

# Approximate Minimum-Weight Partial Matching under Translation

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Wolfgang Mulzer <sup>3</sup>   Günter Rote <sup>3</sup>   Micha Sharir <sup>2</sup>   Allen Xiao <sup>1</sup>

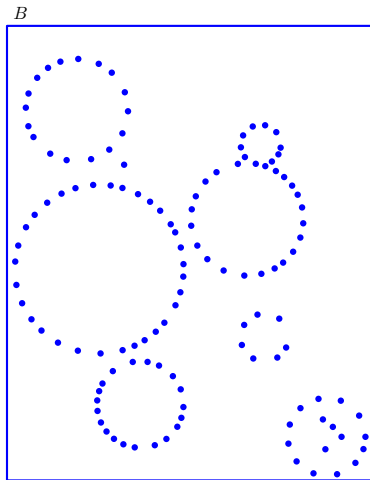
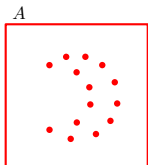
<sup>1</sup>Duke University

<sup>2</sup>Tel Aviv University

<sup>3</sup>Freie Universität Berlin

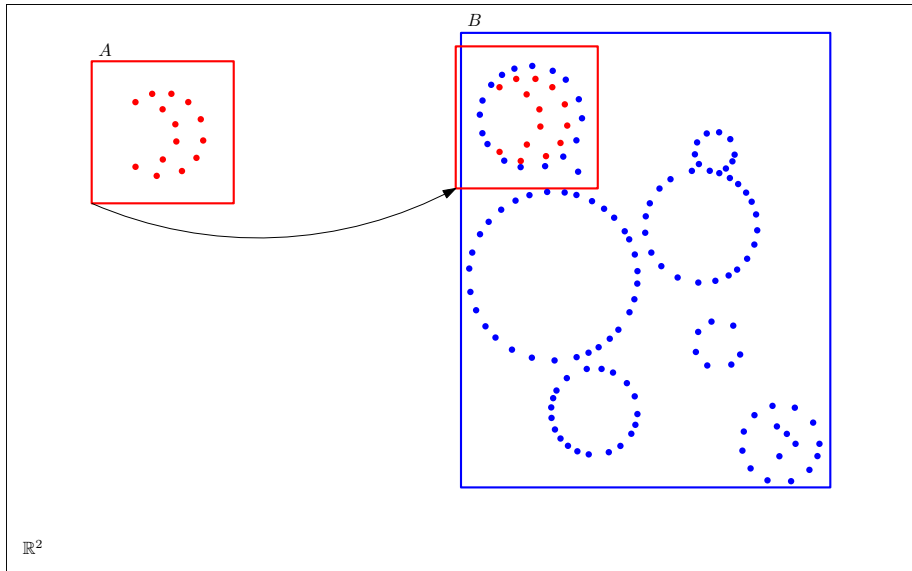
December 2018

# Example

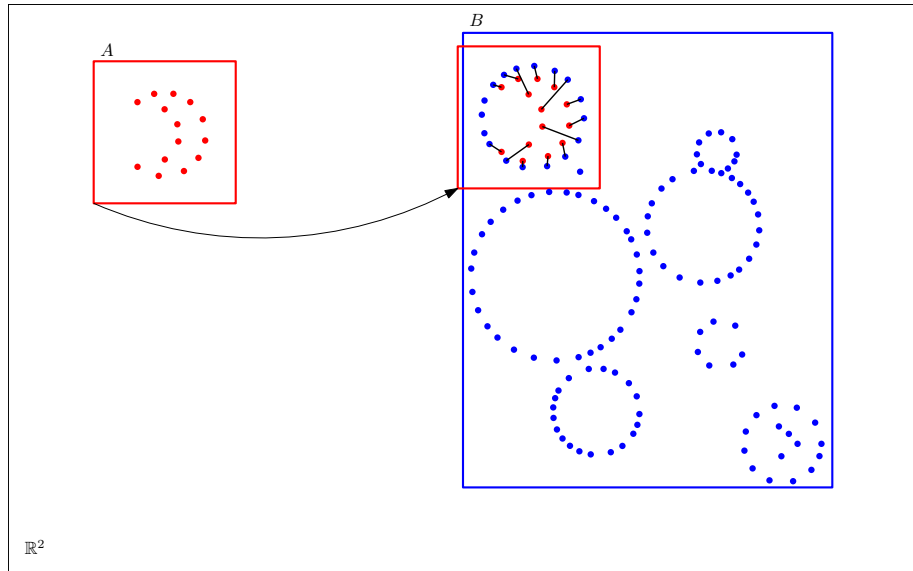


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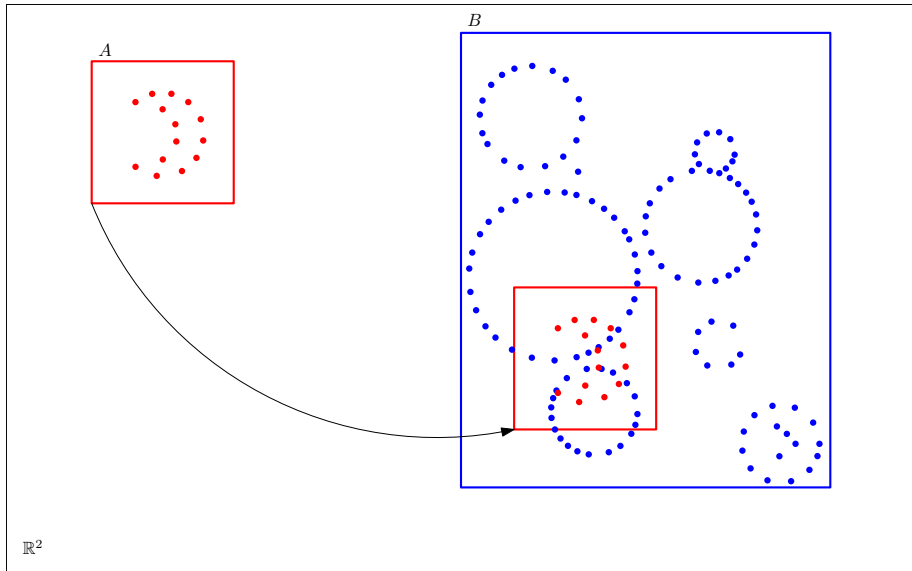
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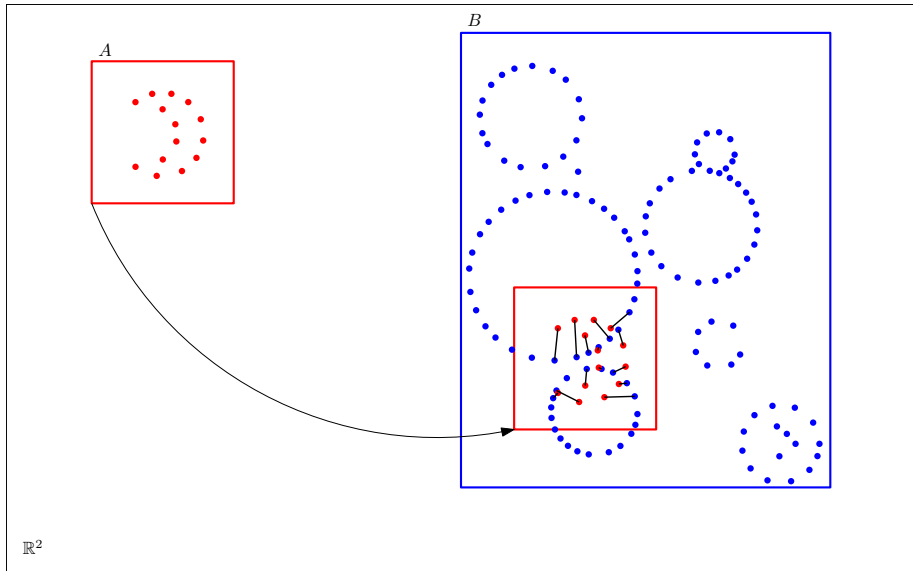
