

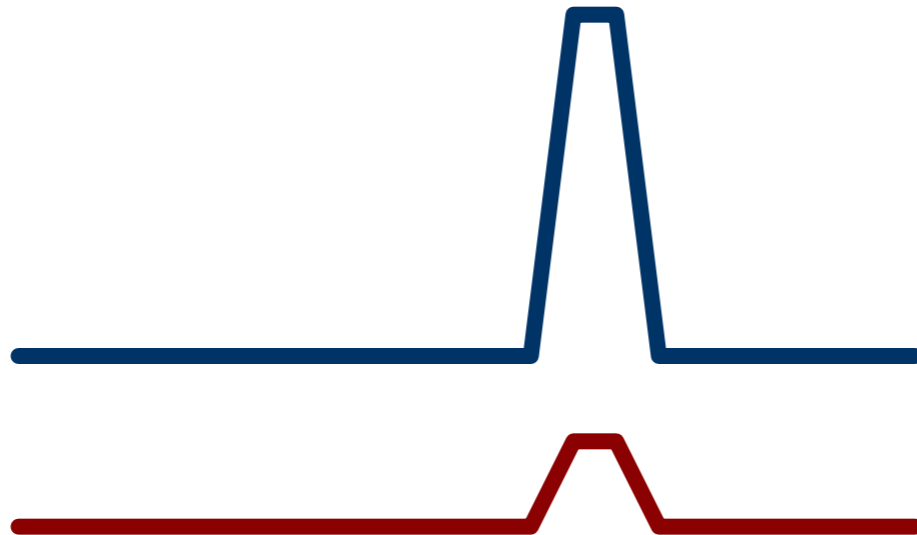
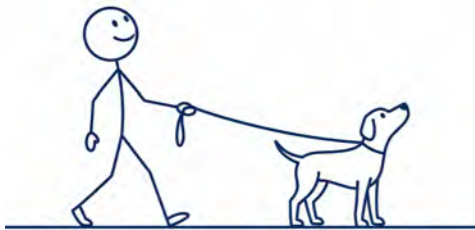
A natural metric for curves – 35 years of Fréchet distance computation

Maïke Buchin

EuroCG 2026, Hagen



Question: How similar are two polygonal curves?



Introduction

An example of a metric for parameterized curves



Maurice Fréchet, 1906

<https://perso.lpsm.paris/mazliak/ConfFrechetang.html>

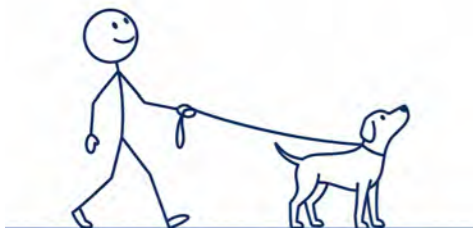
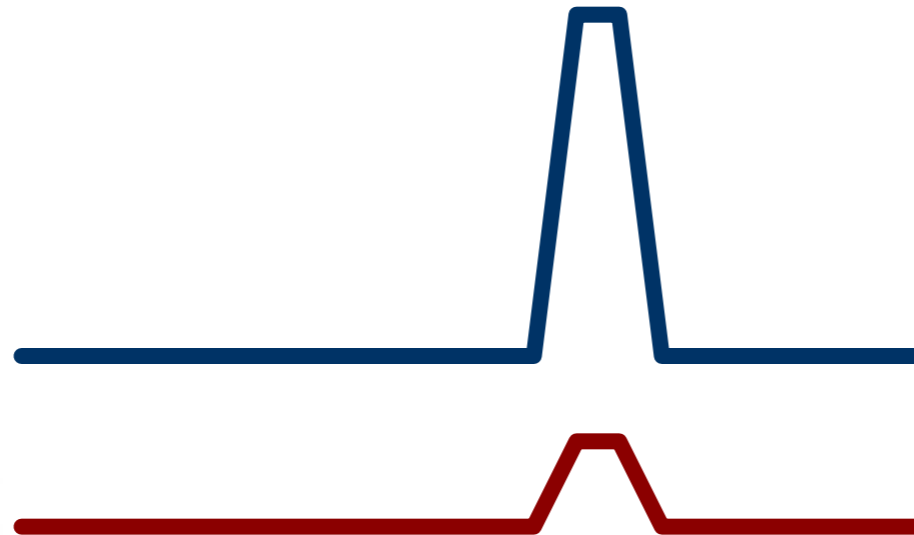
SUR QUELQUES POINTS DU CALCUL FONCTIONNEL;

Par M. Maurice Fréchet (Paris *).

Adunanza del 22 aprile 1906.

INTRODUCTION.

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$P, Q: [0, n] \rightarrow \mathbb{R}^d$ parametrised curves

$$d_F(P, Q) := \inf_{\substack{\sigma: [0, n] \rightarrow [0, n] \\ \text{homeomorphism}}} \max_{x \in [0, n]} d(P(x), Q(\sigma(x)))$$



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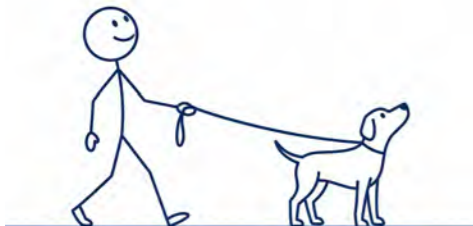
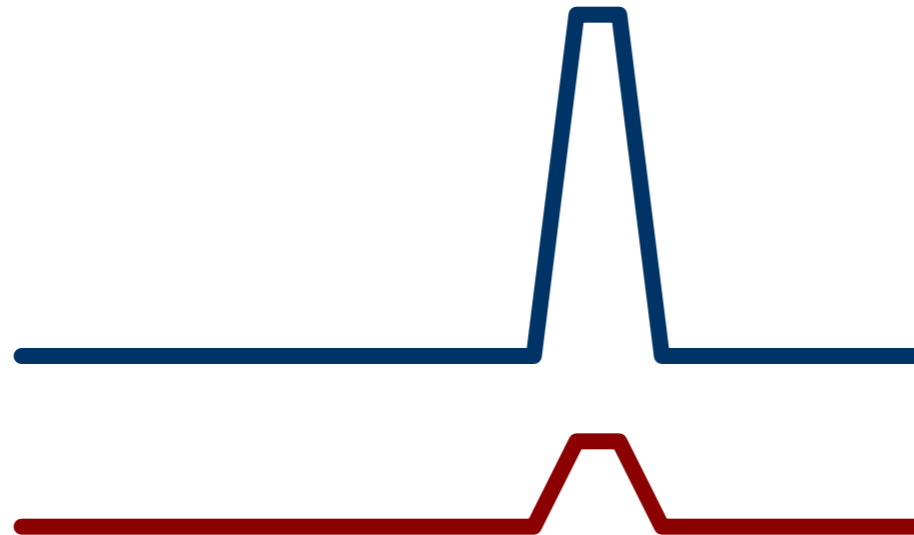
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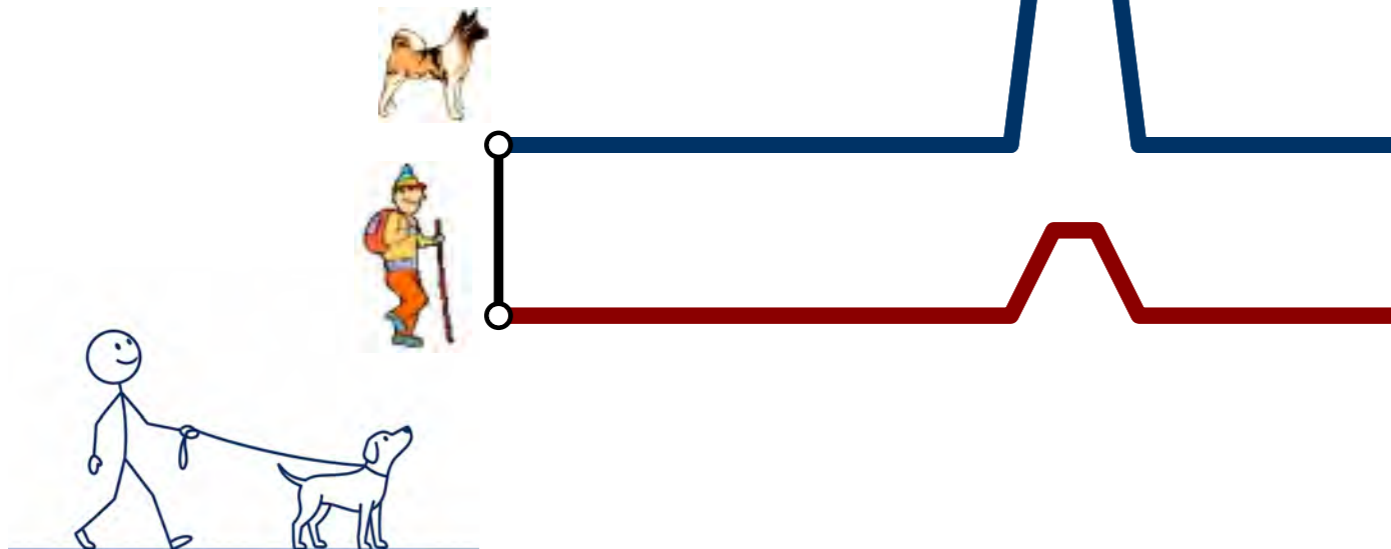
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Illustration: Person & Dog



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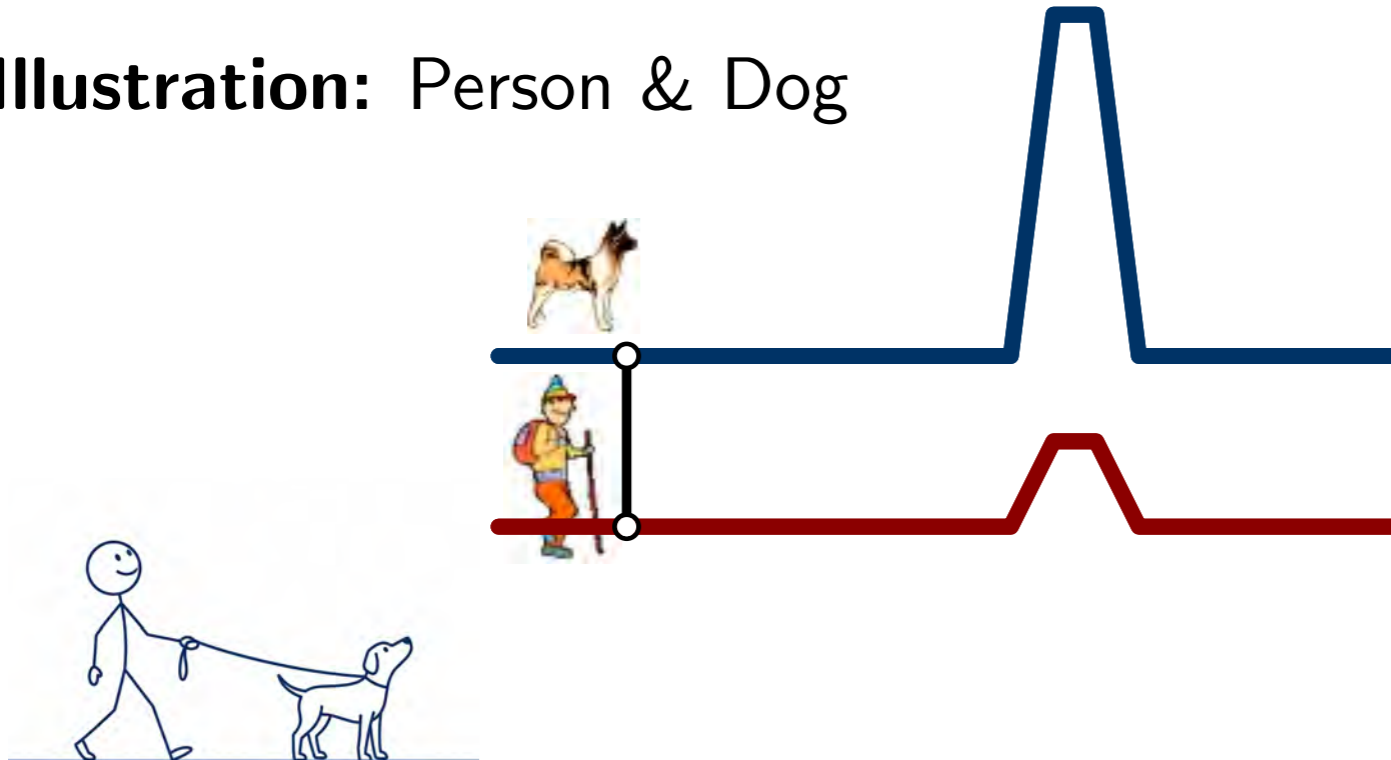
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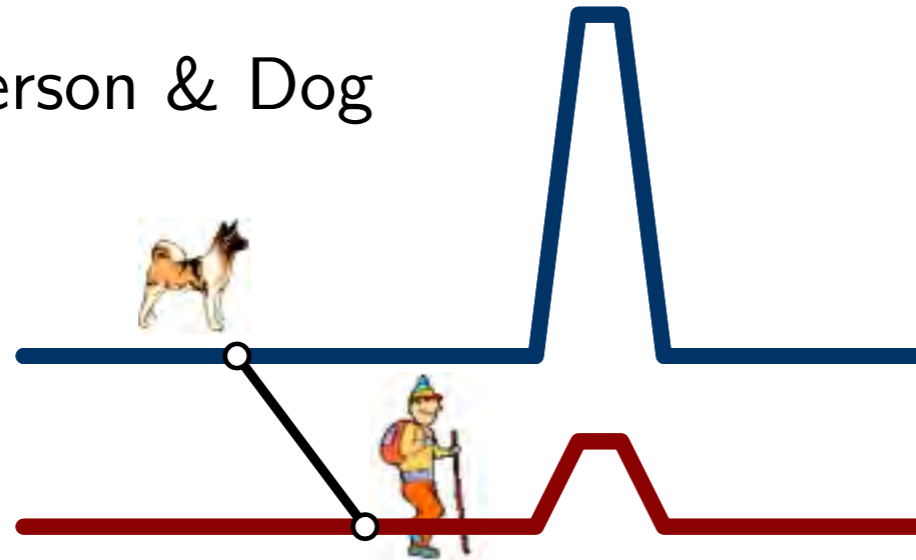
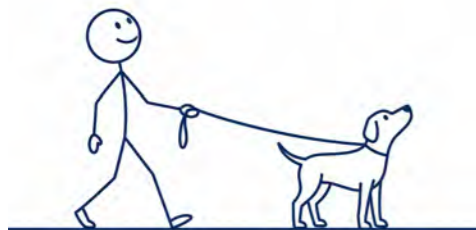
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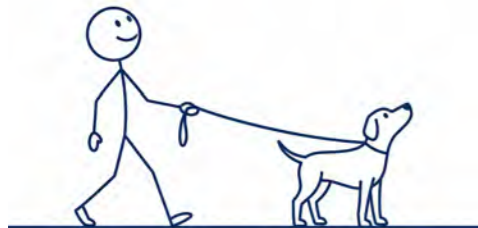
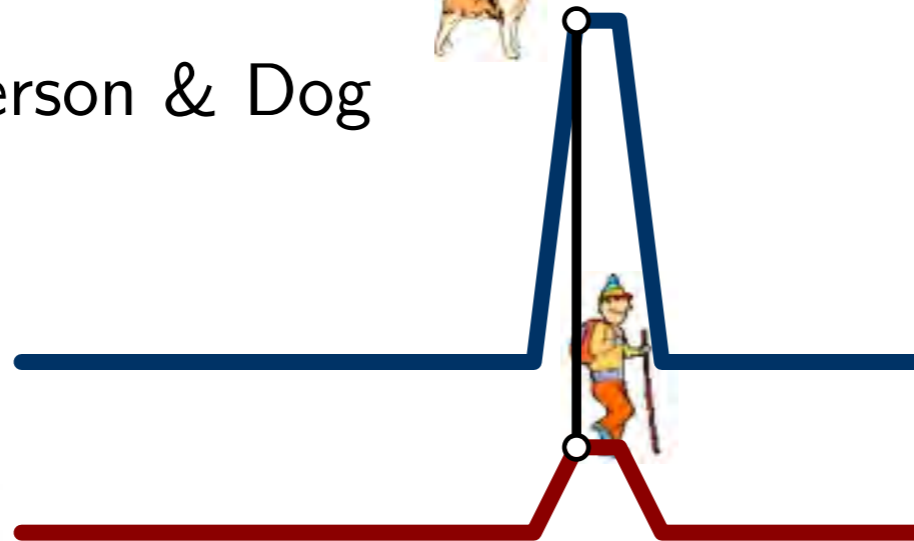
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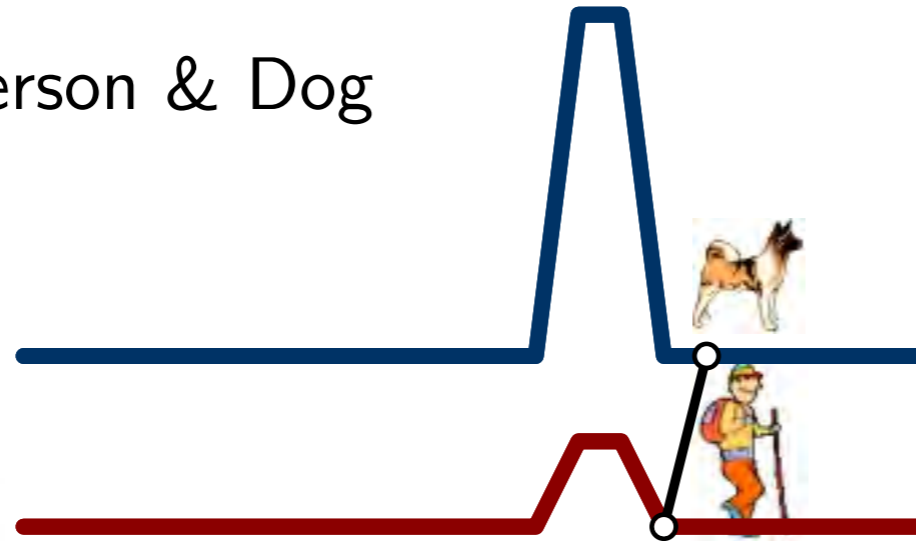
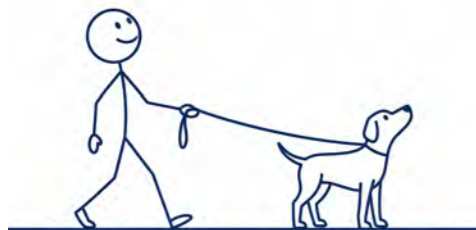
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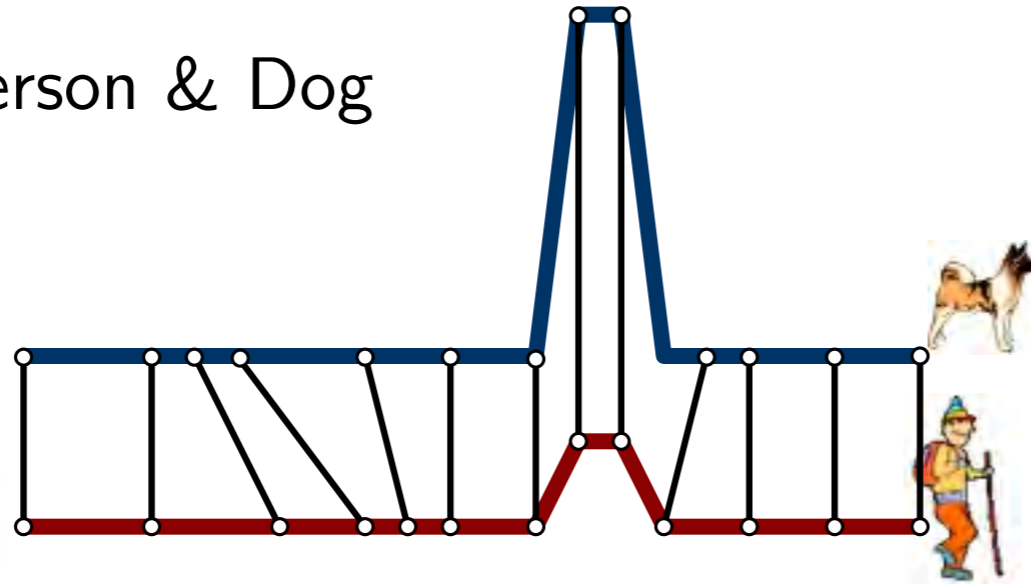
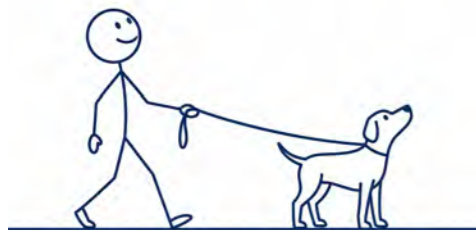
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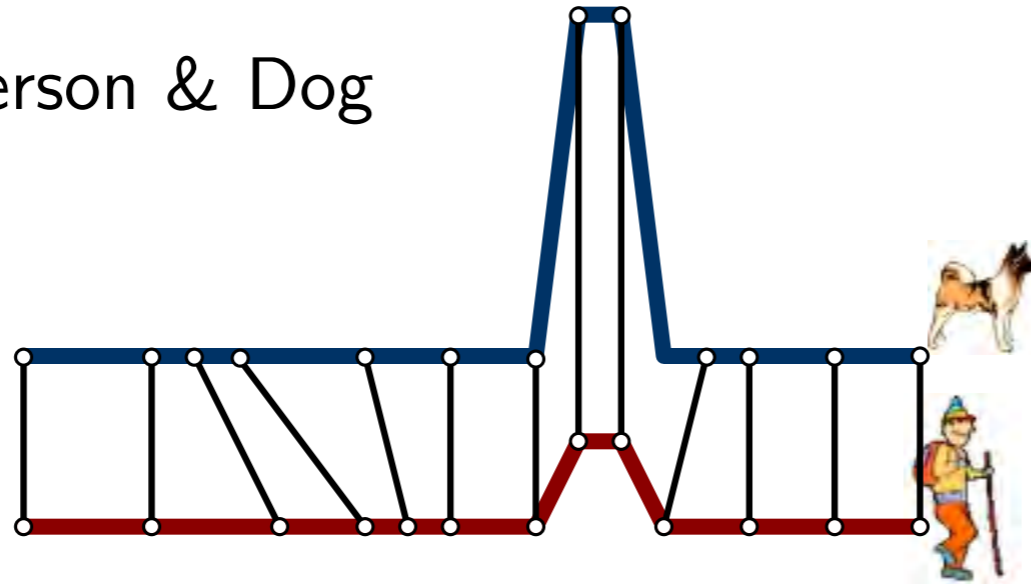
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Fréchet distance equals shortest leash length

