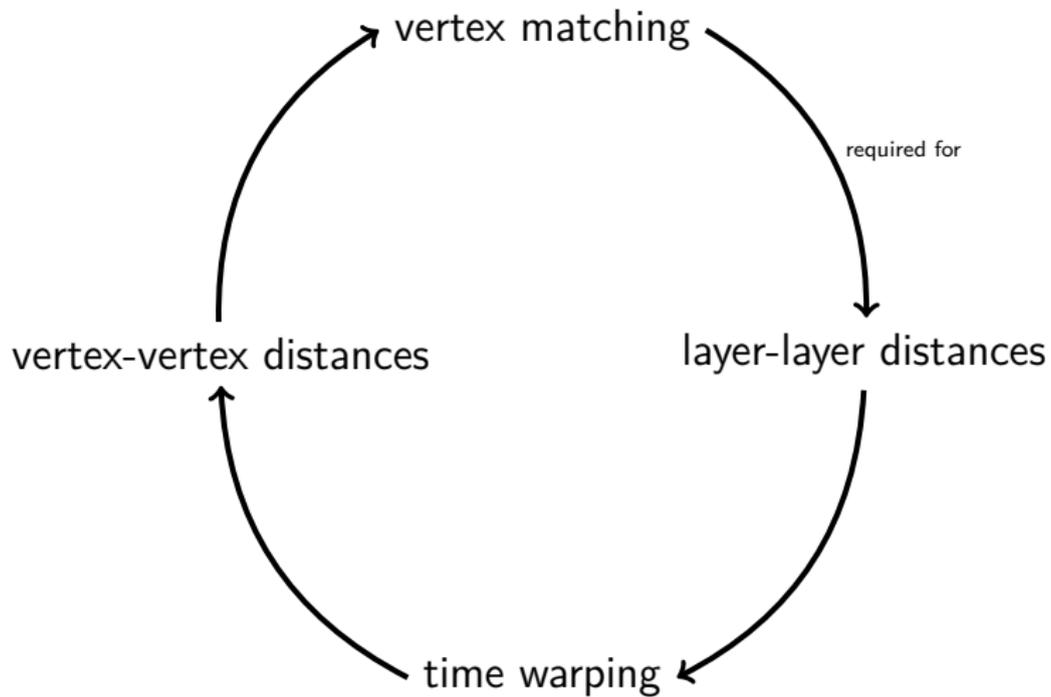


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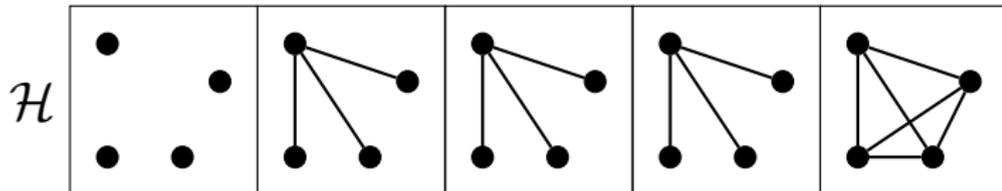
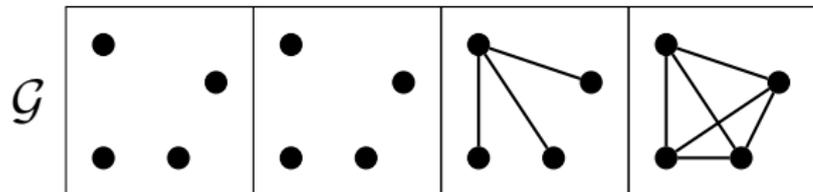
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- ▶ ... you can check in polynomial time whether  $\text{dtgw-dist}(\mathcal{G}, \mathcal{H}) = 0$ .