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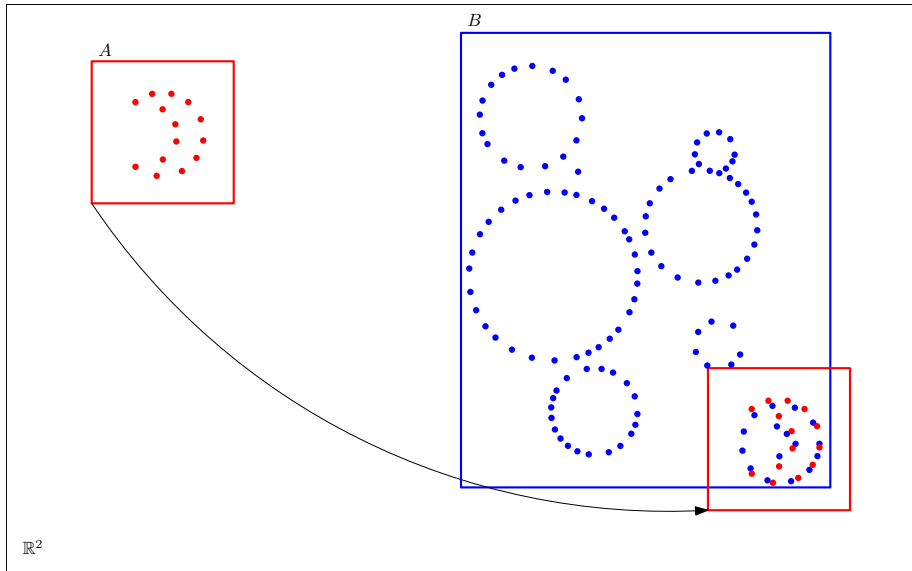
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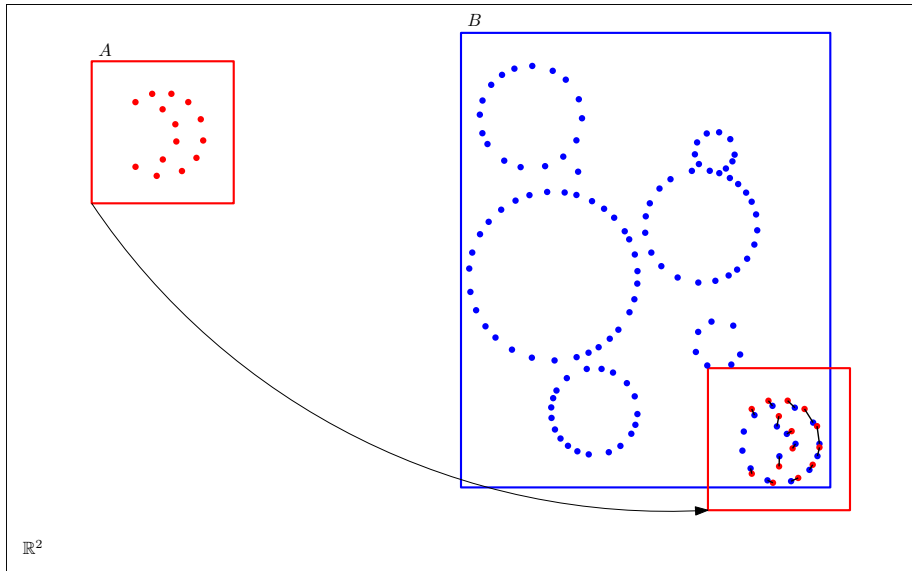
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Find the minimum cost matching over all translations



# Cost function

- ▶ Given point sets  $A, B$ , with  $|A| = m$  and  $|B| = n$ , match  $A$  into an  $m$ -subset of  $B$  after translation  $t$ .
- ▶ For matching  $M$  and translation  $t$ , define the **root-mean-square (RMS) cost** of the matching:

$$\text{cost}(M, t) := \left[ \frac{1}{m} \sum_{(a,b) \in M} \|a + t - b\|^p \right]^{1/2}$$

- ▶ Can generalize to  $p$ -th power/ $p$ -th root (versus 2), and matching of size  $k$  (versus  $m$ ).
- ▶ Best matching cost at each translation:

$$\text{cost}^*(t) := \min_{\text{matching } M} \text{cost}(M, t)$$

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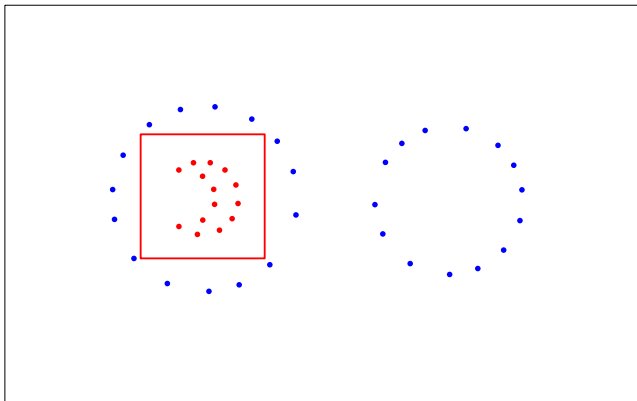
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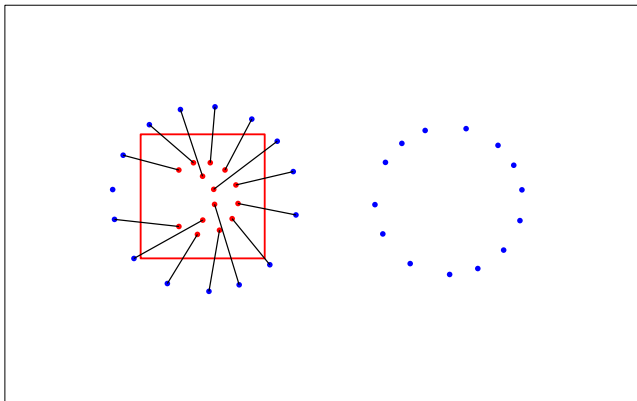
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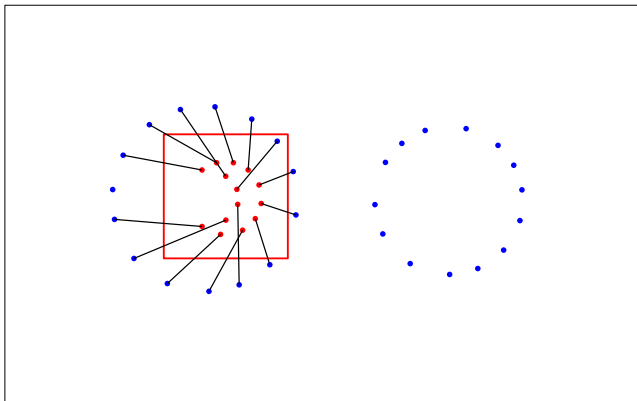
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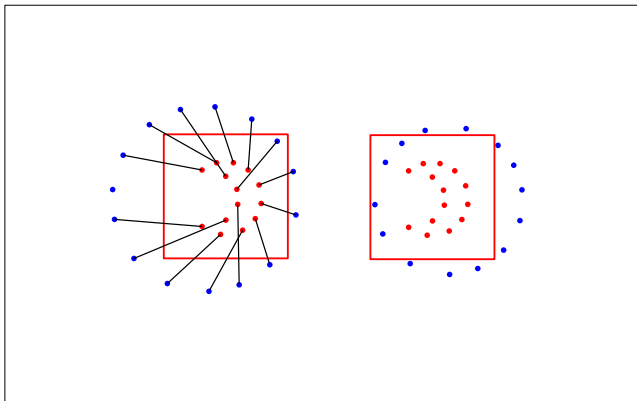
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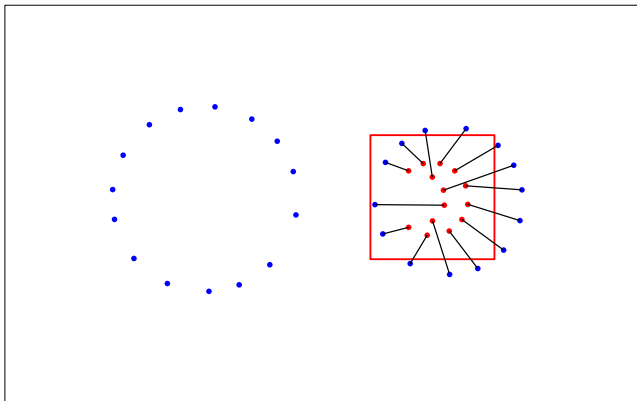
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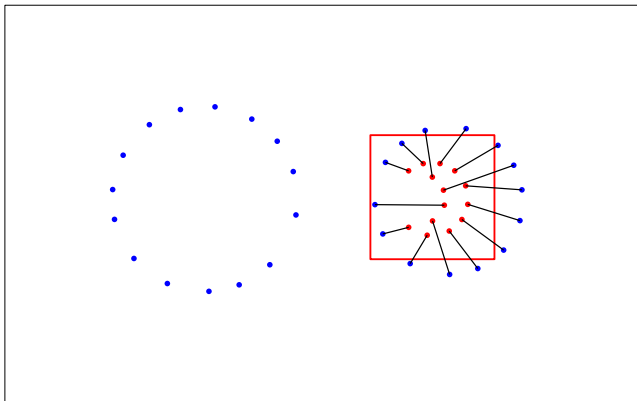