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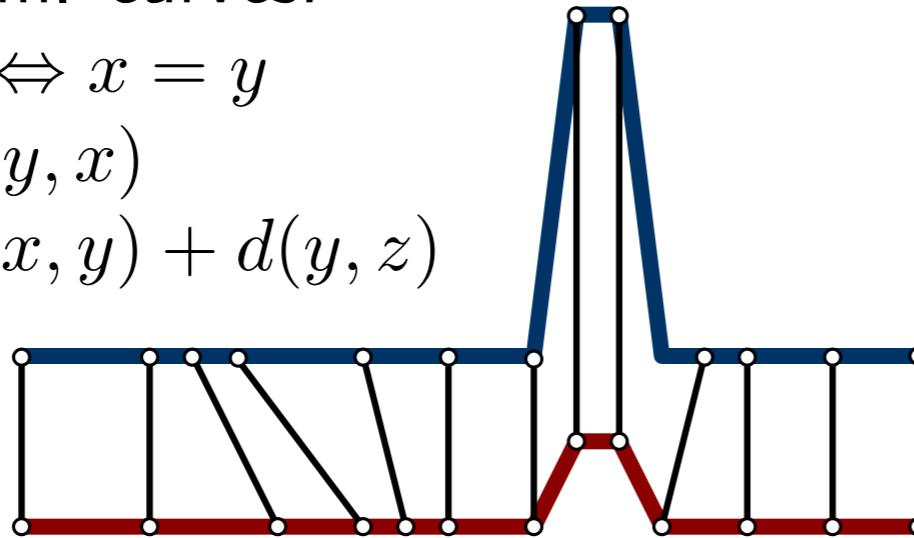


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metric for param. curves:

- $d(x, y) = 0 \Leftrightarrow x = y$
- $d(x, y) = d(y, x)$
- $d(x, z) \leq d(x, y) + d(y, z)$



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Research on computing the Fréchet distance

in 1992 Helmut Alt and Michael Godau presented the first algorithm

A natural metric for curves — Computing the distance for polygonal chains and approximation algorithms

Michael Godau*
godau@leibniz.math.fu-berlin.de
Freie Universität Berlin, FB Mathematik, WE 03
Arnimallee 2-6, D-1000 Berlin 33

COMPUTING THE FRÉCHET DISTANCE BETWEEN TWO POLYGONAL CURVES *

HELMUT ALT and MICHAEL GODAU
Freie Universität Berlin, Fachbereich Mathematik und Informatik, Takustraße 9
14195 Berlin, Germany
e-mail: alt@inf.fu-berlin.de

Measuring the Resemblance of Polygonal Curves *

(Extended Abstract)

Abstract

The often explored problem to approximate a given polygonal chain by a constant factor worse than the global optimum. The often explored problem to approximate a given polygonal chain considered from a computational geometric point of view or model it reasonably we give a natural definition of the distance. Furthermore we give algorithms to calculate this distance chains in the d -dimensional space for arbitrary d . With known polynomial time algorithms to approximate polygonal chains an optimal solution under some constraints. We will show that by a constant factor worse than the global optimum.

1 Introduction, Definitions

In many applications, two-dimensional shapes are given by the planar curves forming their boundaries. Consequently, a natural problem in shape comparison and recognition is to measure, how much two given curves “resemble each other”. Naturally, the first question to be answered, is what metric between curves should be used to reflect the intuitive notion of “resemblance”.

Helmut Alt

Michael Godau †



Figure 1: Two curves with HAUSDORFF-distance δ .

In arbitrary dimensions we consider parametrizations of the curves. An algorithm of runtime $O(n \log n)$ and Q is developed. Then the distance for closed curves, on derived from the Fréchet-curve Q .

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The often explored problem to approximate a given polygonal chain is considered from a computational geometric point of view or modeled reasonably. We give a natural definition of the distance between two polygonal chains in the d -dimensional space for arbitrary d . Furthermore we give algorithms to calculate this distance. With known polynomial time algorithms to approximate polygonal chains an optimal solution under some constraints. We will show that the approximation is by a constant factor worse than the global optimum.

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In arbitrary dimensions we consider polygonal chains. We give algorithms of runtime $O(n^3)$ and $O(n^2)$ respectively. Then the Hausdorff-distance for closed curves, on derived from the Fréchet-curve Q .



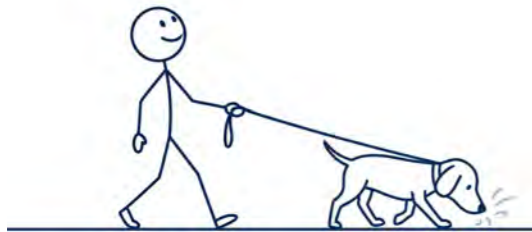
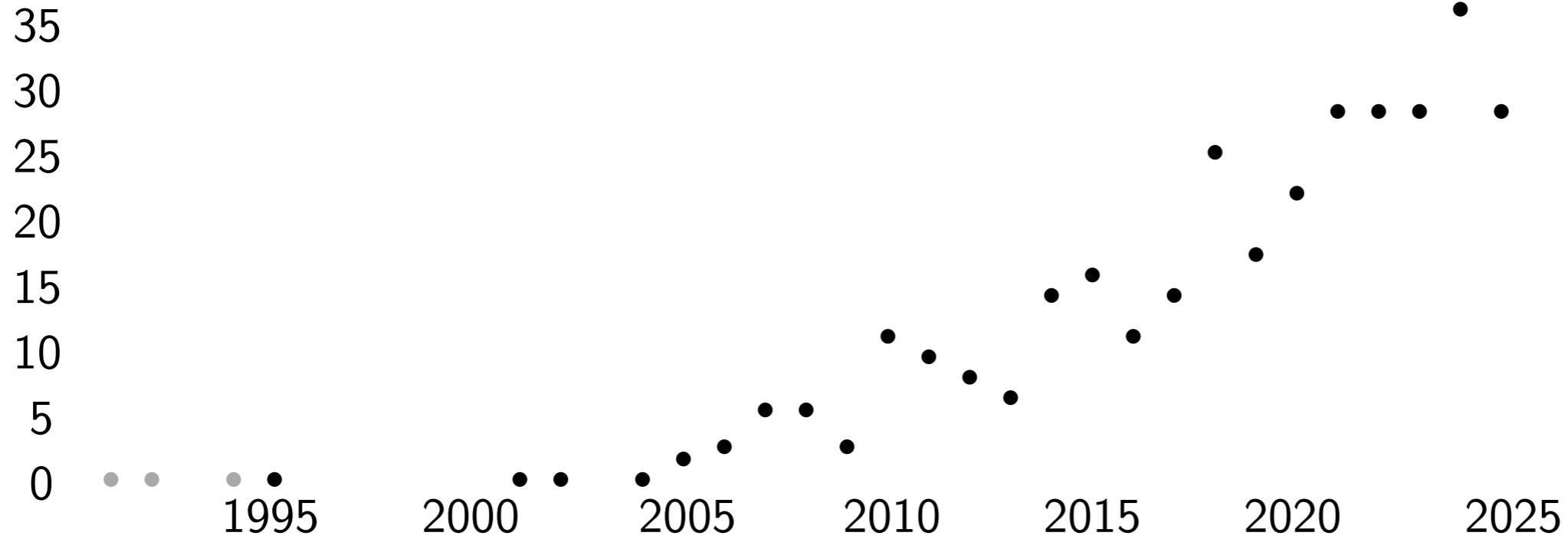
Figure 1: Two curves with HAUSDORFF-distance δ .



and many more papers followed..

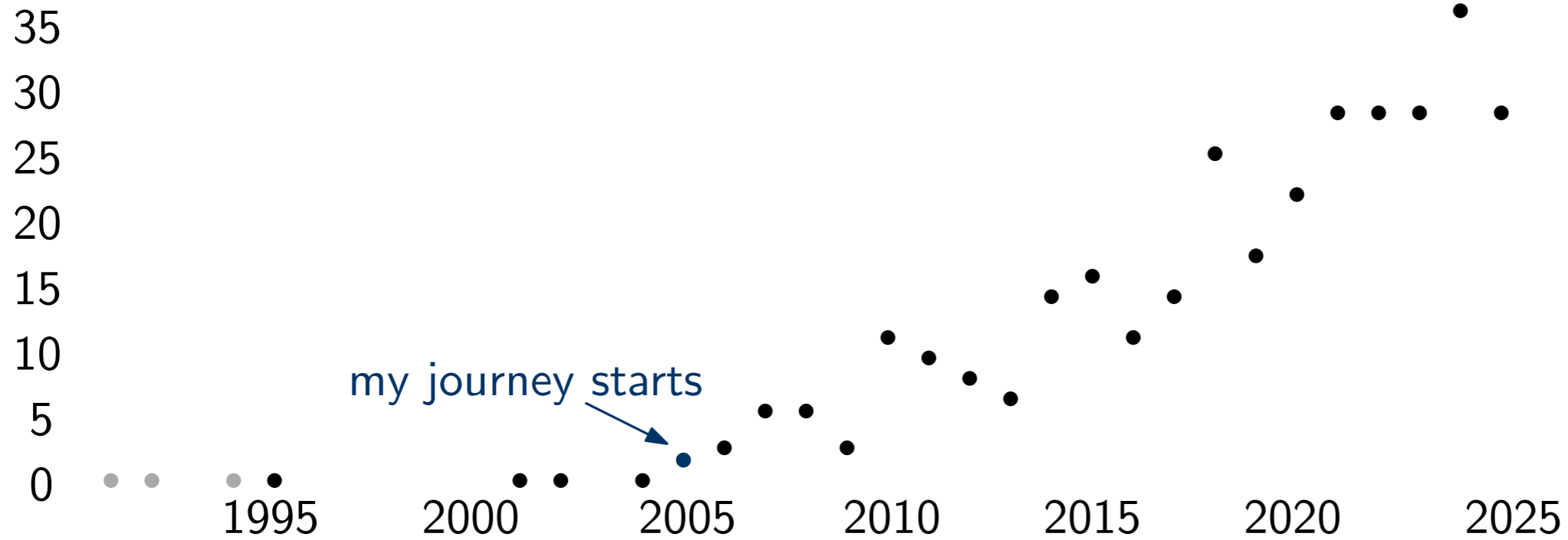
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number of papers on dblp



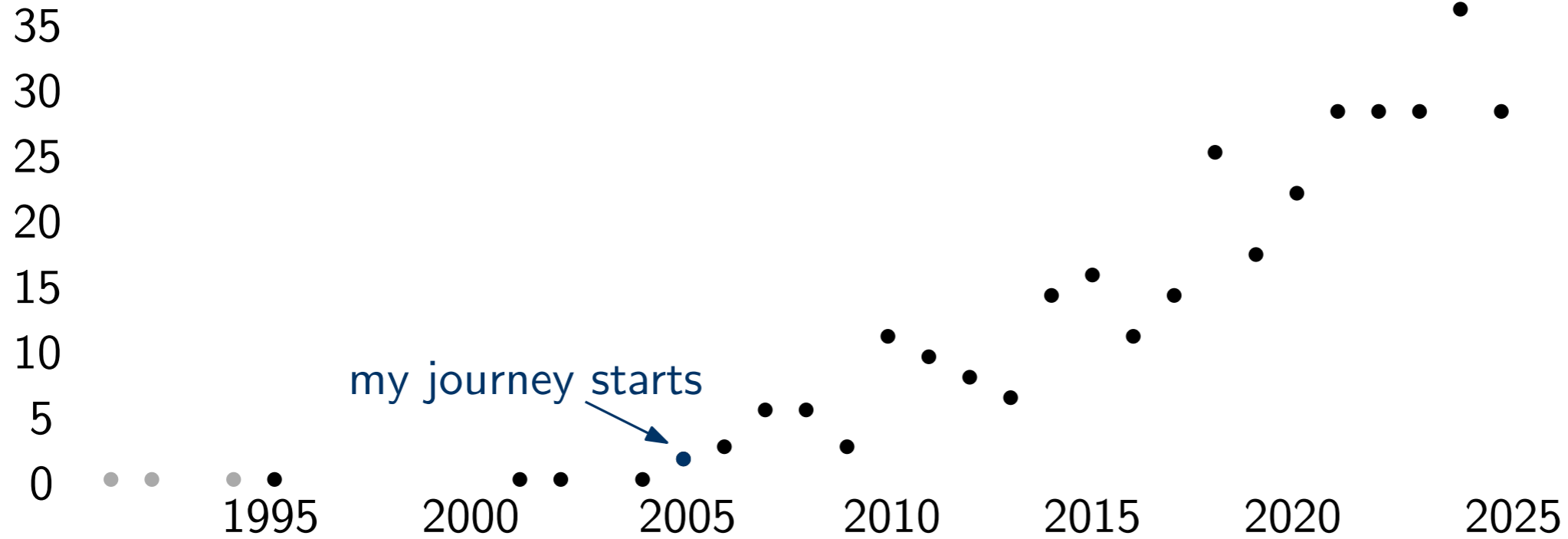
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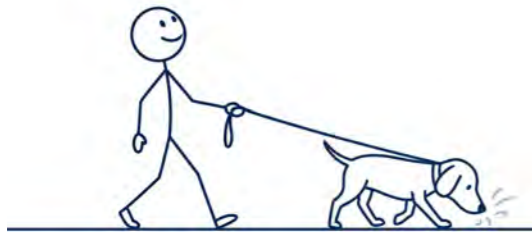


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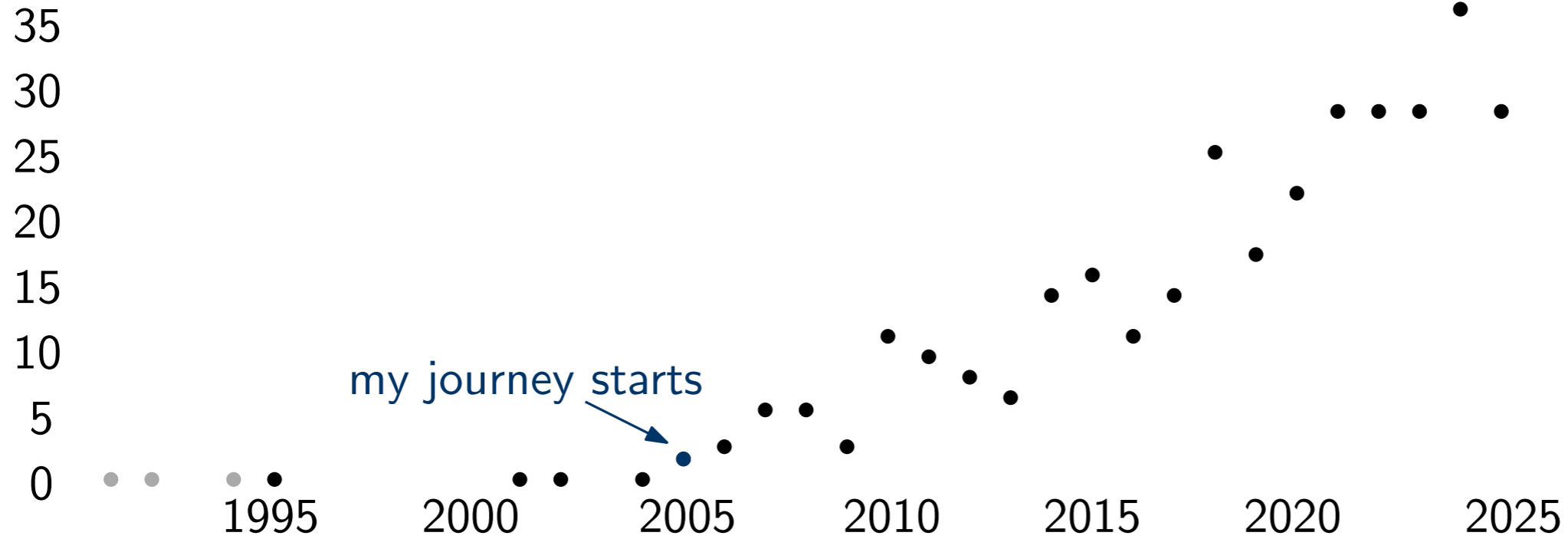


why so popular?