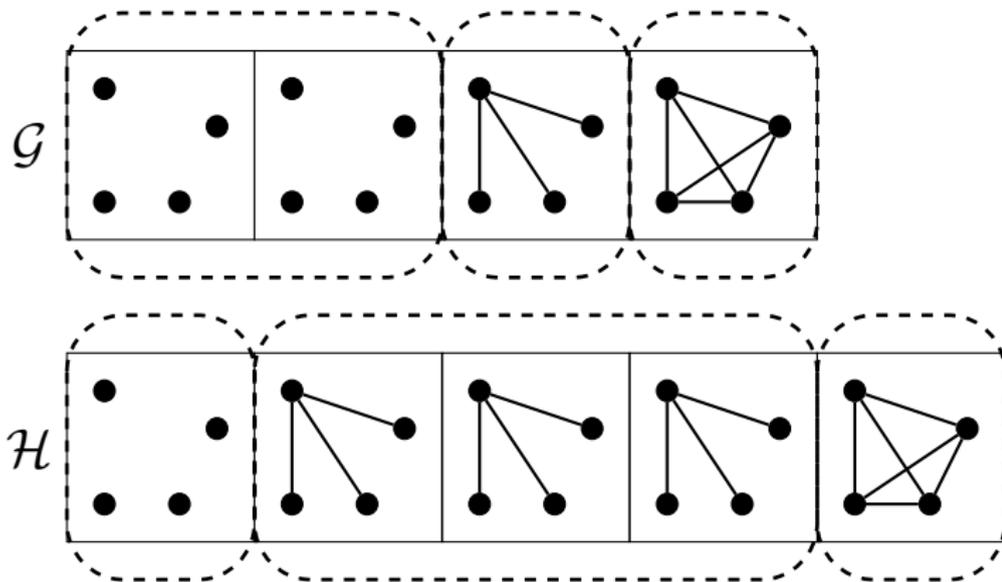
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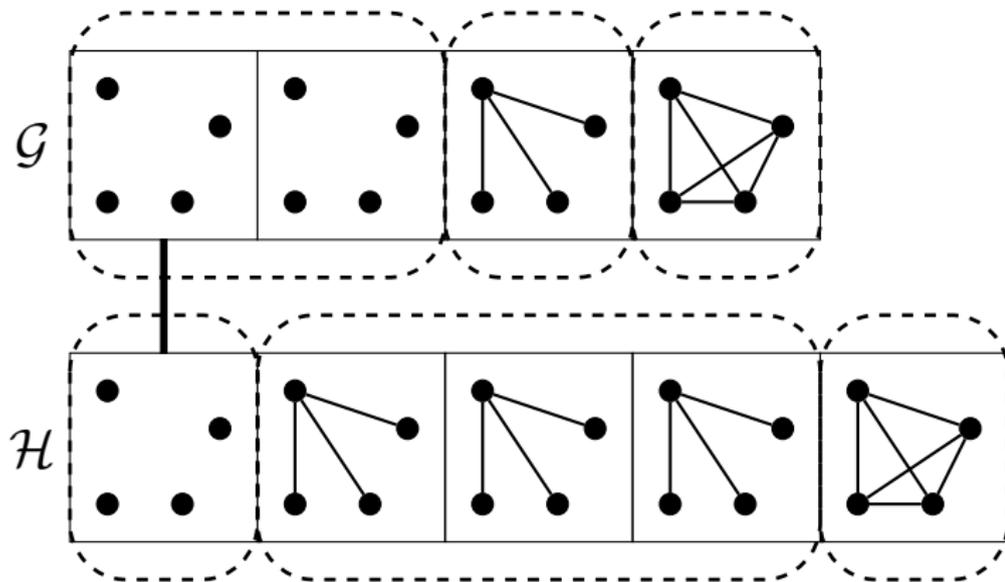
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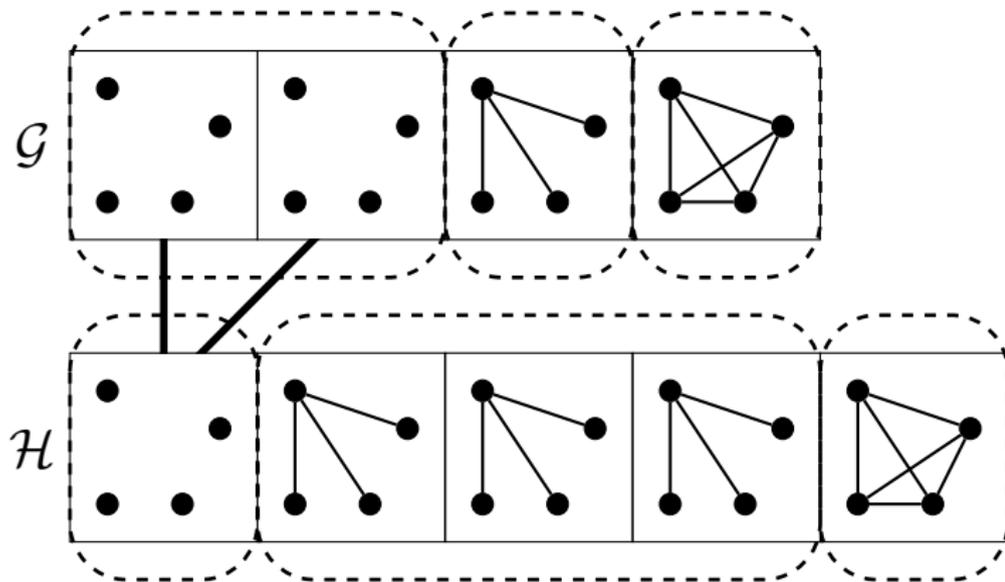
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