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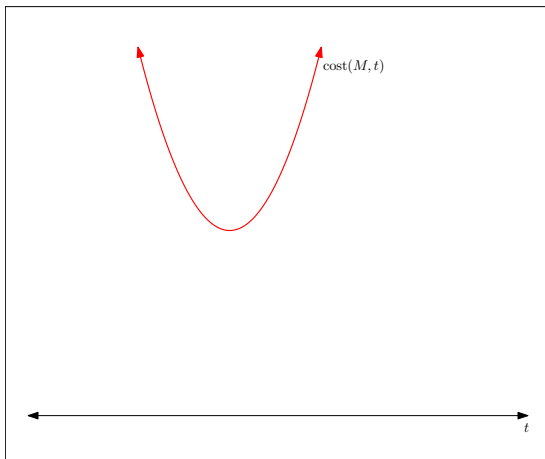
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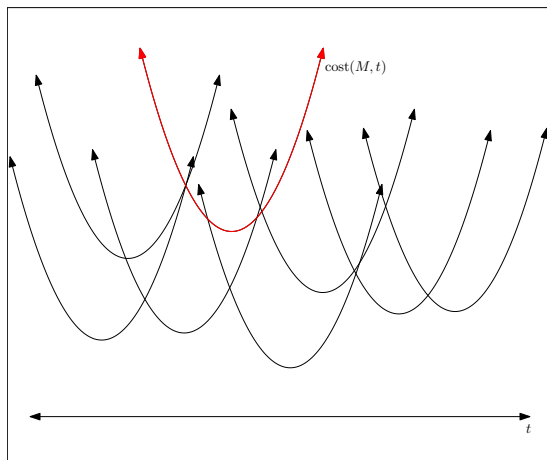
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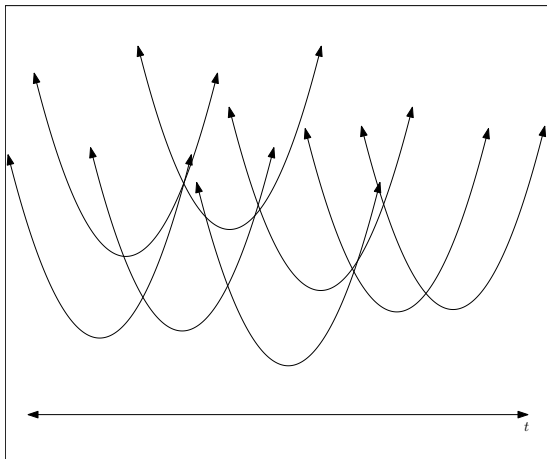
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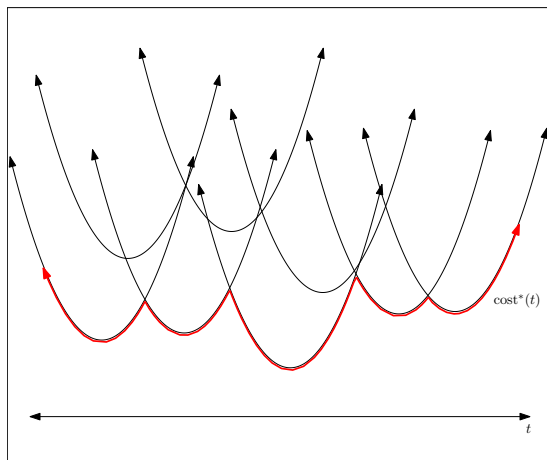
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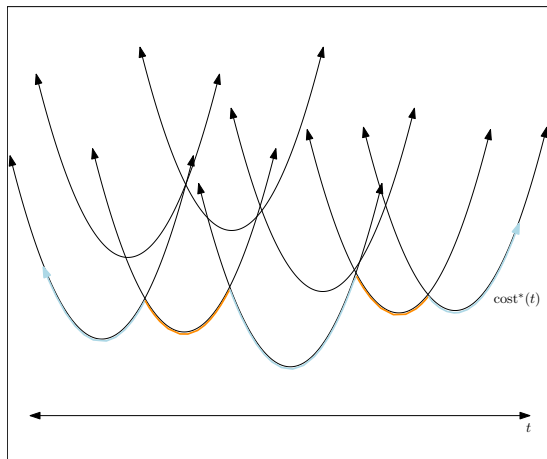
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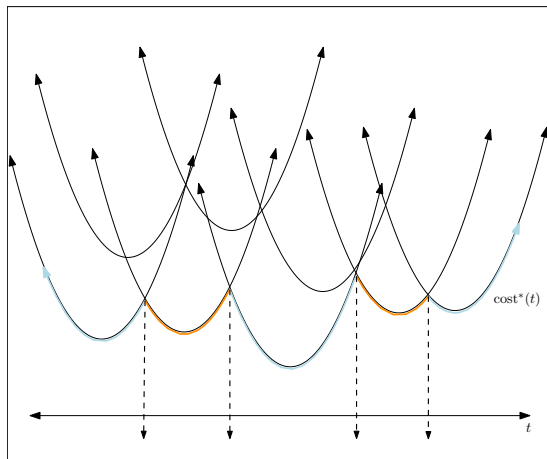
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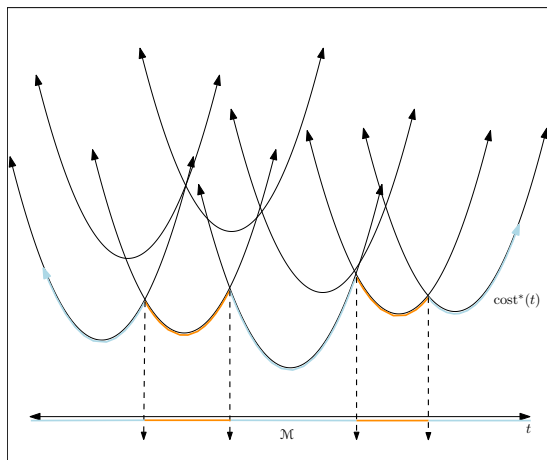
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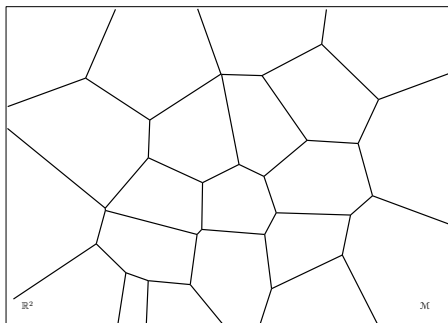
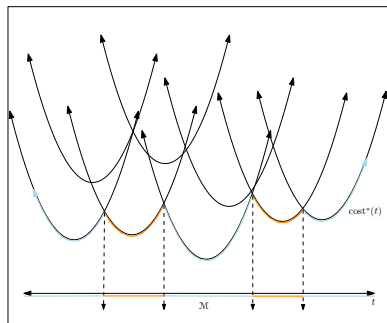


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# How many distinct matchings? (2-dimensional)



- ▶ How many distinct matchings appear in  $\text{cost}^*(t)$ ?
- ▶ What is the combinatorial complexity of  $\mathcal{M}$ ?

# Questions and prior results

1. How quickly can we **compute**  $t^*$ , a global minimum of  $\text{cost}^*(t)$ ?
  2. What is the **combinatorial complexity** of  $\mathcal{M}$ ?
    - ▶ [Rote 10]: in 1D, at most  $m(n - m) + 1$ .
    - ▶ [Ben-Avraham *et al.* 14]:  $O(n^2 m^{3.5} (e \ln m + e)^m)$
- ▶ Open to find  $t^*$  in polytime, and whether  $\mathcal{M}$  has polynomial complexity.
- ▶ Approximation?