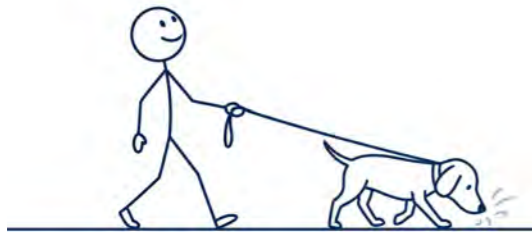


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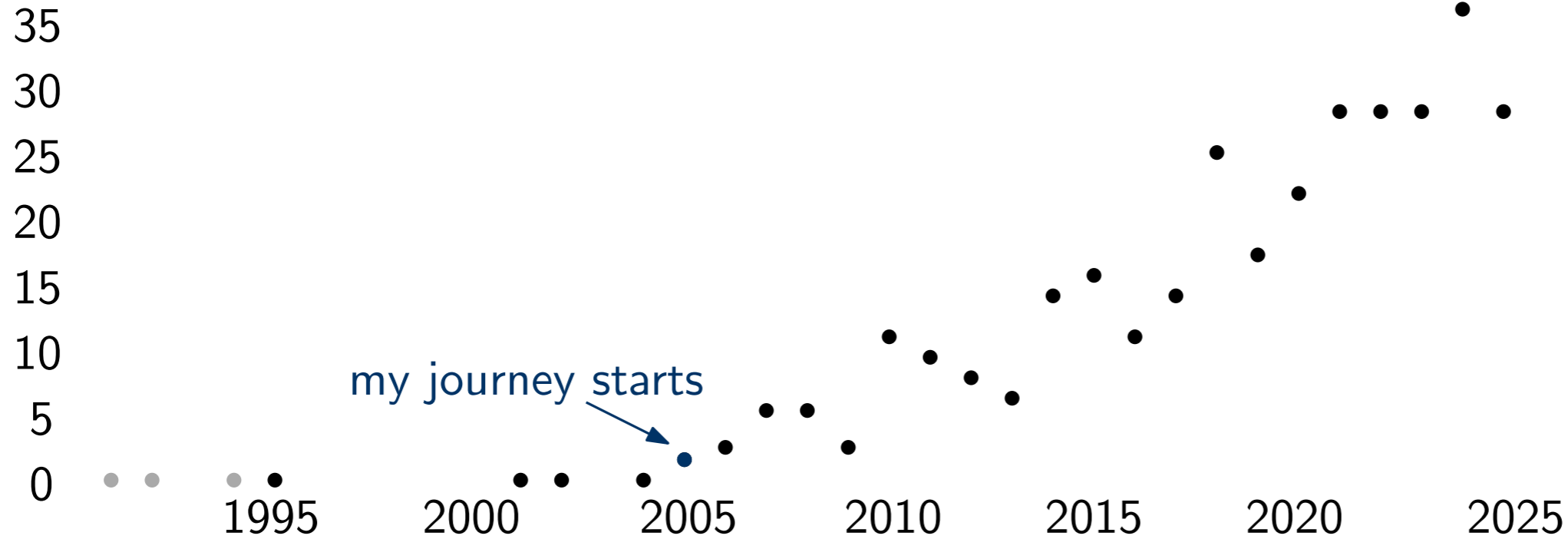


why so popular? because it's *natural*, *versatile* and *challenging* to compute!



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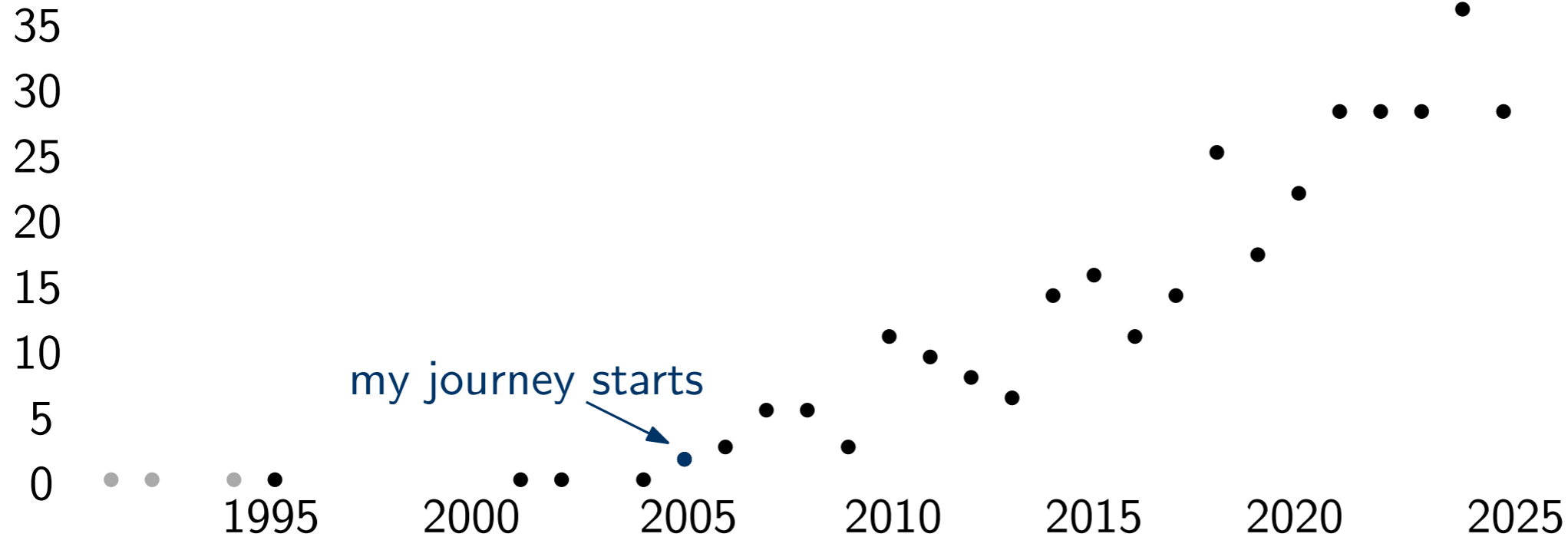
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Plan for this talk:

- some personal highlights
- open problems

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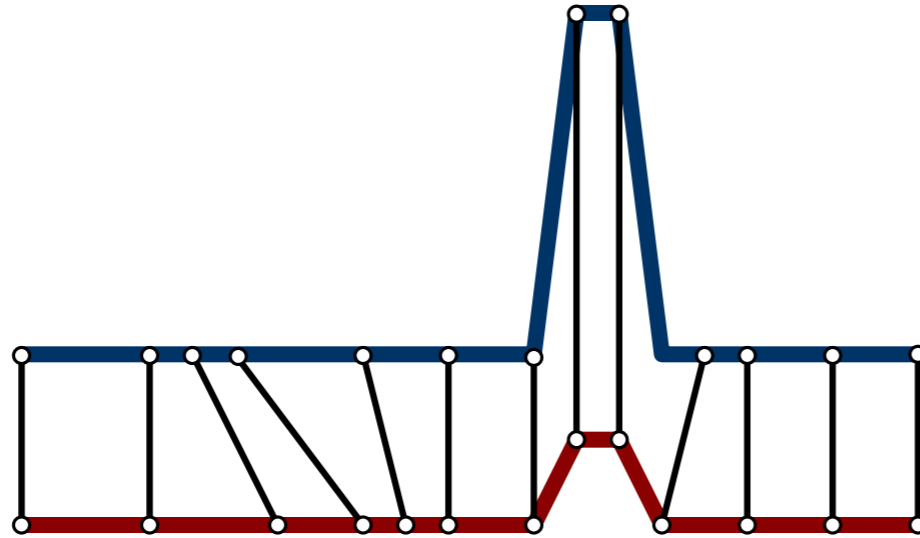
Disclaimer:

only continuous (not discrete) Fréchet distance

Algorithm of Alt & Godau

Let P, Q be two polygonal curves of size n .

- **decision problem:** Is $d_F(P, Q) \leq \epsilon$?
- **computation problem:** What is $d_F(P, Q)$?

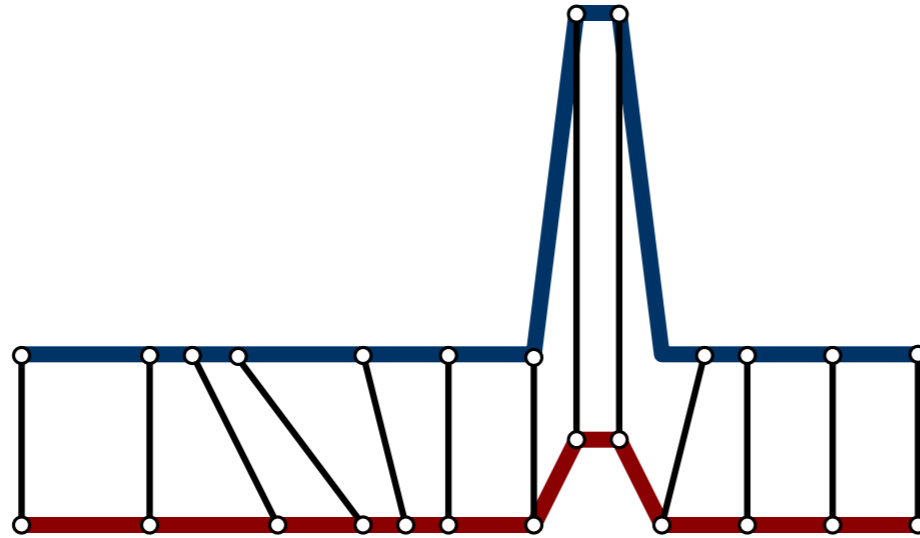


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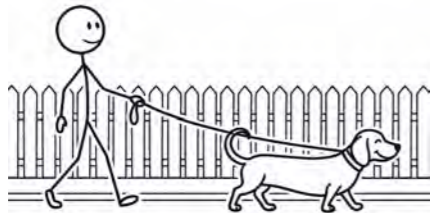
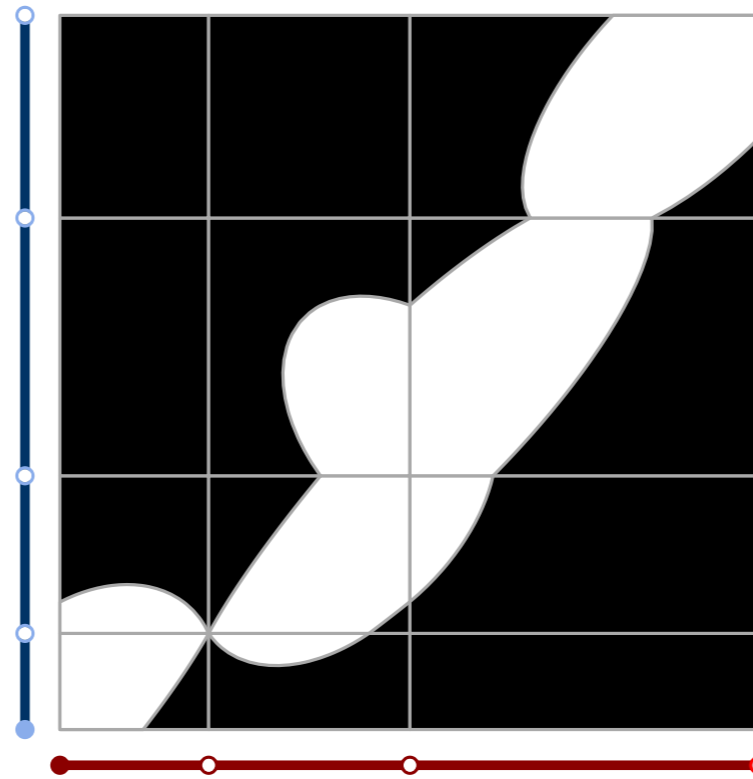
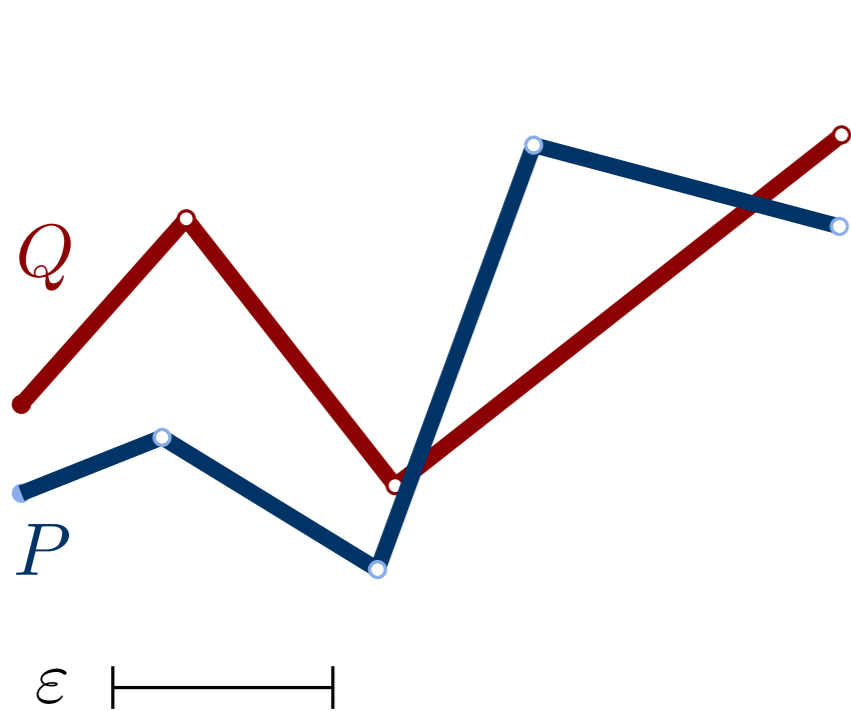
use the *Free Space Diagram* for the decision problem



Free Space Diagram

Let $P, Q: [0, n] \rightarrow \mathbb{R}^d$ polygonal curves, and let $\varepsilon > 0$.