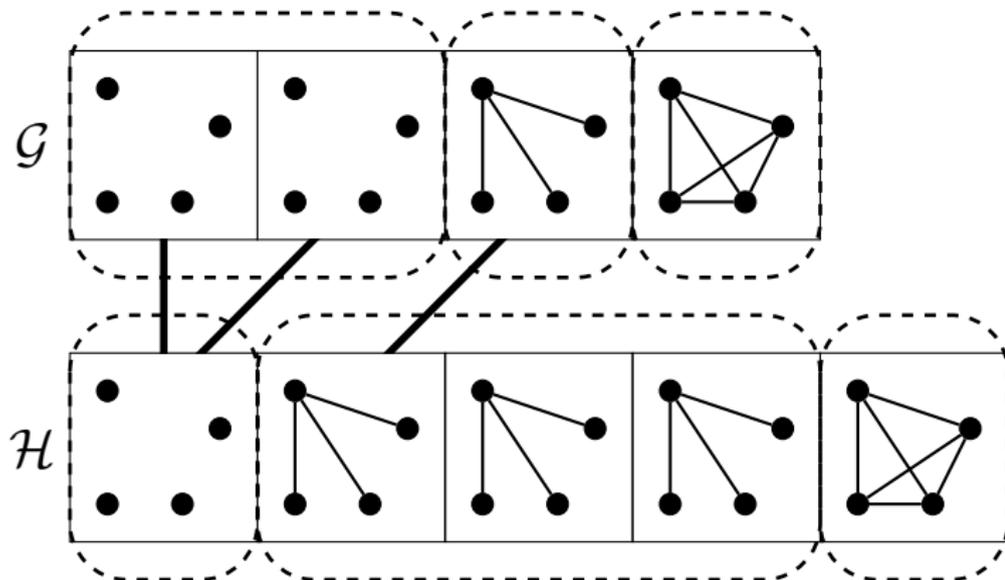
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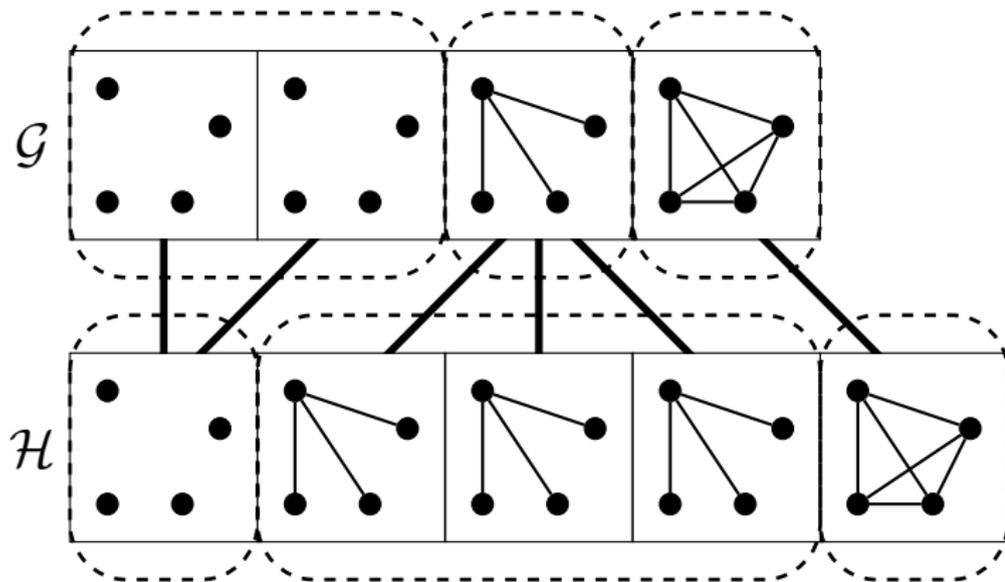
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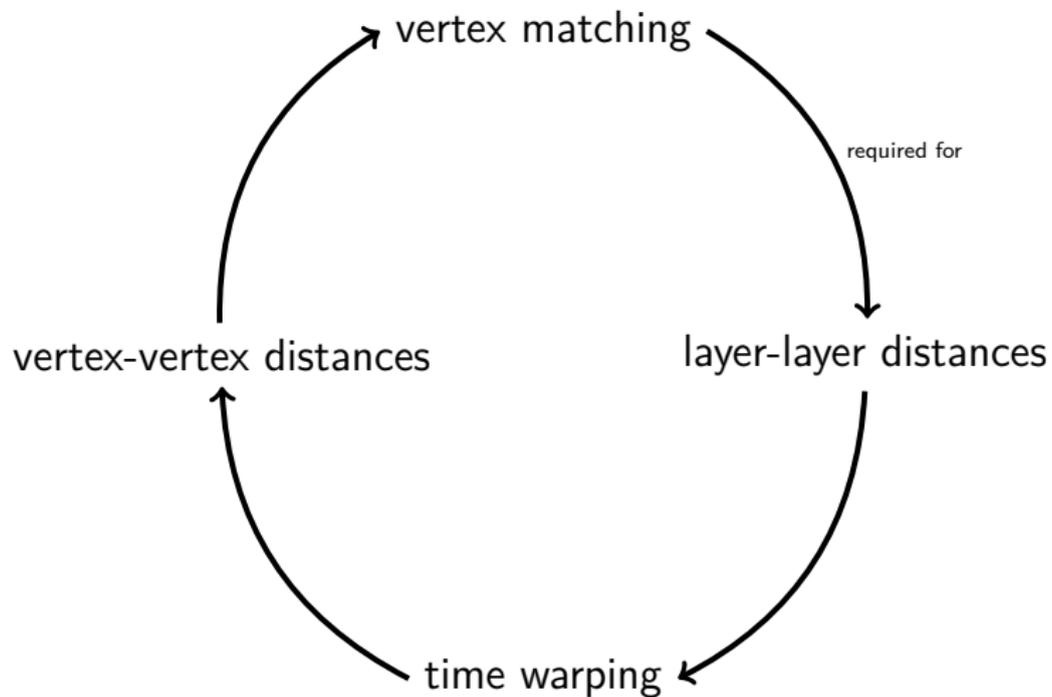