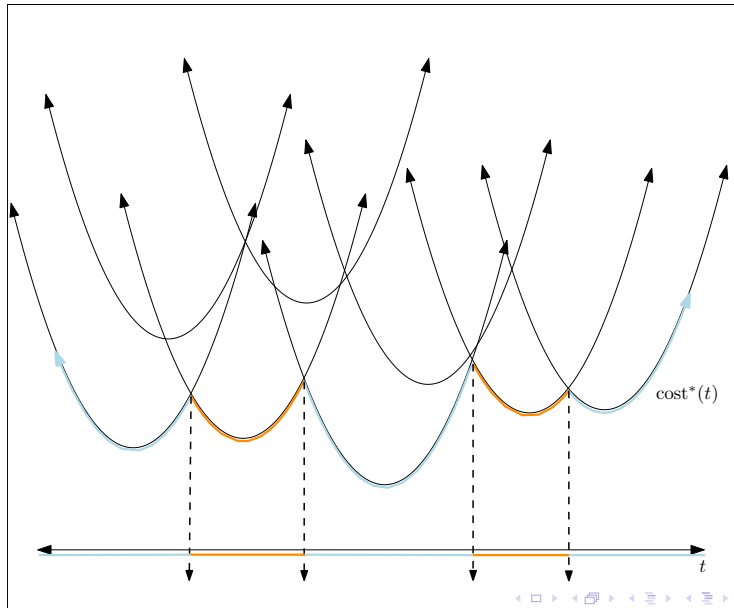
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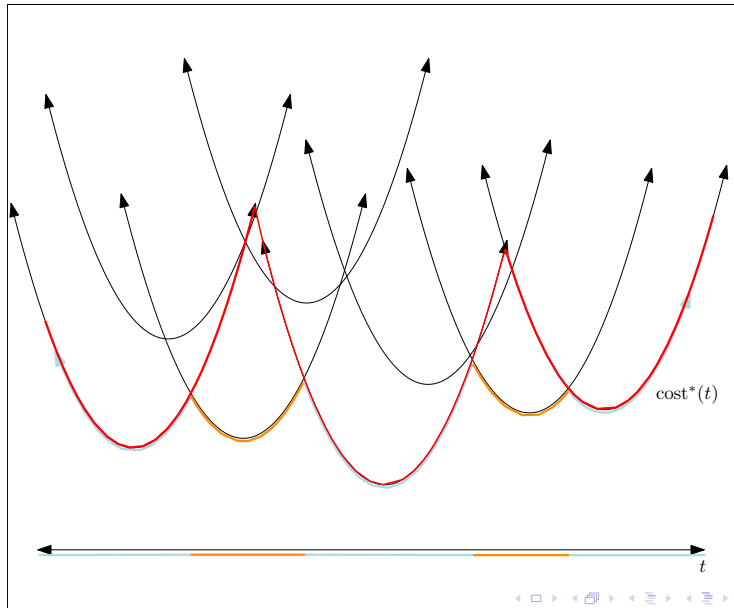
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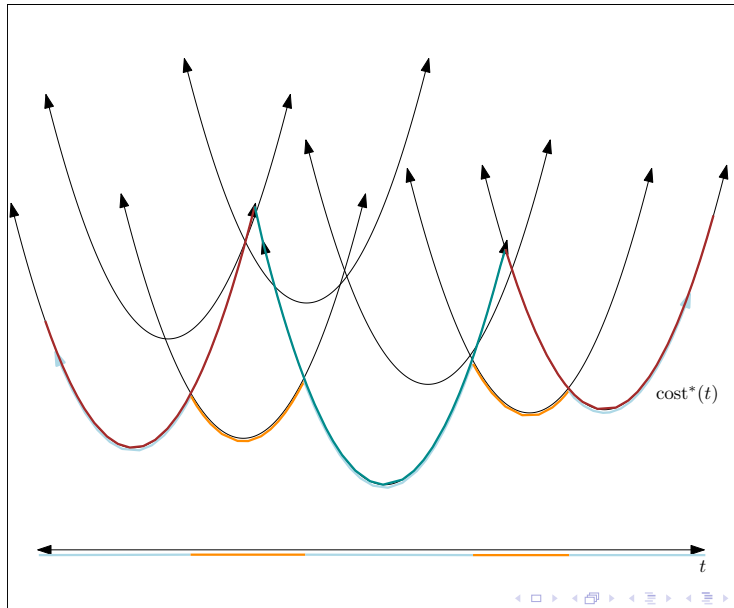
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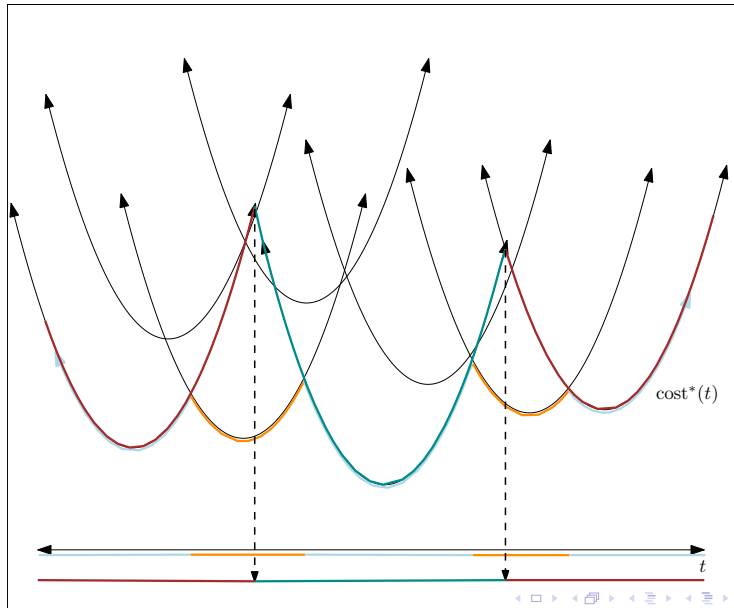


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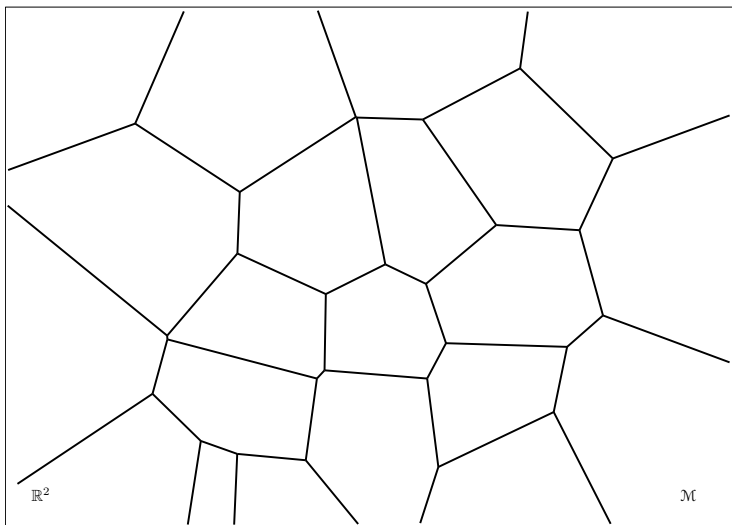