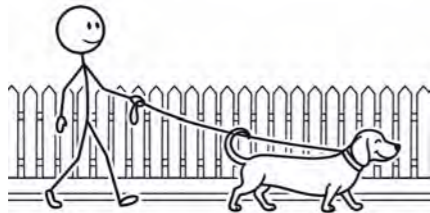
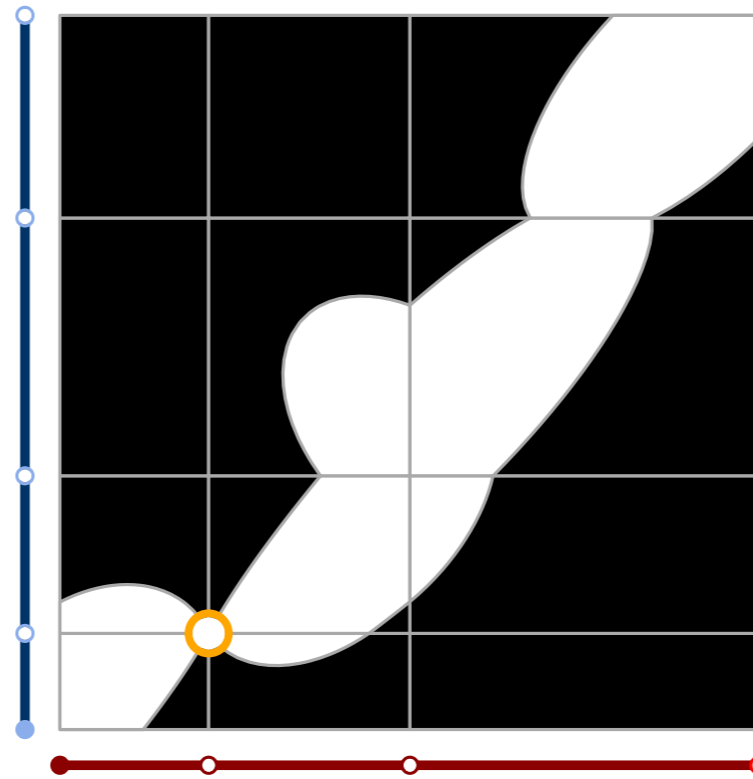
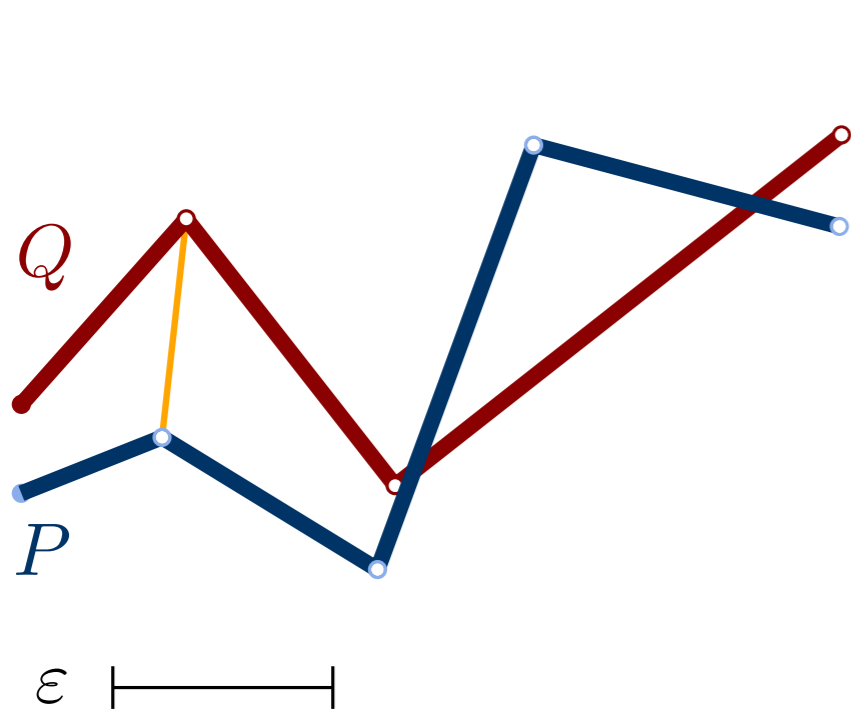
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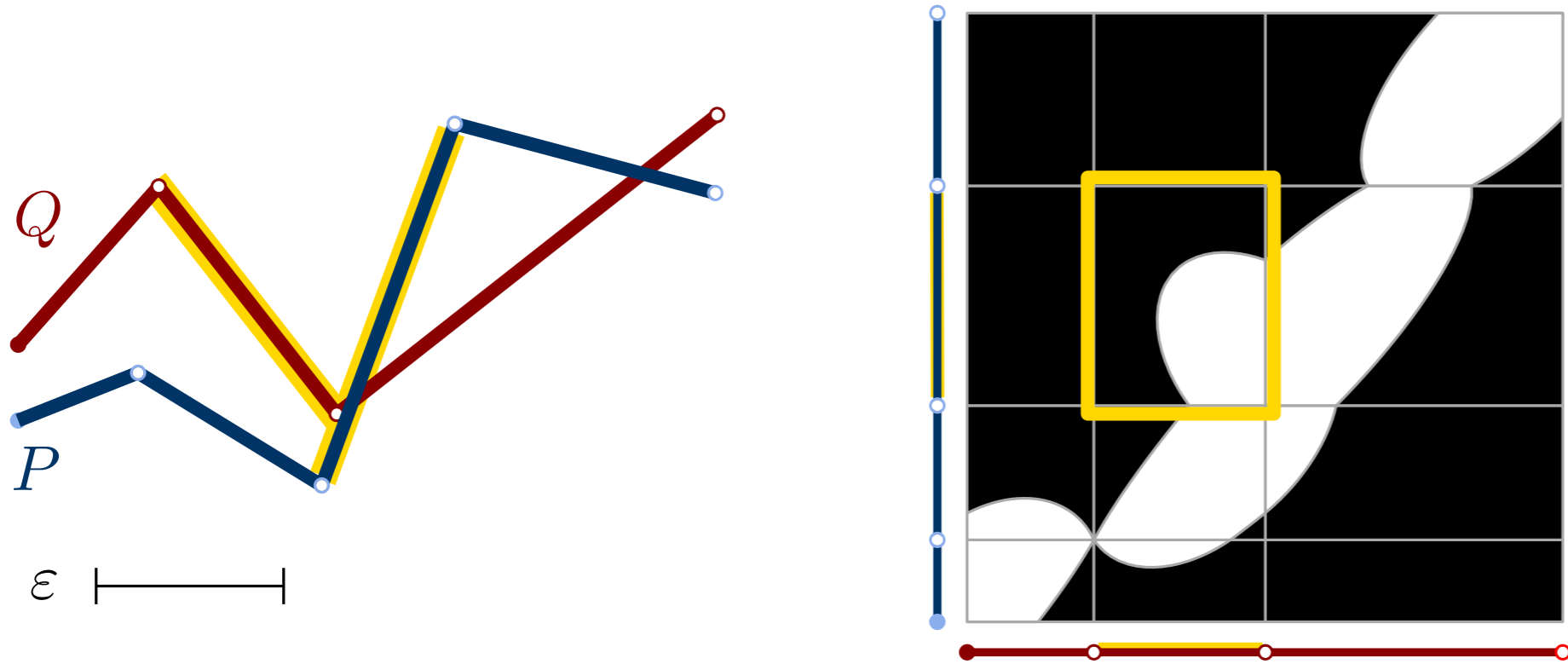
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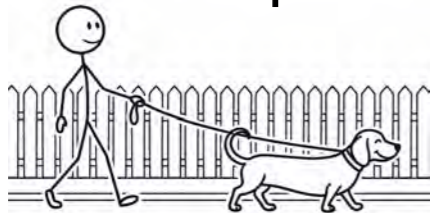
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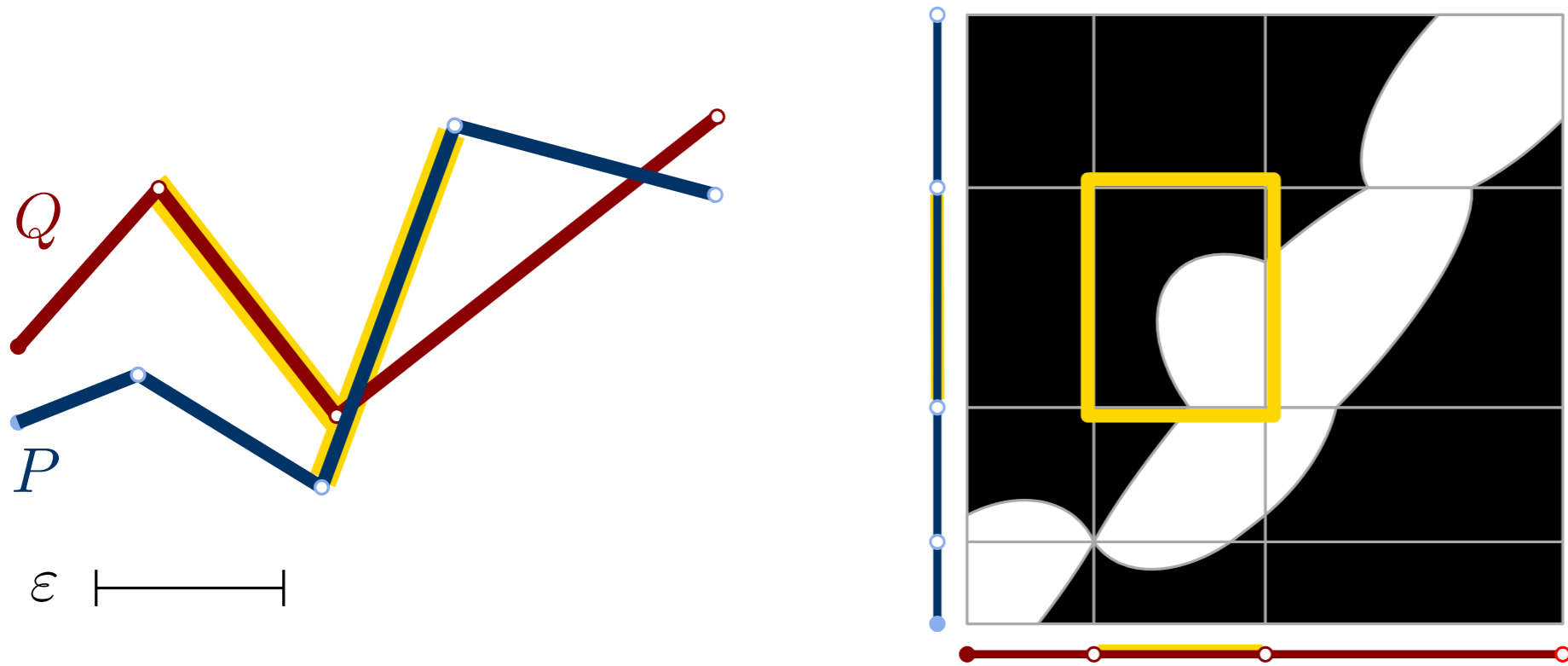
Cell: free space of two segments of the curves



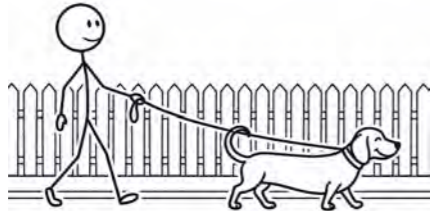
Free Space Diagram

Let $P, Q: [0, n] \rightarrow \mathbb{R}^d$ polygonal curves, and let $\varepsilon > 0$.

Define $F_\varepsilon(P, Q) := \{(s, t) \mid d(P(s), Q(t)) \leq \varepsilon\}$



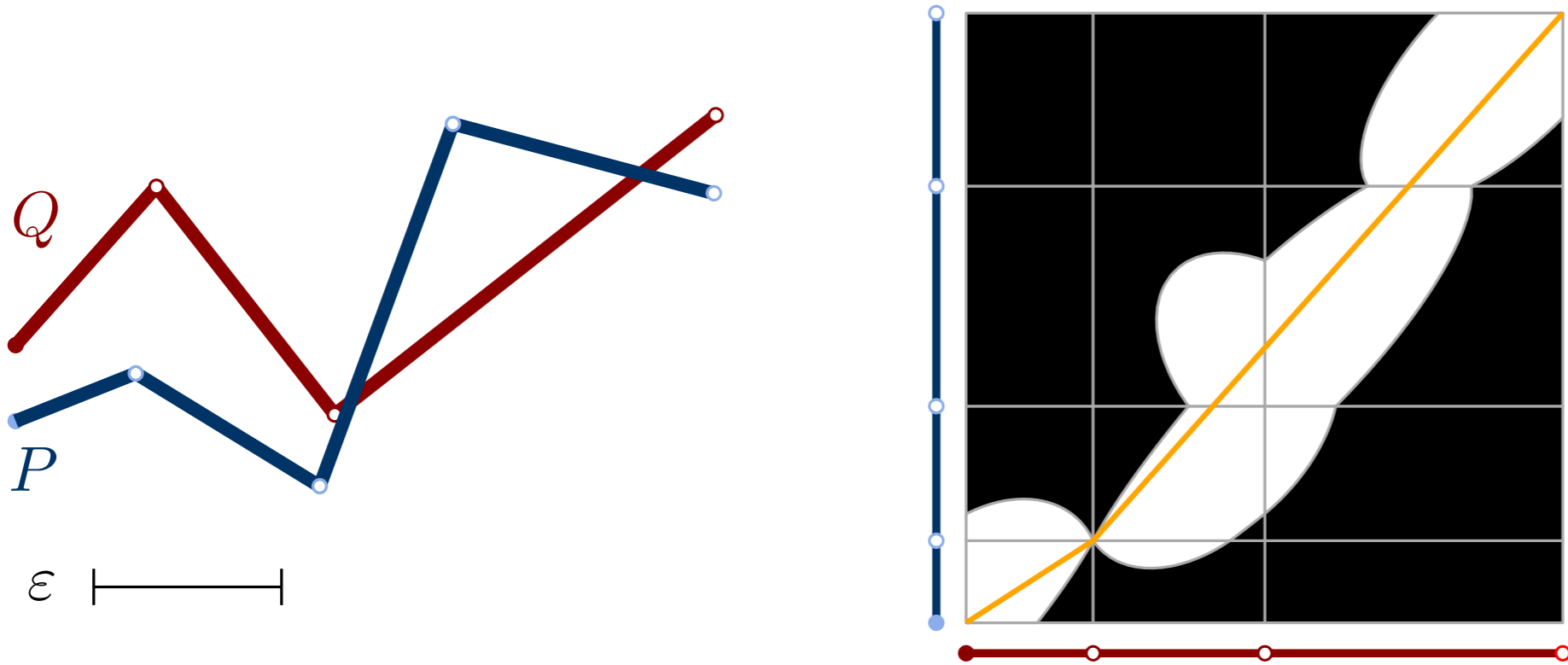
Lemma 1: Cells of the free space are convex.



Free Space Diagram

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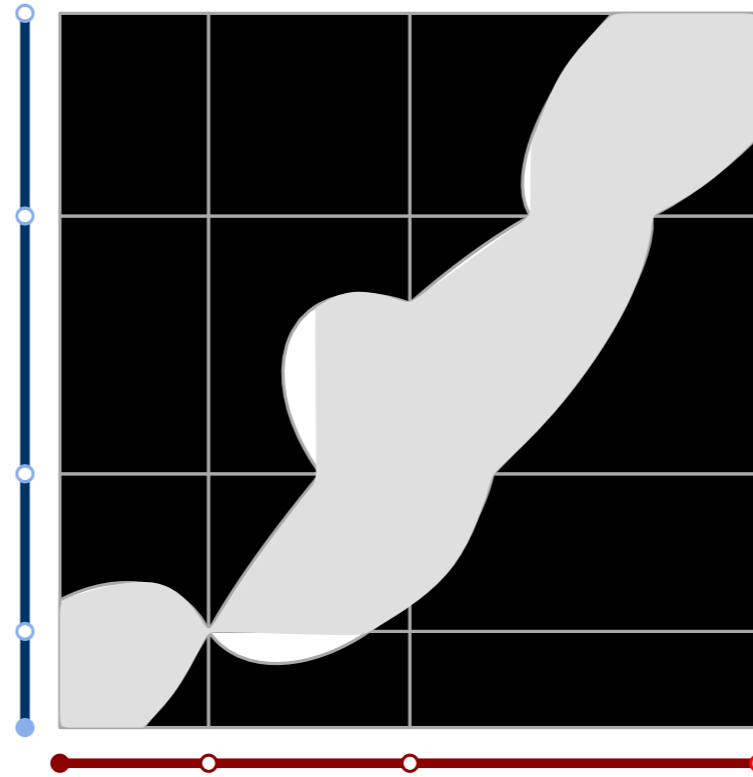
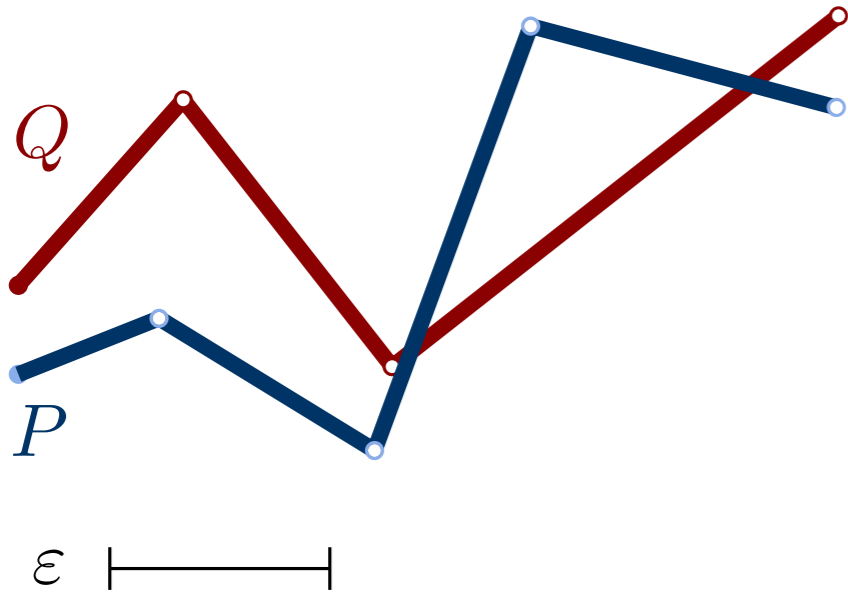
Lemma 1: Cells of the free space are convex.

Lemma 2: $d_F(P, Q) \leq \varepsilon \Leftrightarrow$ there exists a monotone path from $(0, 0)$ to (n, n) in the free space $F_\varepsilon(P, Q)$.

Free Space Diagram

Let $P, Q: [0, n] \rightarrow \mathbb{R}^d$ polygonal curves, and let $\varepsilon > 0$.

Define $F_\varepsilon(P, Q) := \{(s, t) \mid d(P(s), Q(t)) \leq \varepsilon\}$



reachable free space: subset of $F_\varepsilon(P, Q)$ that is reachable by a monotone path from $(0, 0)$

Decision Algorithm

Input: Polygonal curves P, Q , and $\varepsilon > 0$

Output: Is $d_F(P, Q) \leq \varepsilon$

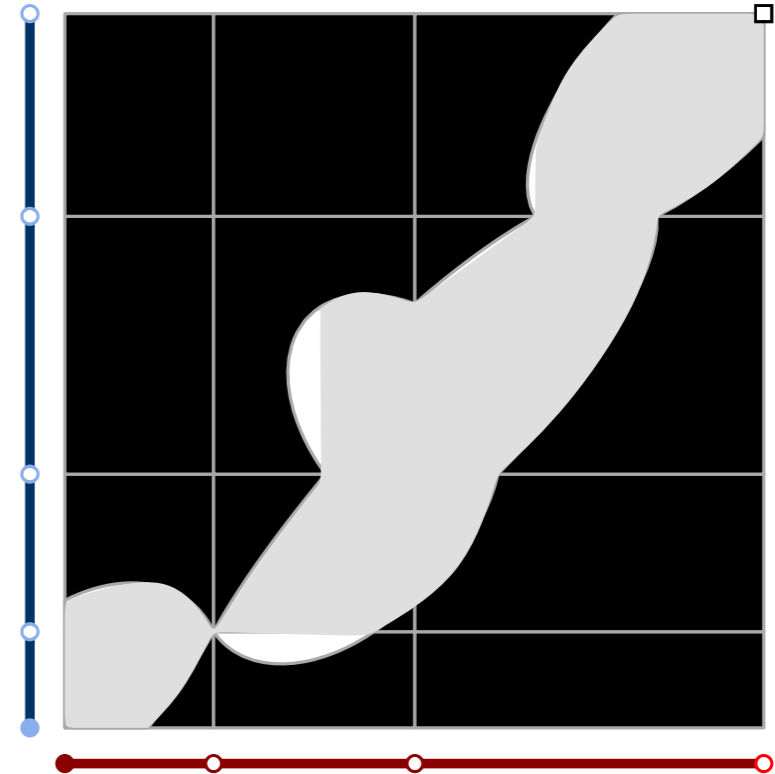
compute lower and left reachable
boundaries of the diagram

for $i \leftarrow 1$ **to** n **do**

for $j \leftarrow 1$ **to** n **do**

 compute upper and right reachable
 boundaries of cell (i, j)

output **true** if (n, n) reachable, else **false**



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Input: Polygonal curves P, Q , and $\varepsilon > 0$

Output: Is $d_F(P, Q) \leq \varepsilon$

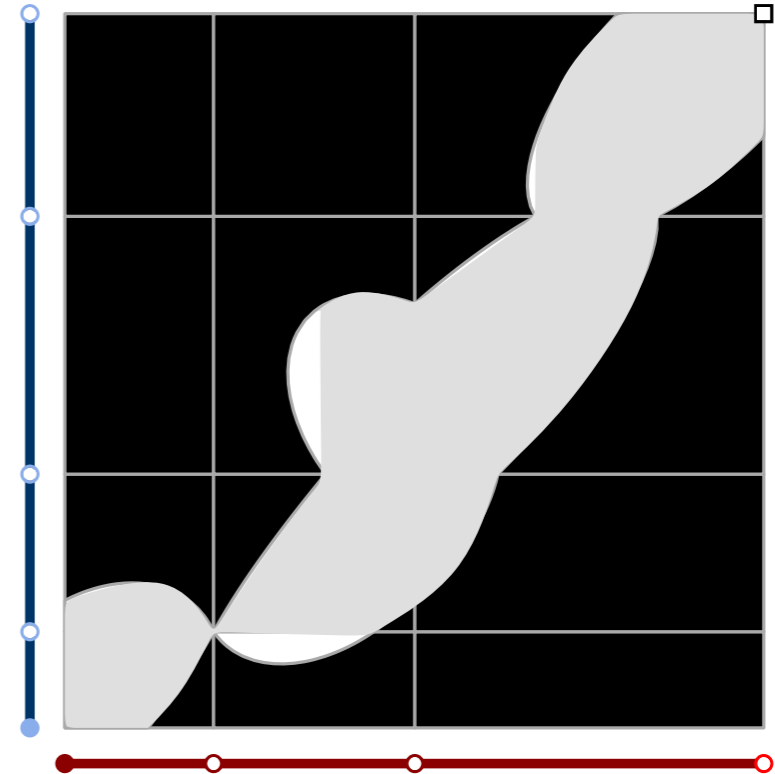
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Theorem (Alt, Godau)

For two polygonal curves P, Q of size n and $\varepsilon > 0$,

$d_F(P, Q) \leq \varepsilon$ can be decided in $O(n^2)$ time and $O(n)$ space.