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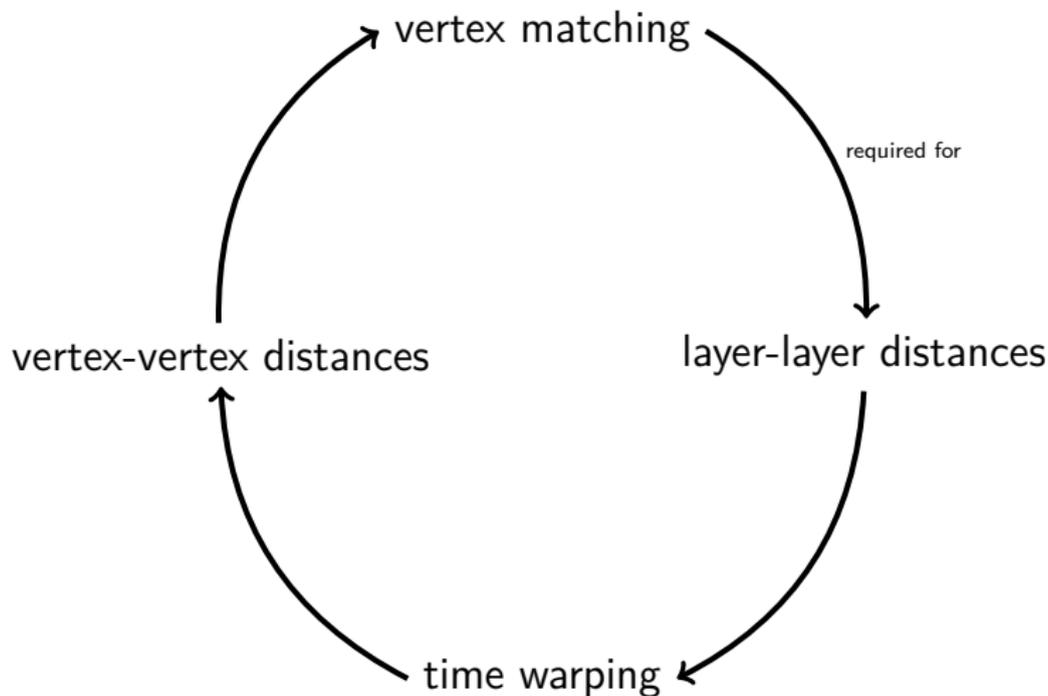
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- ▶ Depends surprisingly little on your initial guess.
- ▶ Seems to produce good results.

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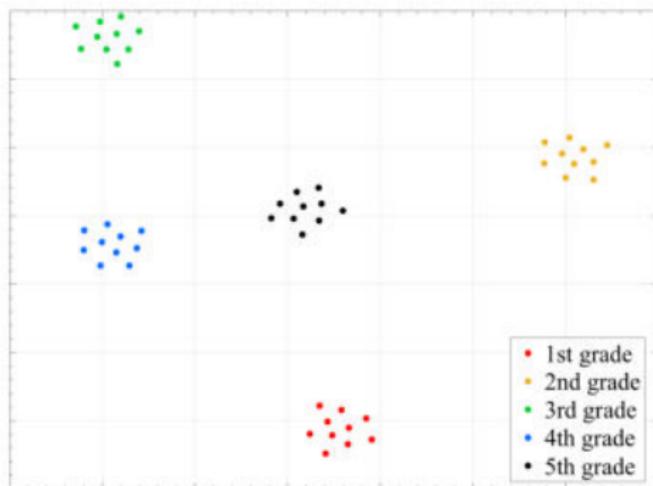
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## Result



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- ▶ Can you find approximation algorithms with guaranteed approximation quality?
- ▶ Which vertex signatures work best in different settings?