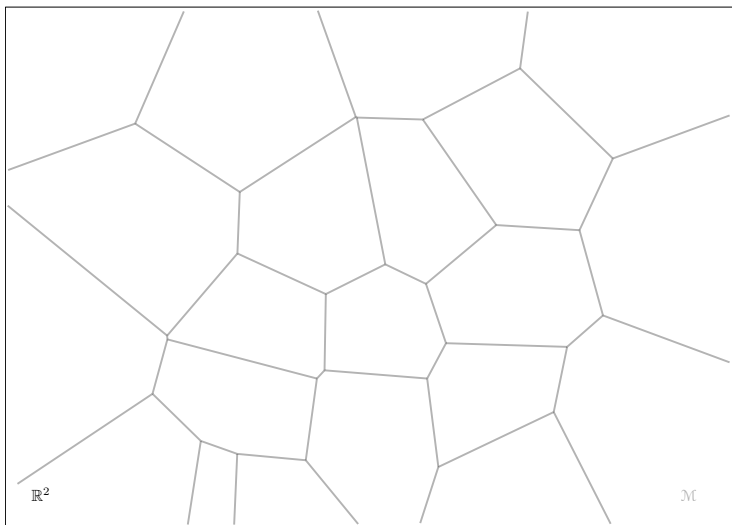
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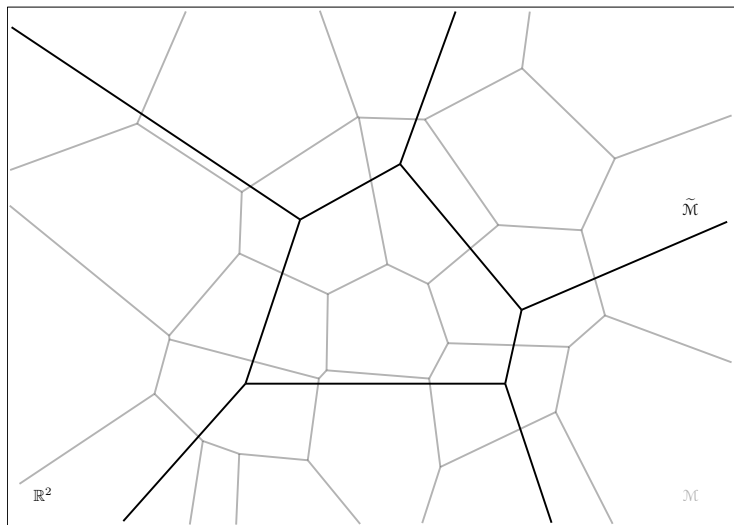
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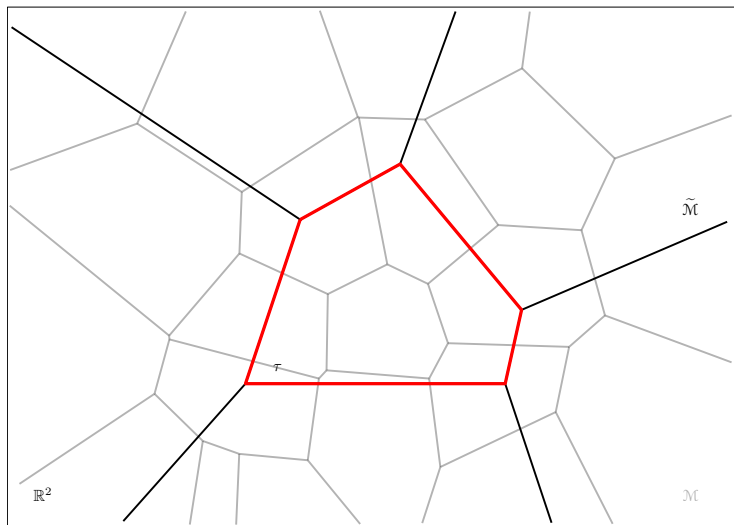
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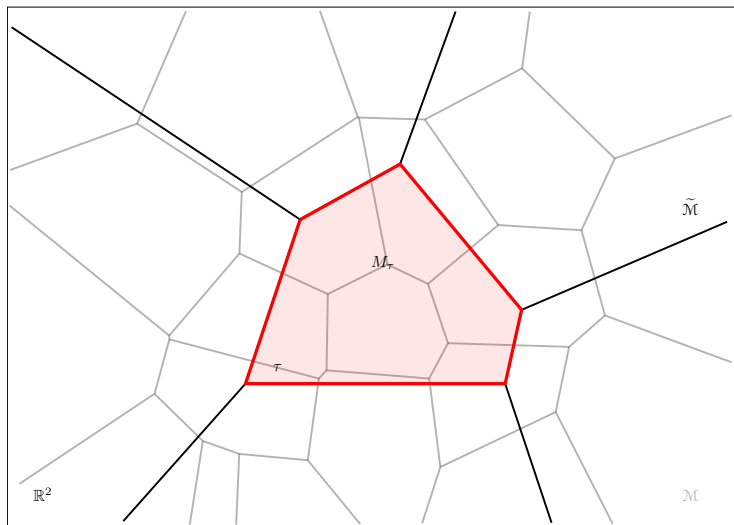
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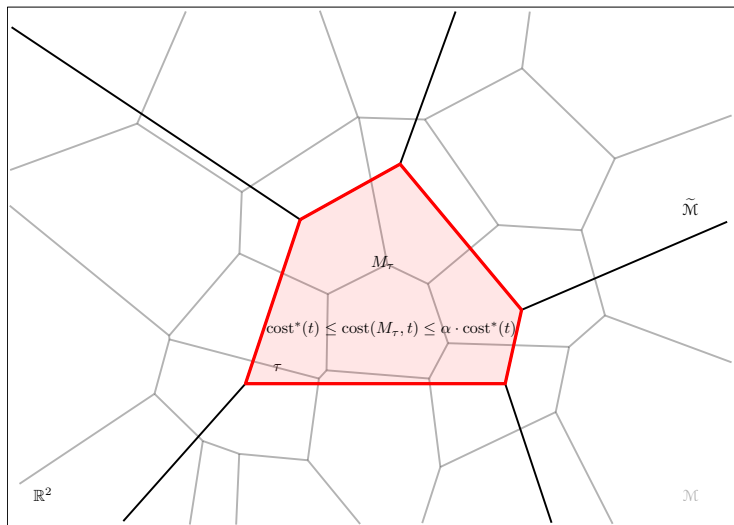
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# Our results (approximation helps)

1. How quickly can we **compute**  $t^*$ , a global minimum of  $\text{cost}^*(t)$ ?

## Theorem

*In  $\text{poly}(m, n, \varepsilon^{-1})$  time, can compute a  $(1 + \varepsilon)$  **approximation to  $\text{cost}^*(t^*)$**  by exploring the faces of  $\tilde{\mathcal{M}}$ .*

2. What is the **combinatorial complexity of  $\mathcal{M}$** ?

## Theorem

*In  $\text{poly}(m, n, \varepsilon^{-1})$  time, can construct a  $(1 + \varepsilon)$  **approximate diagram  $\tilde{\mathcal{M}}$**  of complexity  $O(n\varepsilon^{-2} \log \varepsilon^{-1})$ .*



1. The set of **point-to-point translations** give a constant approximate diagram of size  $mn$ .
2. Using exponential grids, constant  $\rightarrow (1 + \varepsilon)$  **approximation** of size  $O(mn\varepsilon^{-2} \log \varepsilon^{-1})$ .
3. Reduce size to  $O(n\varepsilon^{-2} \log \varepsilon^{-1})$  by clustering.

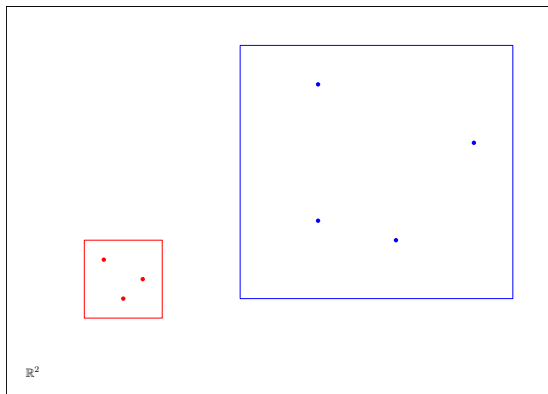
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$$T := \{t_{ba} = (b - a) \mid a \in A, b \in B\}$$