Computation Algorithm

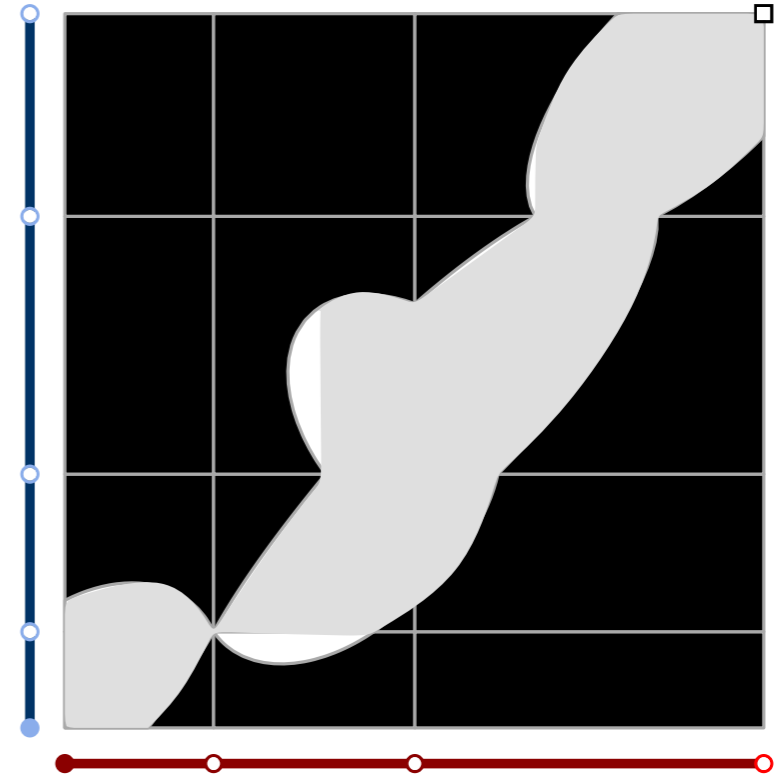
Input: Polygonal curves P, Q

Output: $d_F(P, Q)$

determine all critical values

sort the critical values

do a **binary search** over the values,
solving the decision problem in each step



critical values:

values of ε where reachability
in the free space changes

Computation Algorithm

Input: Polygonal curves P, Q

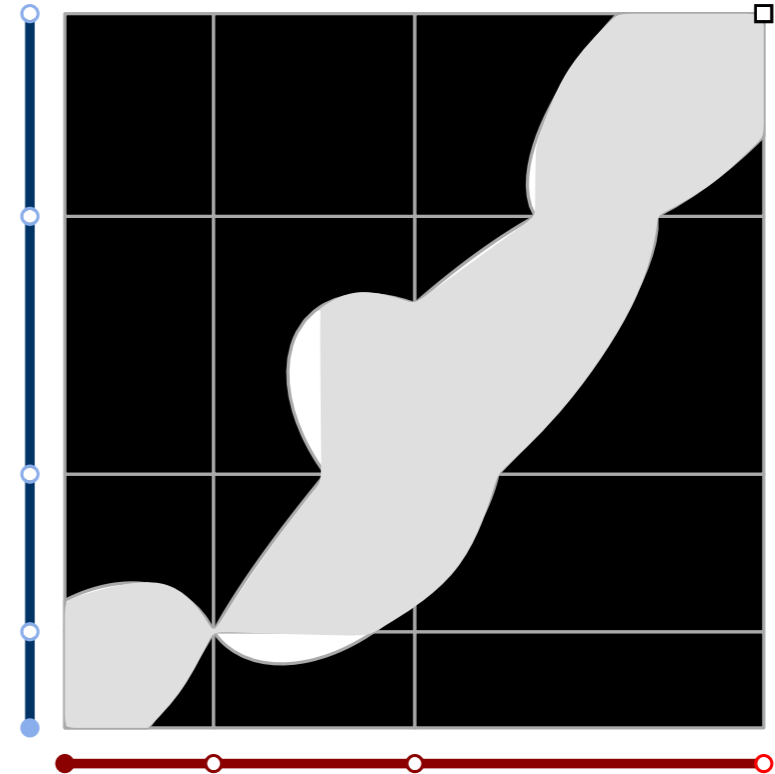
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Computation Algorithm

Input: Polygonal curves P, Q

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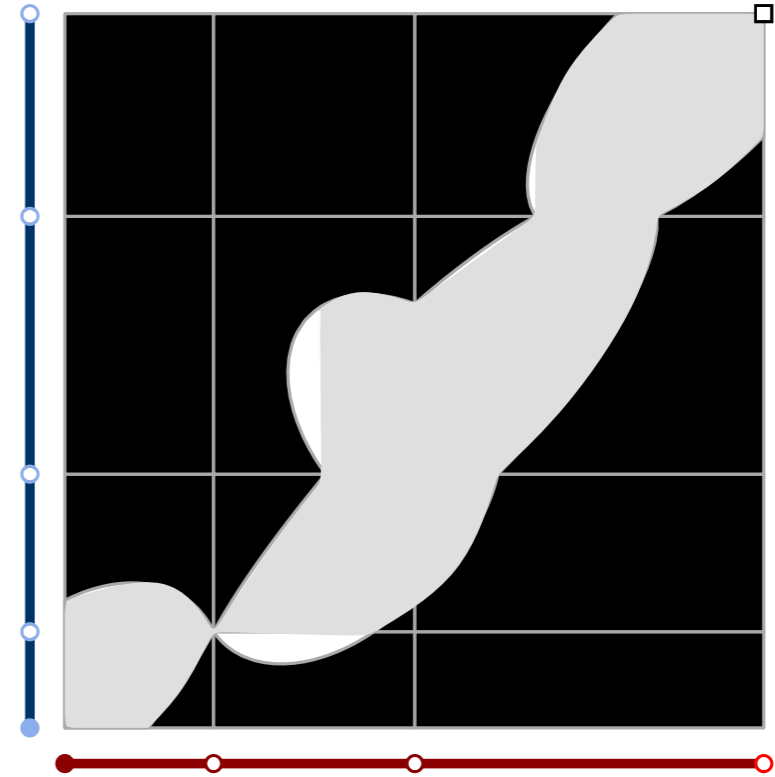
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Runtime: $O(n^3 \log n)$

faster: with *parametric search* in $O(n^2 \log n)$ time



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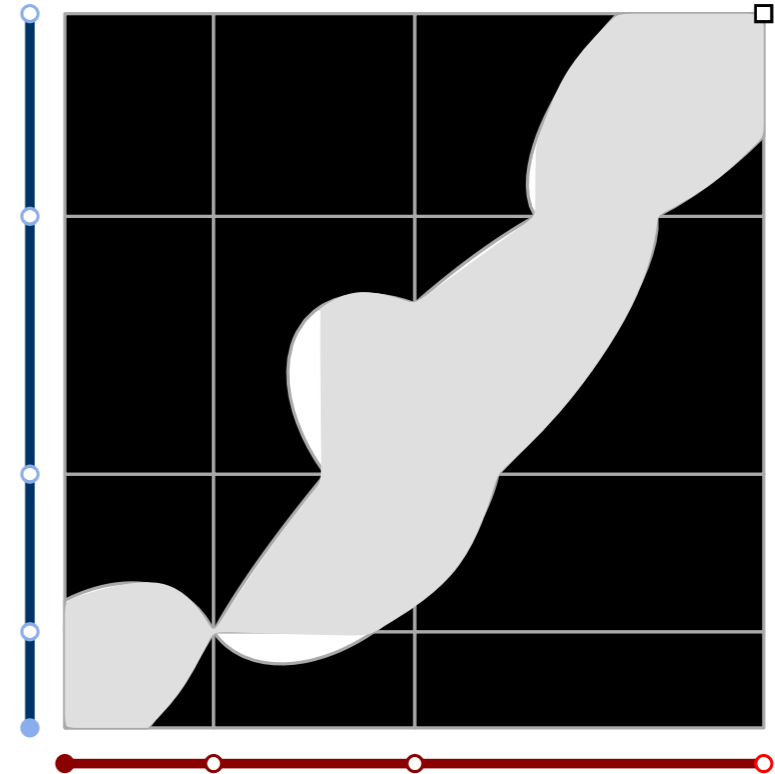
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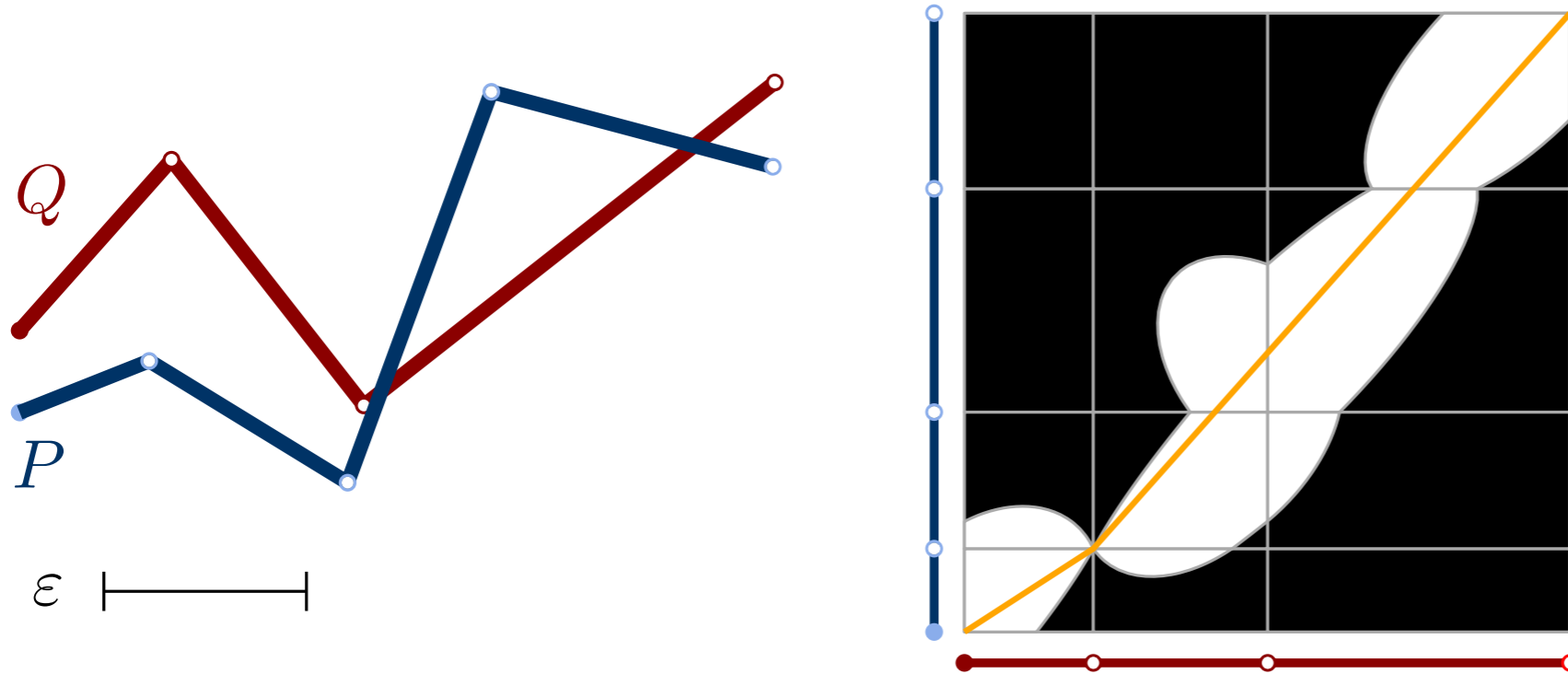
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Questions:

- **faster?**
- implementation?
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- applications?

Breaking the quadratic runtime

Can the Fréchet distance be computed in subquadratic time?

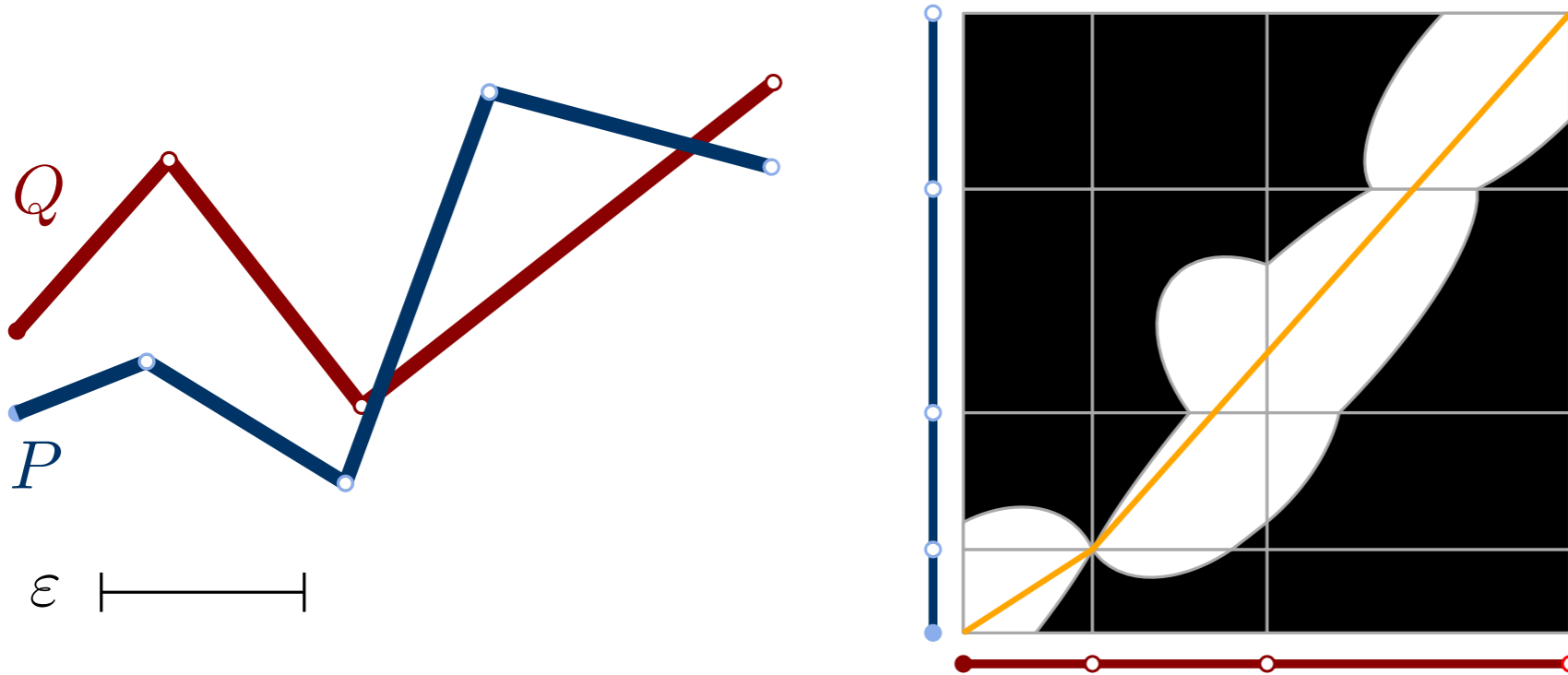


Bottleneck: size of the free space diagram

Breaking the quadratic runtime

Can the Fréchet distance be computed in subquadratic time?

- conditional lower bound by Bringmann (2014)
- slightly faster algorithm by Buchin et al. (2014)
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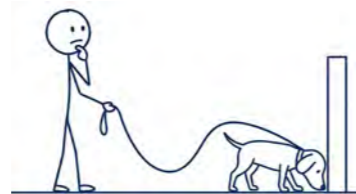
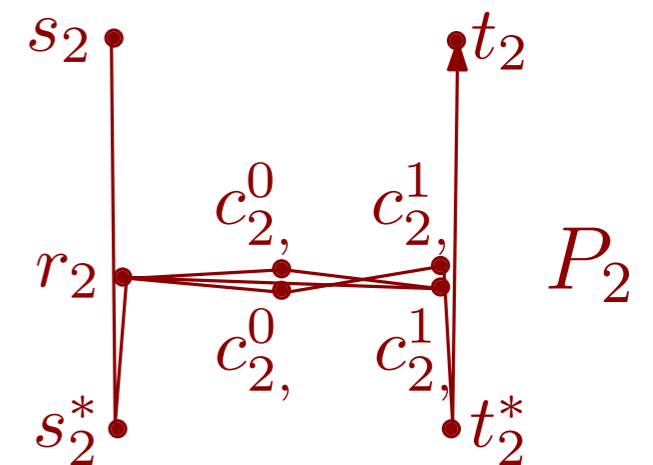
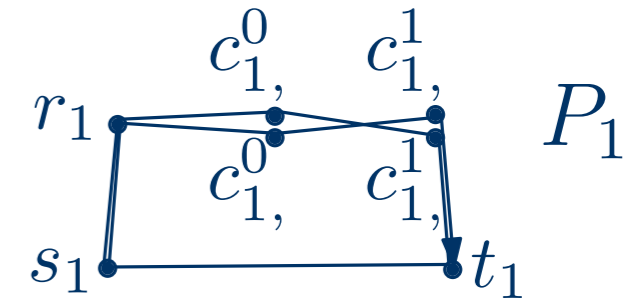
Reduction from k -CNF SAT

Given CNF-SAT formula F with N variables and M clauses.

Construct two curves P_1, P_2 such that

$$d_F(P_1, P_2) \leq 1 \iff F \text{ is satisfiable.}$$

Hence an $O(n^{2-\delta})$ algorithm for d_F would imply an $O(M^2 2^{(1-\delta/2)N})$ SAT algorithm.

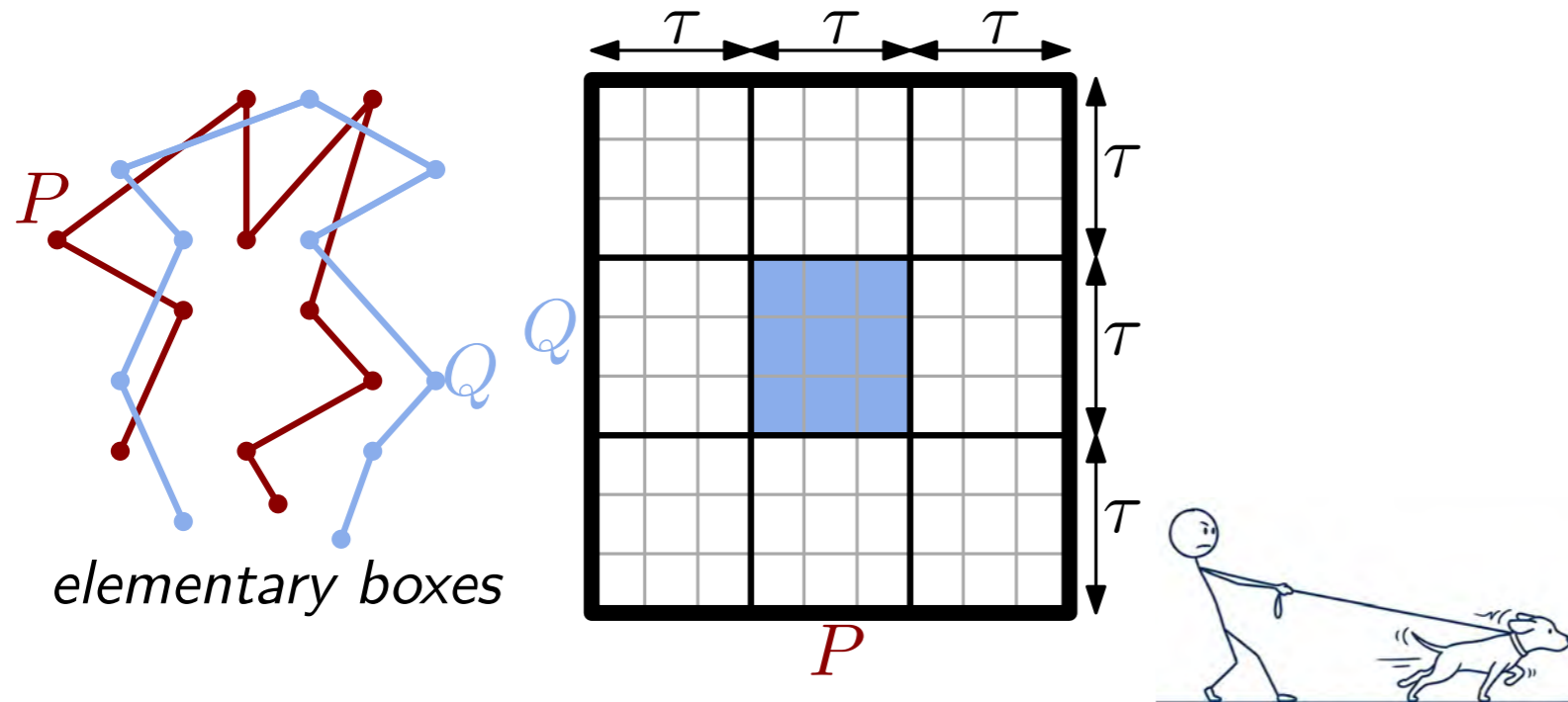


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Pre-processing to speed up the computation (4 Russians trick)

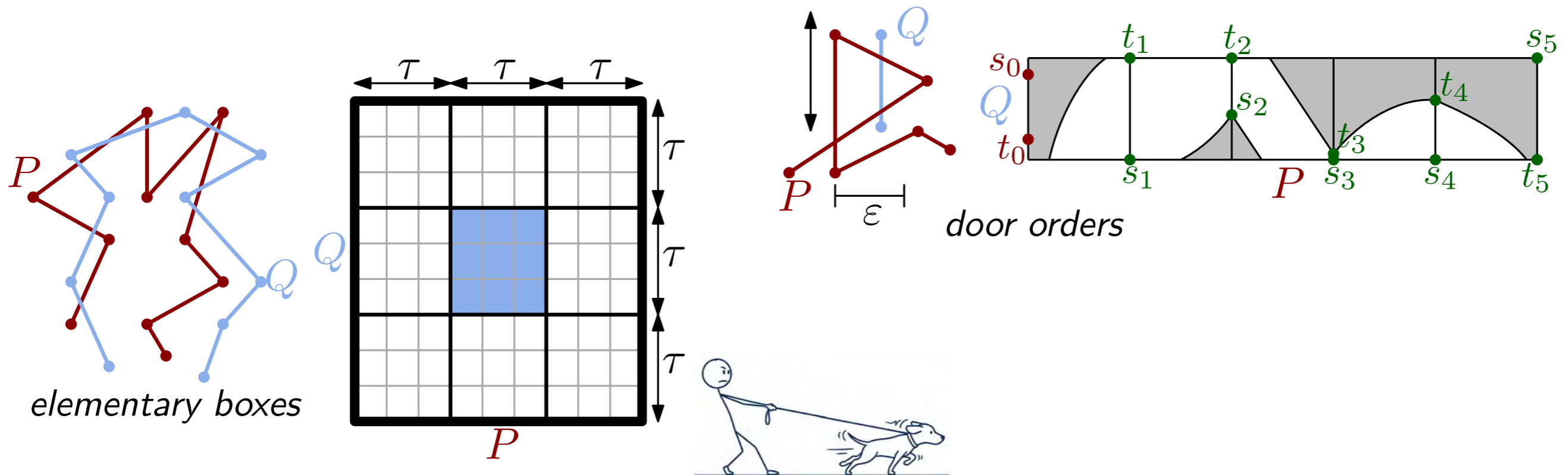


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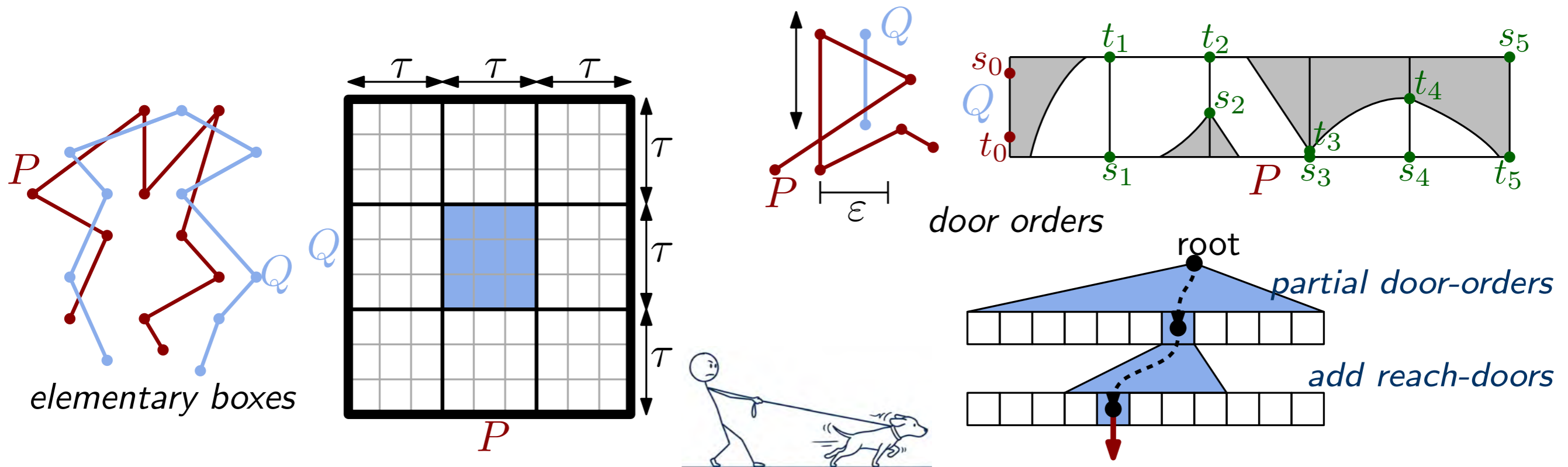


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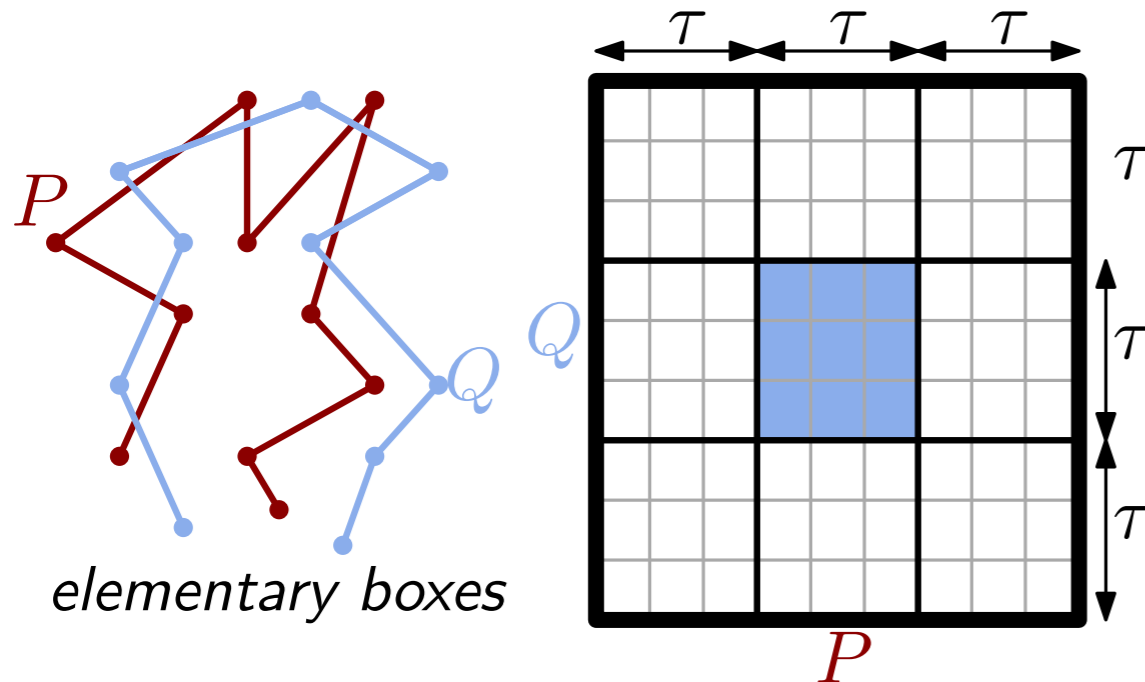


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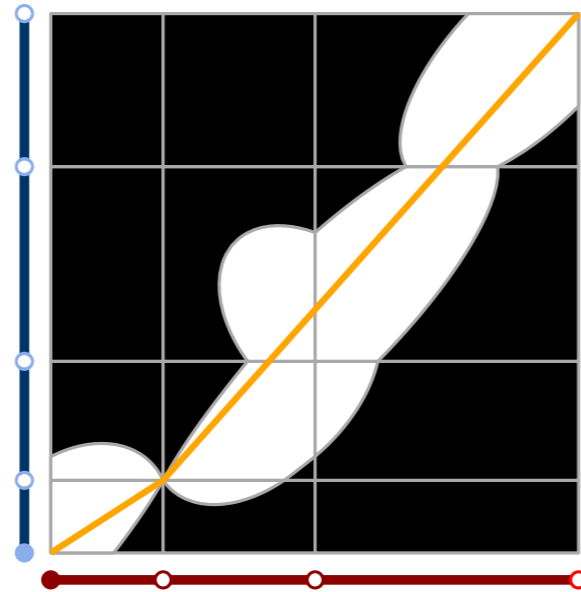
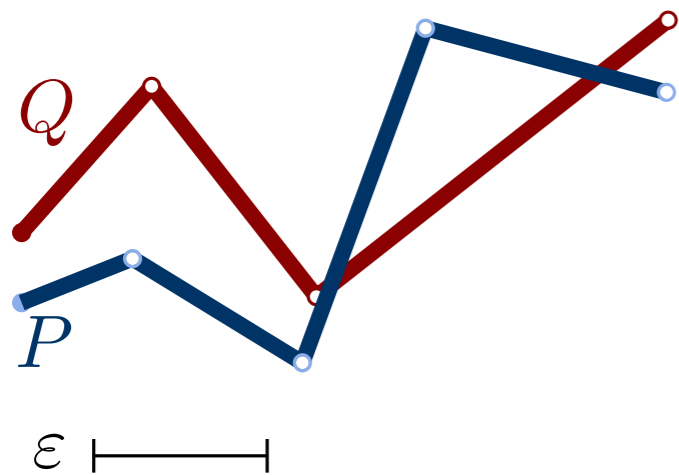
Pre-processing to speed up the computation (4 Russians trick)



use algebraic geometric tools to show only a subquadratic number of distinct encodings



Can the Fréchet distance be computed faster for certain inputs?

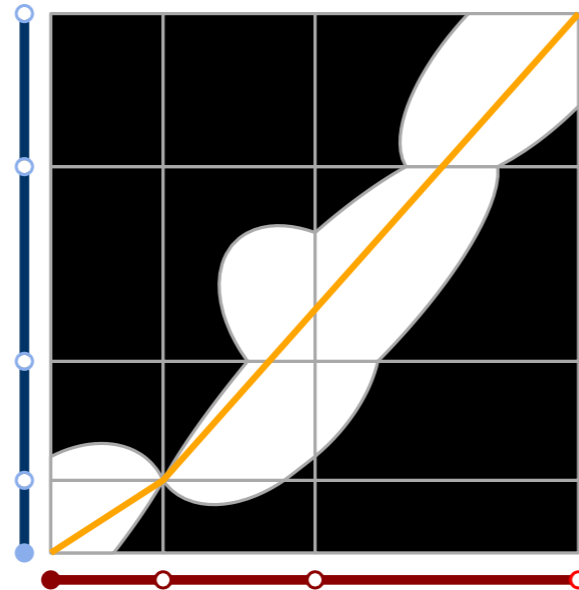
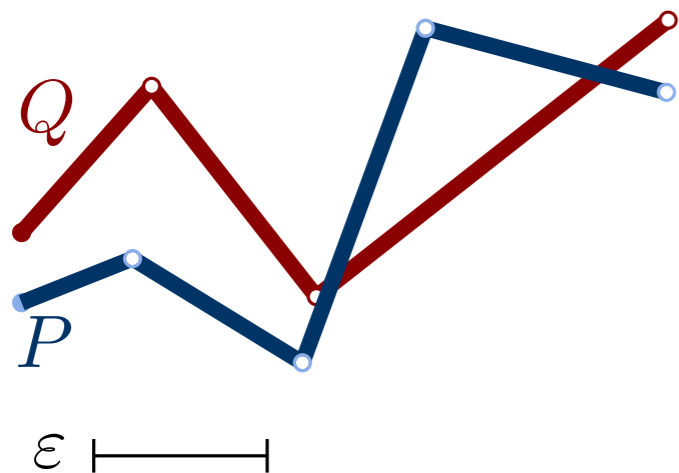


Bottleneck: size of the free space diagram

Realistic inputs

Can the Fréchet distance be computed faster for certain inputs?

- κ -straight, κ -bounded (Alt, Knauer, Wenk, 2004)
- backbone curves (Aronov et al. 2006)
- c -packed curves (Driemel, Har-Peled, Wenk, 2012),



Bottleneck: size of the free space diagram

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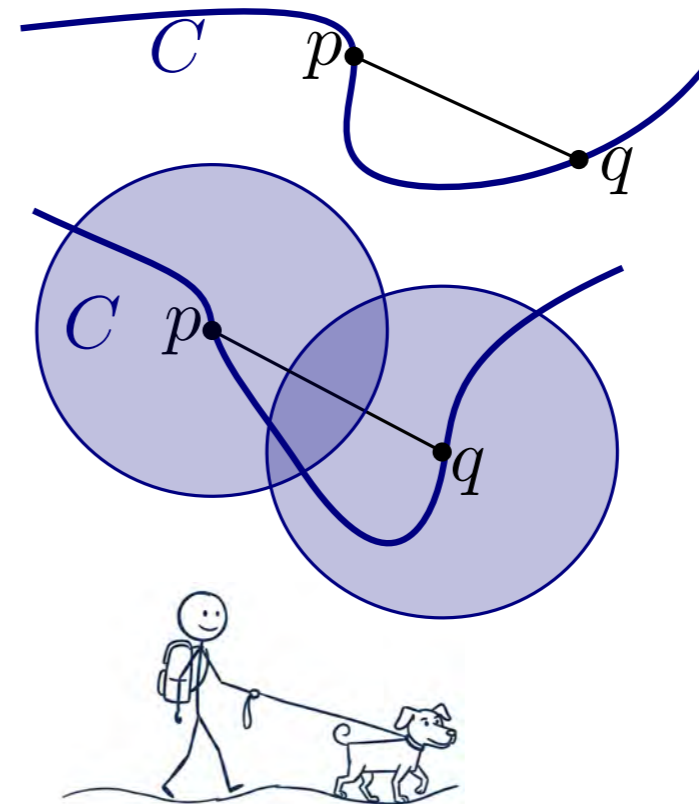
κ -straight

for p, q on curve C : $|C[p, q]| \leq \kappa \|p - q\|$

κ -bounded

for p, q on curve C : $C[p, q] \subseteq B(p, r) \cup B(q, r)$,
where $r = \kappa \|p - q\| / 2$

for these curves Fréchet distance can be bounded by $\kappa + 1$ times Hausdorff distance



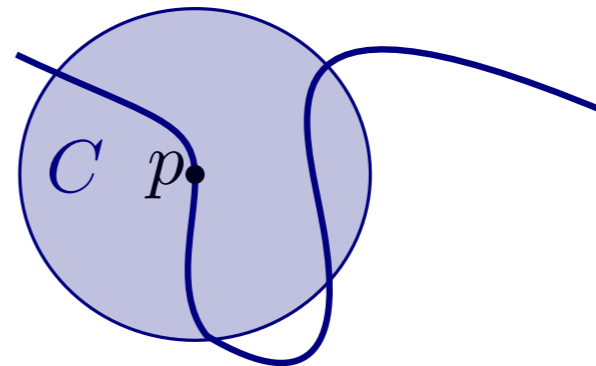
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c -packed

for any p : $|C \cap B(p, r)| \leq cr$

- simplify curves (c -packed remains $6c$ -packed, complexity of fsd reduced)
- approximately decide Fréchet distance between simplifications



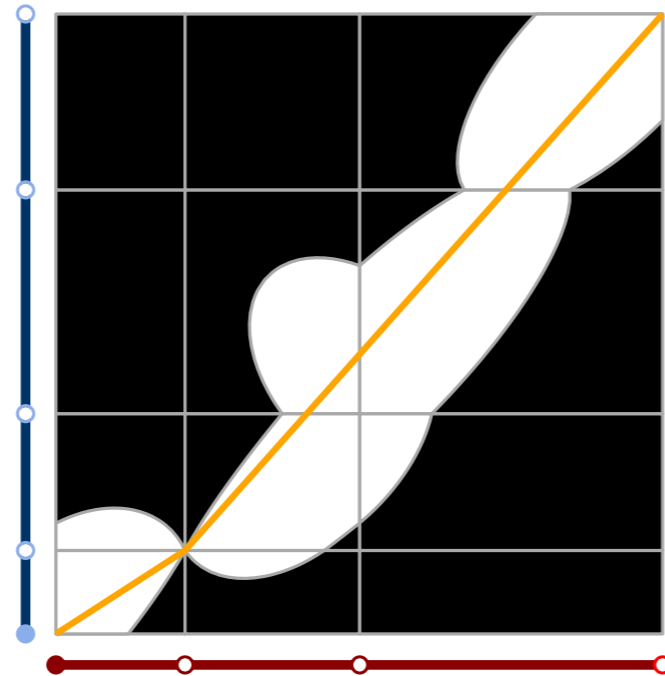
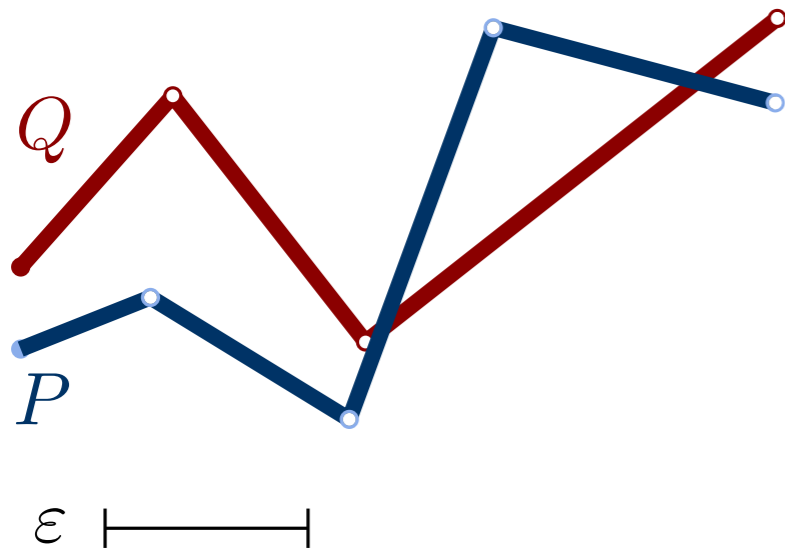
Can the Fréchet distance be computed faster for certain inputs?