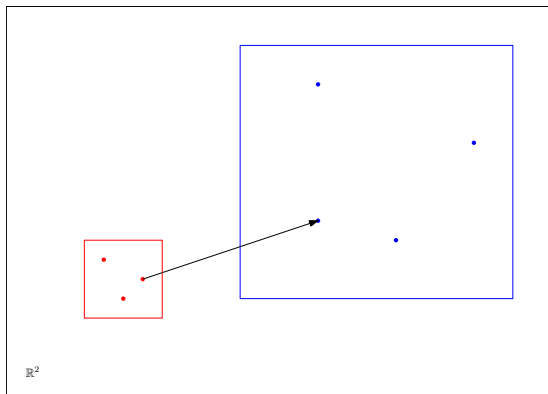


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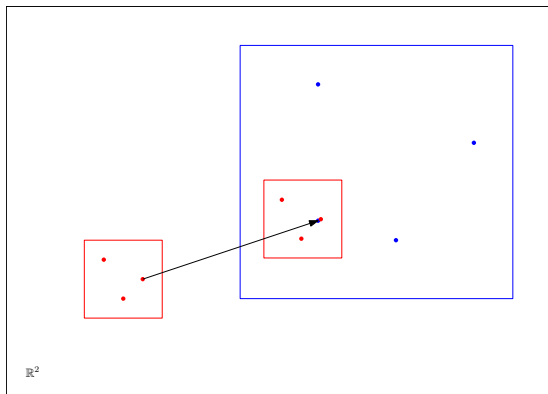


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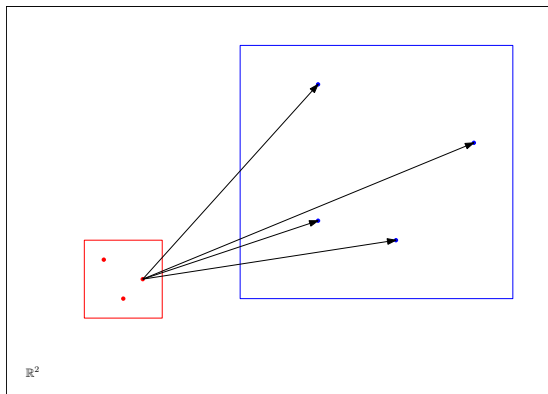


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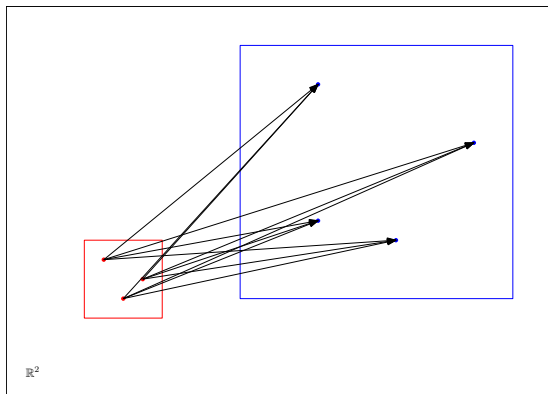


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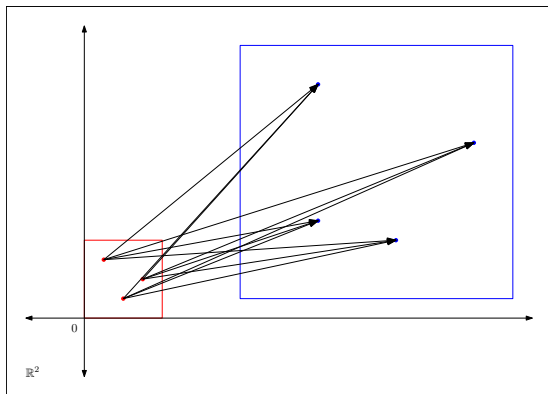
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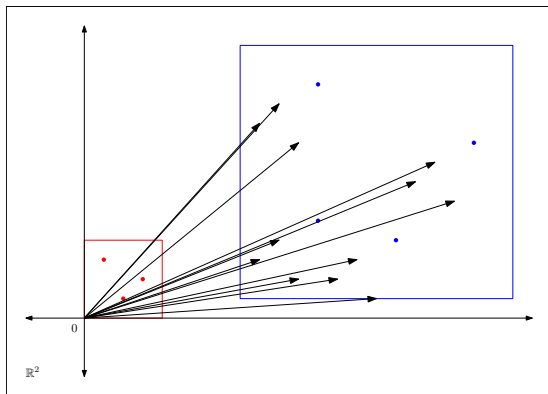


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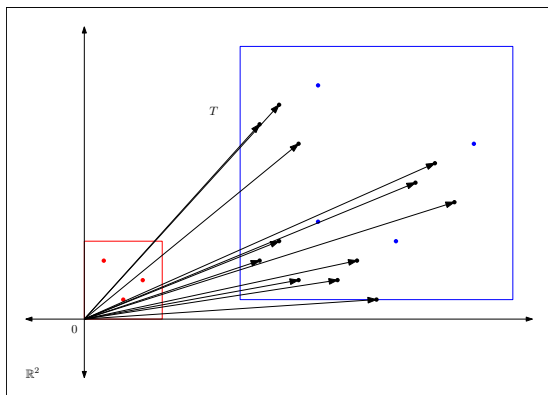


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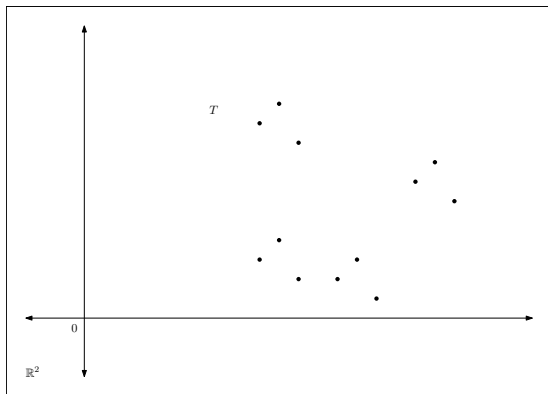


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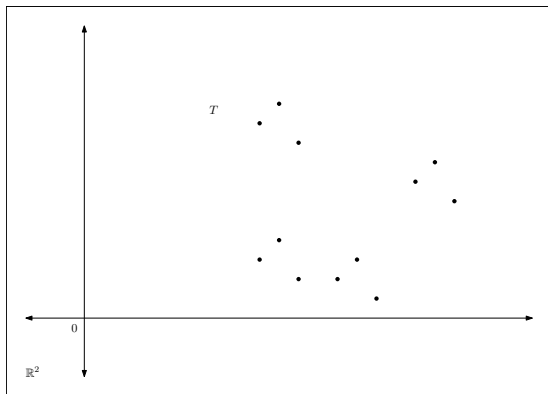