- κ -straight, κ -bounded (Alt, Knauer, Wenk, 2004)
- backbone curves (Aronov et al. 2006)
- c -packed curves (Driemel, Har-Peled, Wenk, 2012),
- long edges (Gudmundsson, Mirzanezhad, Mohades, Wenk, 2019)
- one curve c -packed (Gudmundsson, Mai, Wong, 2024),
(Conradi, van der Hoog, van der Horst, Ophelders, 2026)
- imbalanced (Blank, 2026)



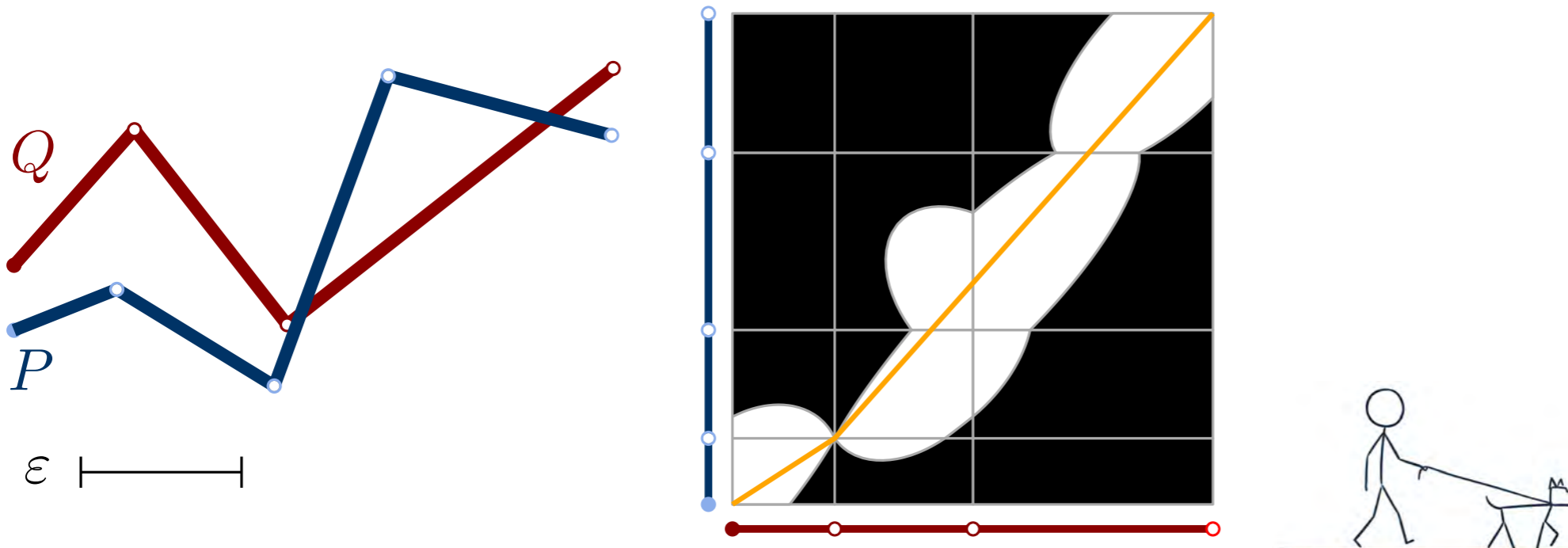
Approximation

Can the Fréchet distance be approximated faster?



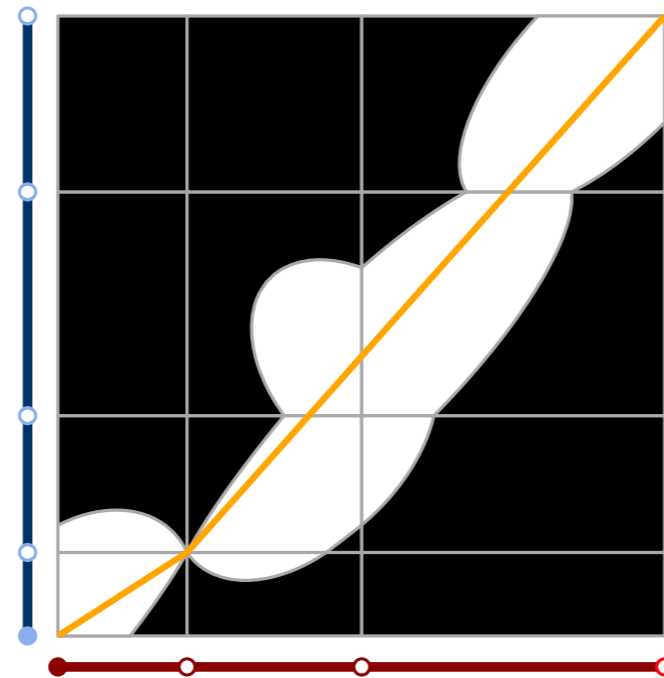
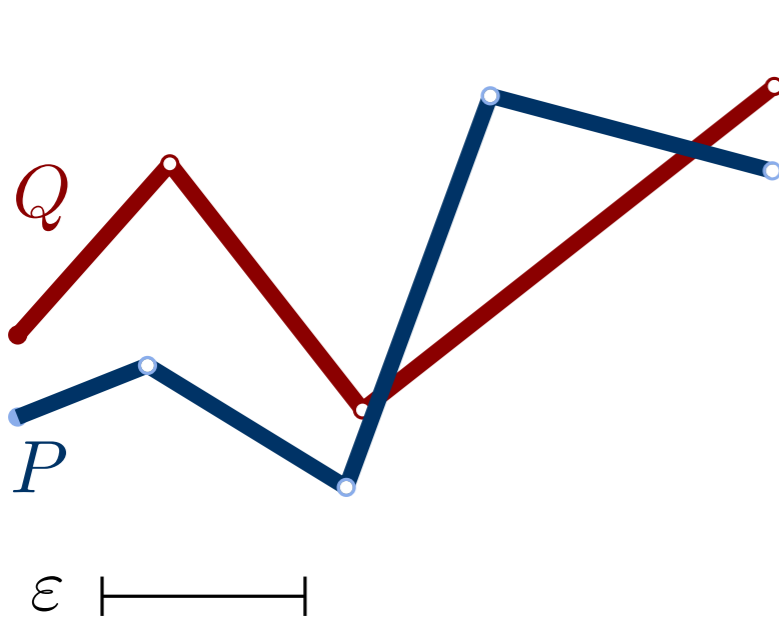
Can the Fréchet distance be approximated faster?

- no 3-approx in subquadratic time (Buchin, Ophelders, Speckmann, 2019)
- $O(\alpha)$ -approx for $\alpha \in [\sqrt{n}, n]$ in $O((n^3/\alpha^2) \log^3 n)$ (Colombe & Fox, 2021)
- $O(\alpha)$ -approx for $\alpha \in [1, n]$ in $O((n + mn/\alpha) \log^3 n)$ (van der Horst, van Kreveld, Ophelders & Speckmann, 2023)
- $O(\alpha)$ -approx for $\alpha \in [1, n]$ in $O((n + mn/\alpha) \log^2 n)$ (van der Horst, Ophelders, '24)
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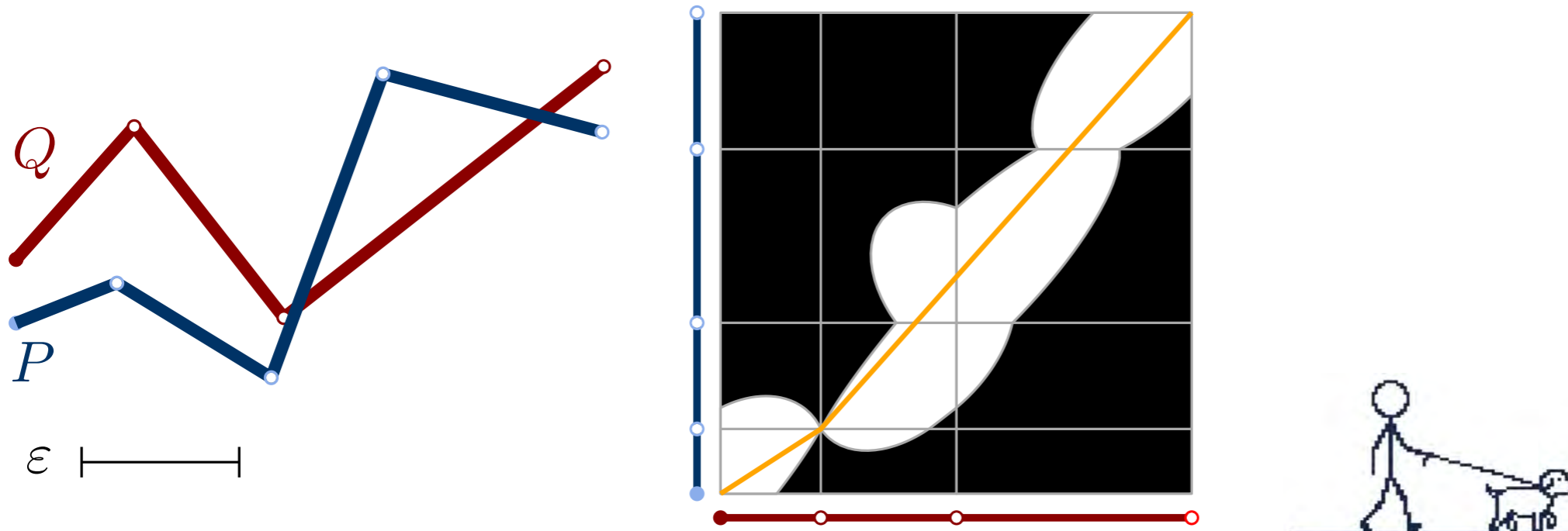
combines

- simplification
- grids
- greedy (for long edges)



Implementation

How to efficiently implement the algorithms?



How to efficiently implement the algorithms?



Trajectory similarity is a very basic problem when dealing with this type of data and many different ways of comparing trajectories have been proposed. In this challenge, we concentrate on a very intuitive measure called **Fréchet distance** with hard computational complexity, but superior quality in theory and practice.

Paper of the Winning Teams

All winning teams were invited to publish a paper describing their approach in the proceedings of SIGSPATIAL GIS.

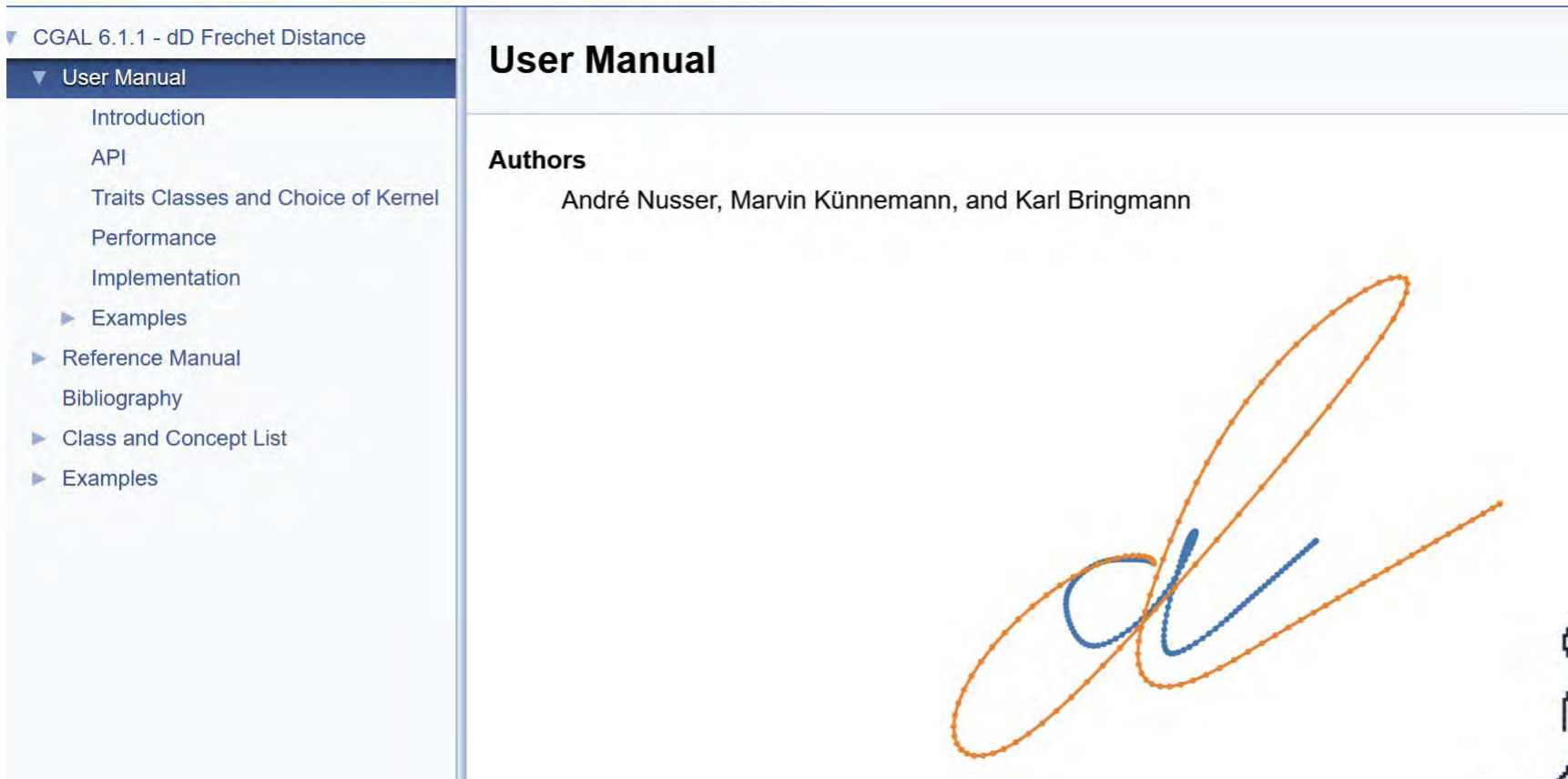
- **A fast implementation of near neighbors queries for fréchet distance.**
J. Baldus and K. Bringmann.
- **Efficient trajectory queries under the fréchet distance.**
K. Buchin, Y. Diez, T. van Diggelen, and W. Meulemans.
- **A filter-and-refinement-algorithm for range queries based on the fréchet distance.**
F. Dütsch and J. Vahrenhold.



How to efficiently implement the algorithms?

- Walking the Dog Fast in Practice: Algorithm Engineering of the Fréchet Distance (Bringmann, Künneemann, Nusser, 2019)

CGAL 6.1.1 - dD Frechet Distance



The image shows a screenshot of the CGAL 6.1.1 User Manual for the dD Frechet Distance. The left sidebar contains a navigation menu with the following items: Introduction, API, Traits Classes and Choice of Kernel, Performance, Implementation, Examples, Reference Manual, Bibliography, Class and Concept List, and Examples. The main content area is titled "User Manual" and includes the authors: André Nusser, Marvin Künneemann, and Karl Bringmann. Below the text, there is a diagram illustrating the Fréchet distance between two curves. One curve is a blue line, and the other is an orange line. A stick figure is shown walking along the orange curve, holding a leash that is attached to a dog walking along the blue curve. The leash represents the distance between the two curves at any given point.

Input: Polygonal curves P, Q

Output: $d_F(P, Q)$

determine all critical values

sort the critical values

do a **binary search** over the values,
solving the decision problem in each step

Runtime: $O(n^3 \log n)$

faster: with *parametric search* in $O(n^2 \log n)$ time

Theorem (Alt, Godau '95)

For two polygonal curves P, Q of size n , one can

- decide $d_F(P, Q) < \varepsilon$ in $O(n^2)$ time
- compute $d_F(P, Q)$ in $O(n^2 \log n)$ time

Questions:

- faster!
- implementation!
- generalizations?
- applications?

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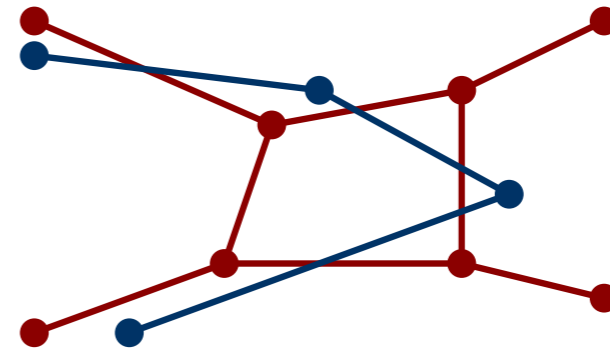
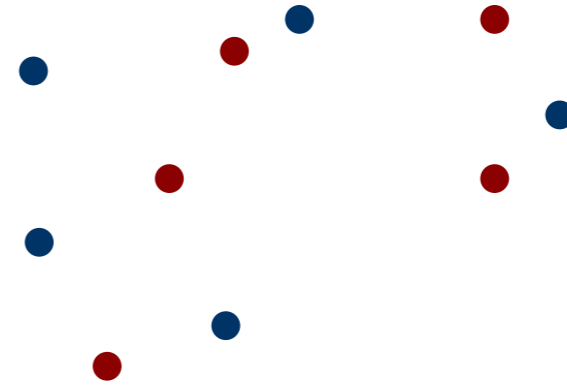
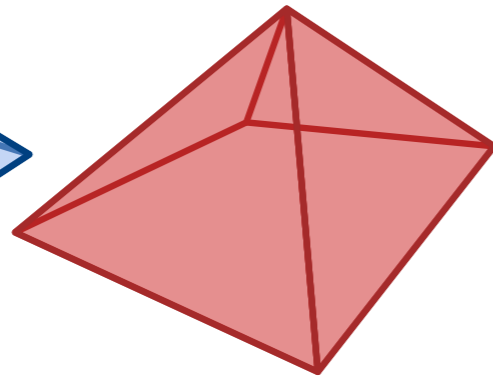
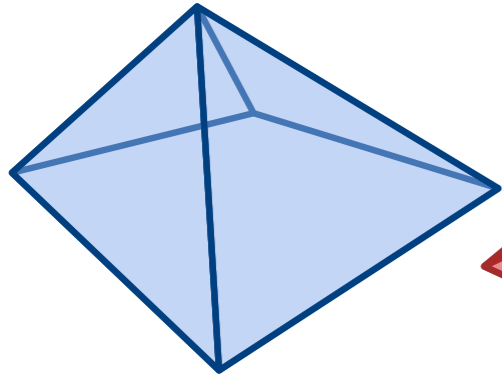
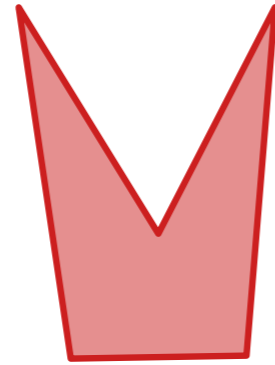
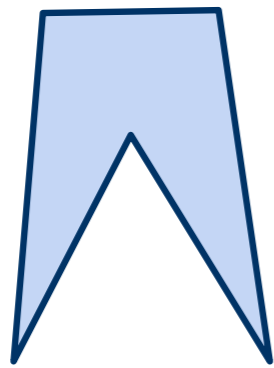
Questions:

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- **generalizations?**
- **applications?**



Generalizations

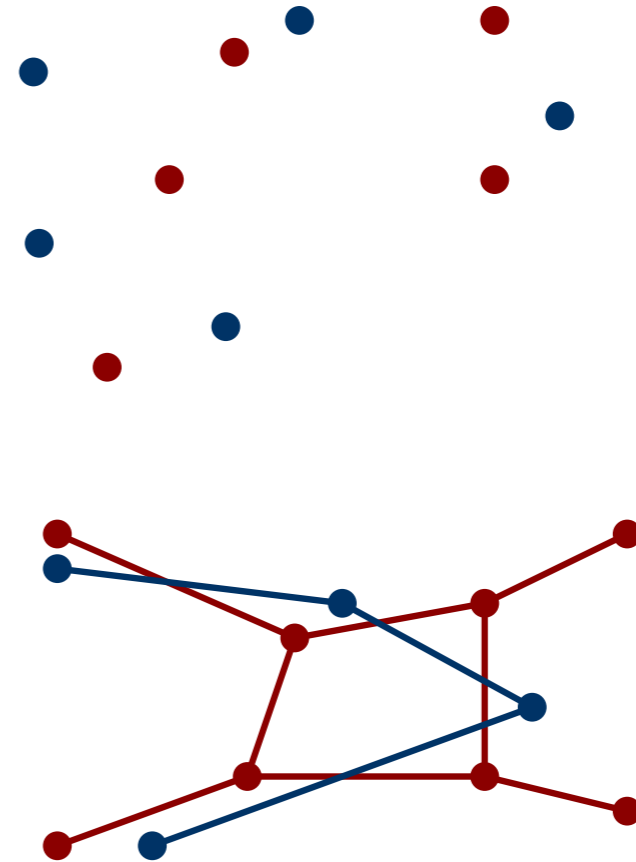
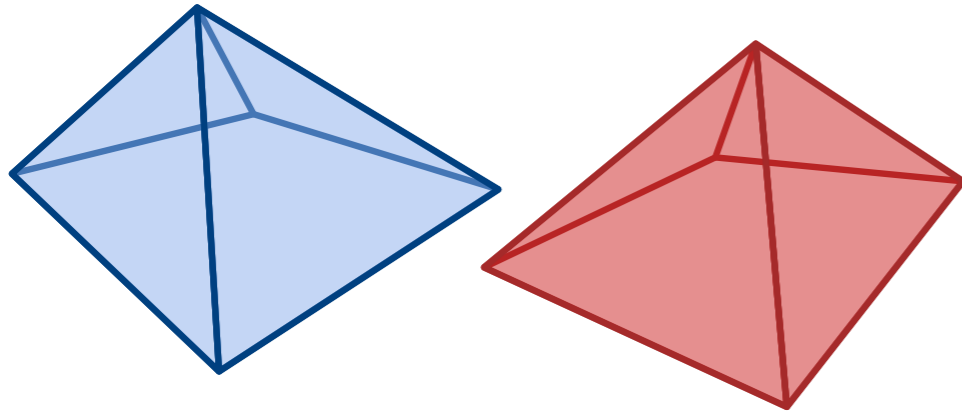
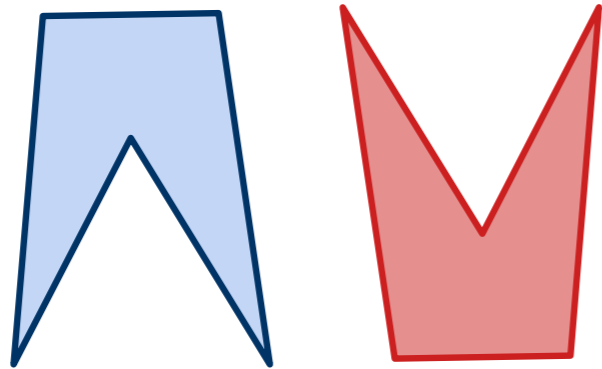
What about other shapes, e.g., surfaces, polygons, graphs, or point sets?