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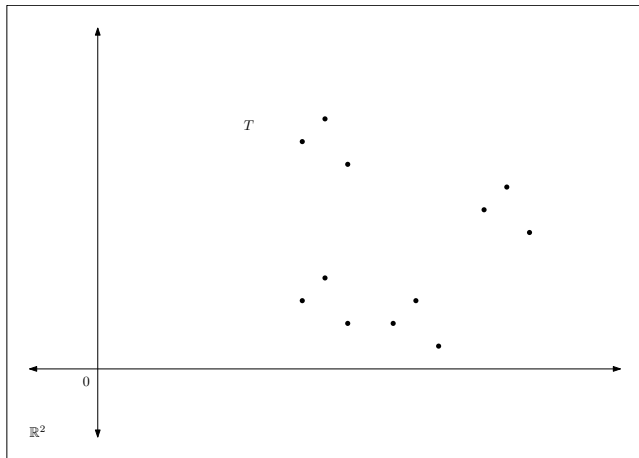
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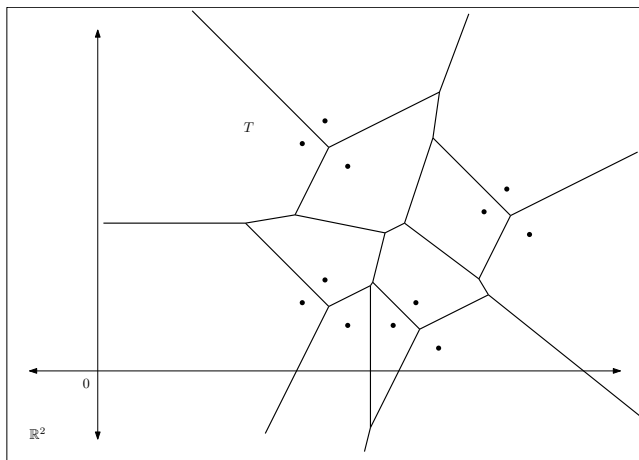
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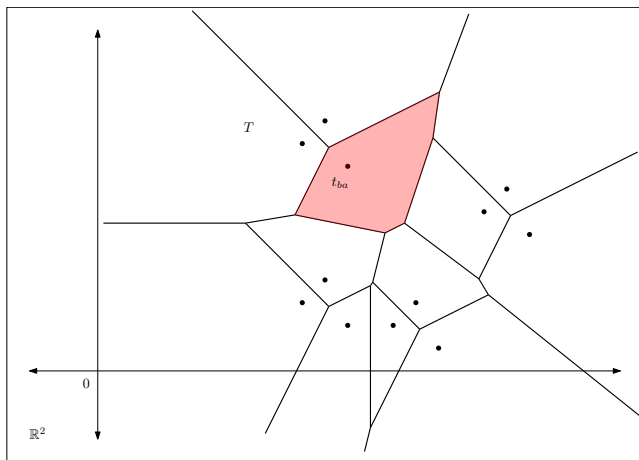
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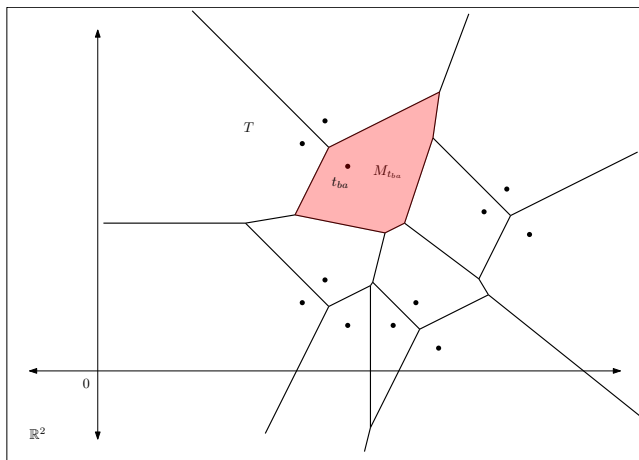
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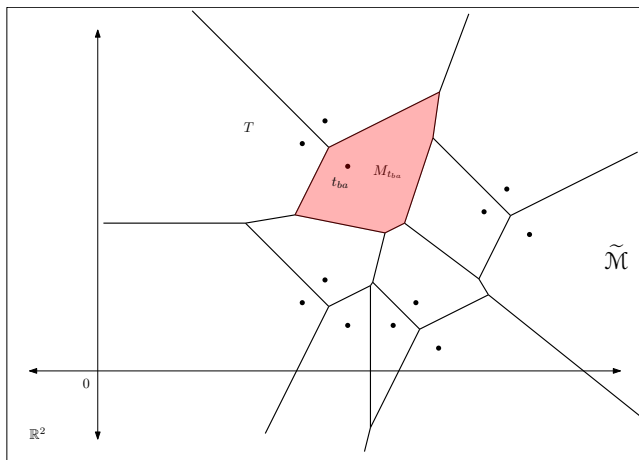


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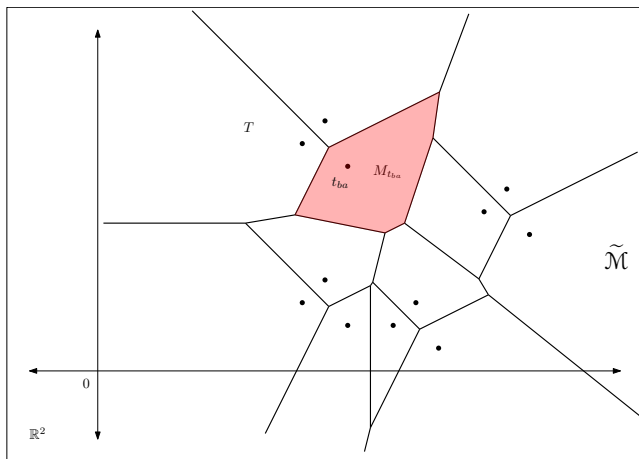
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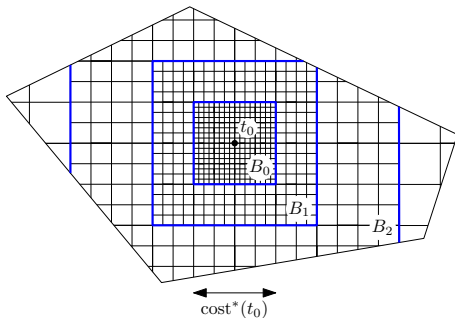
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# $O(1) \rightarrow (1 + \varepsilon)$ approximation



- ▶  $(1 + \varepsilon)$  approximate diagram of size  $O(|T|\varepsilon^{-2} \log \varepsilon^{-1}) = O((mn)\varepsilon^{-2} \log \varepsilon^{-1})$

# Near-linear size: clustering $T$

- ▶ Any  $O(1)$  approx. diagram is enough for the  $(1 + \varepsilon)$  diagram.
- ▶ Partition  $T$  into clusters of size at least  $m/2$ .
- ▶ Reduce diagram size from  $O(mn) \rightarrow O(n)$ .

# Open questions

1. Is the complexity of  $\mathcal{M}$  polynomial?
2. Approximate diagrams for rotations? Rigid transforms?

# The End

Thank you.