Generalizations

What about other shapes, e.g., surfaces, polygons, graphs, or point sets?

→ Fréchet distance has been generalized to all of these



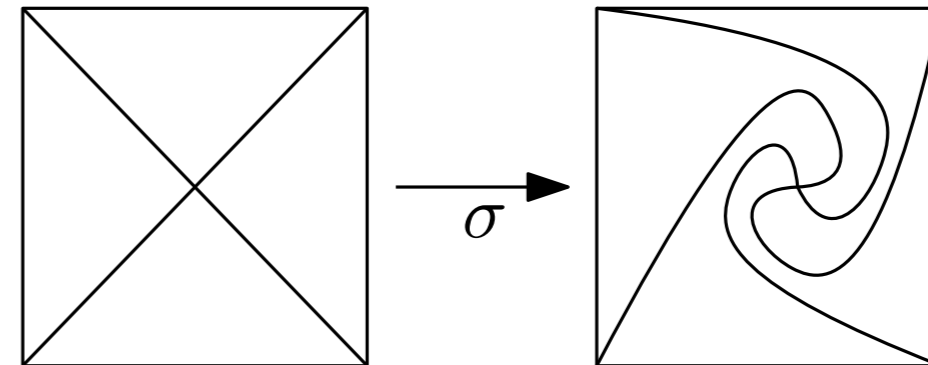
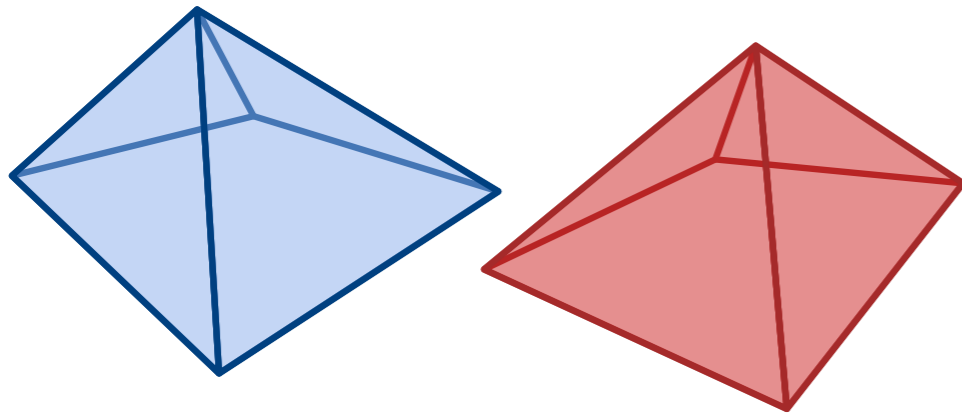
several of these turn out to be NP-hard



What about other shapes, e.g., **surfaces**, polygons, graphs, or point sets?

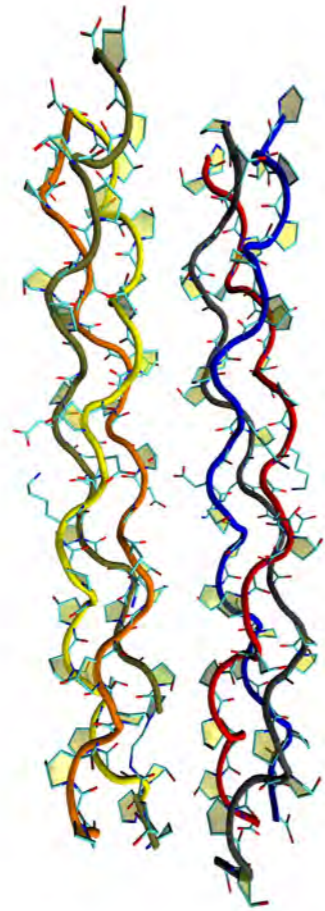
→ Fréchet distance has been generalized to all of these

- generalized to surfaces (Fréchet, 1924)
- NP-hard (Godau, 1998), (Buchin, Buchin, Schulz 2010)
- Semi-computable (Alt, Buchin, 2004)
- Computable (Nayyeri, Xu, 2016)

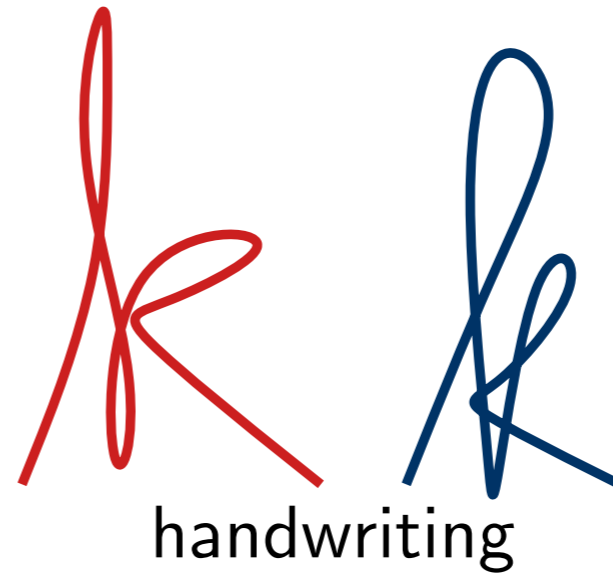


Homeomorphisms on $[0, 1]^2$ are not as nice as on $[0, 1]$

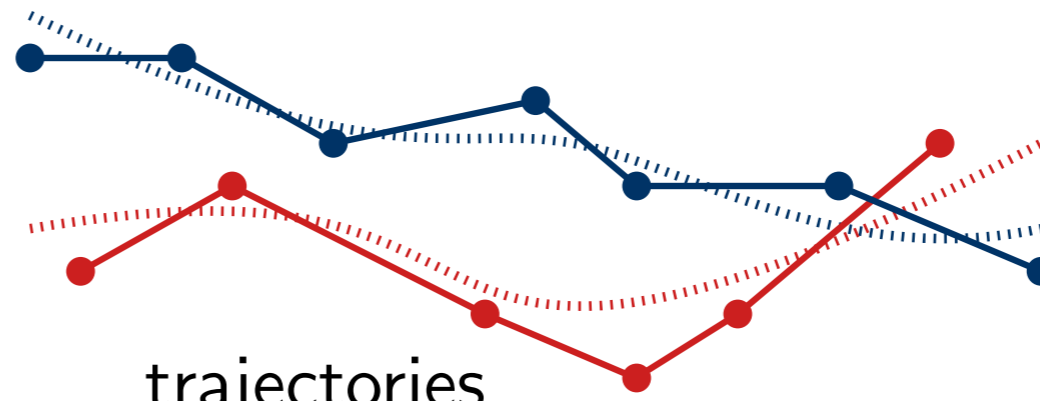
Shape matching: given two curves, find a transformation minimizing their distance.



protein backbones



handwriting



trajectories

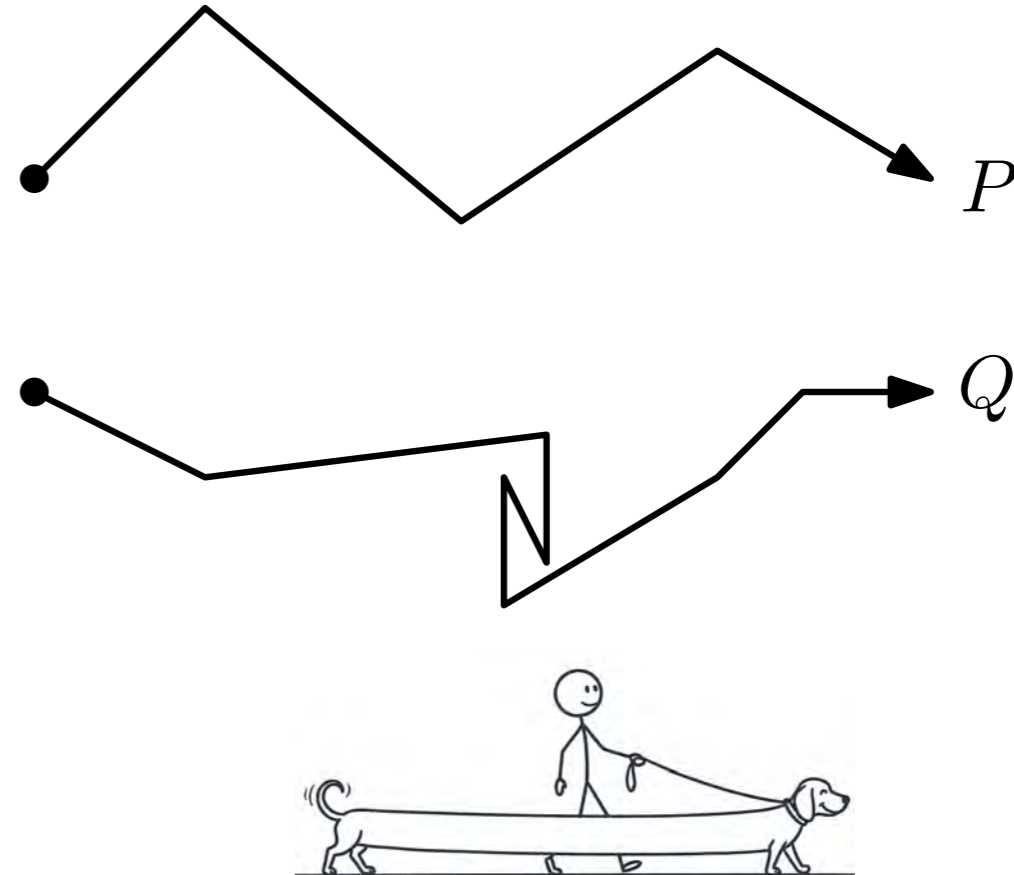
Fréchet distance of Curves under Transformations

- in 2D in $\tilde{O}(n^8)$ time (Alt, Knauer, Wenk, 2001)
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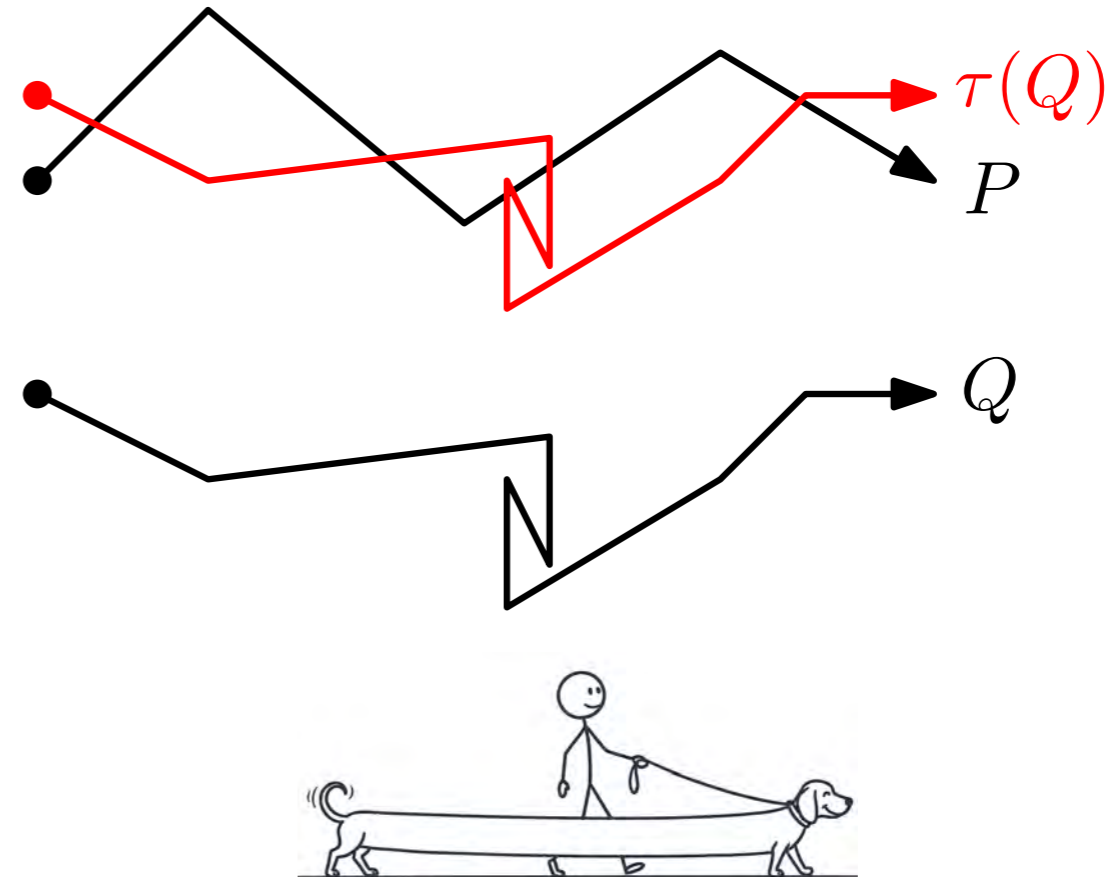
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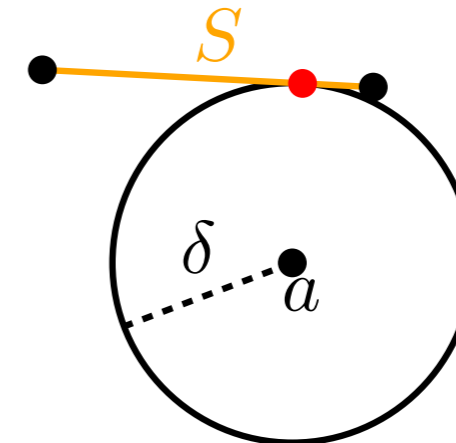
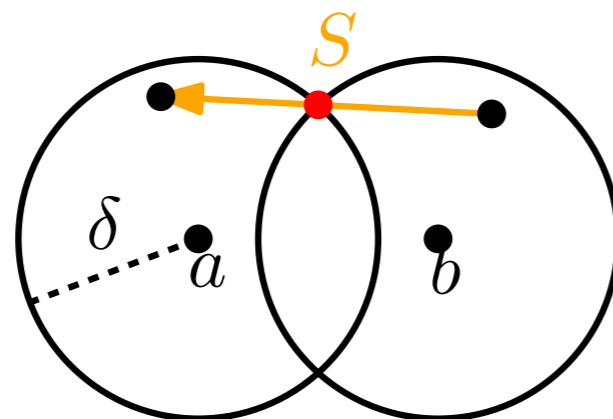
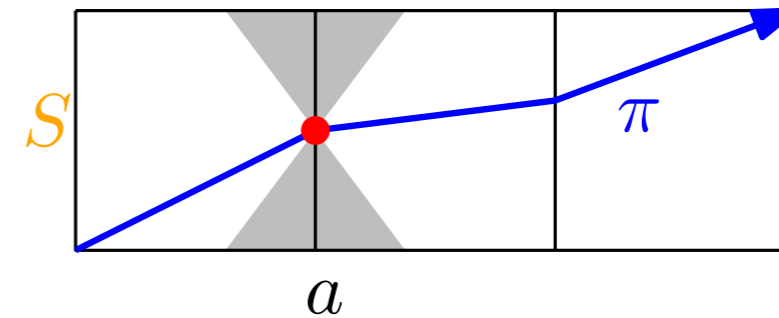
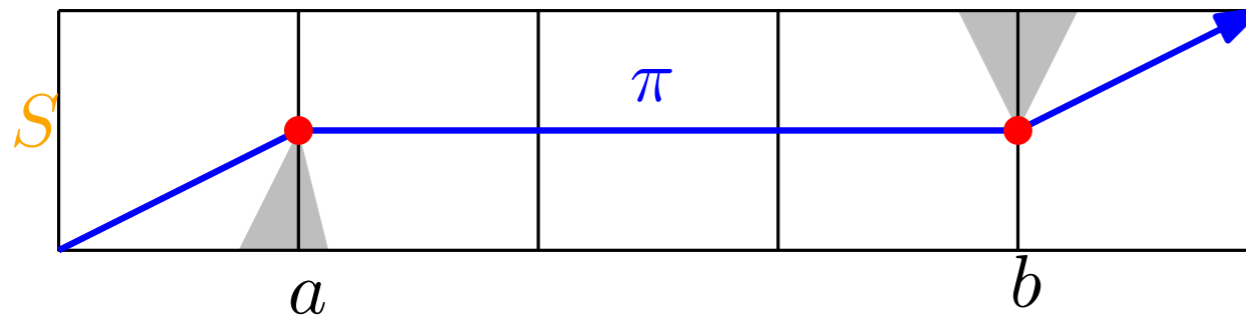
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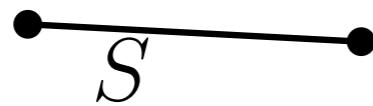
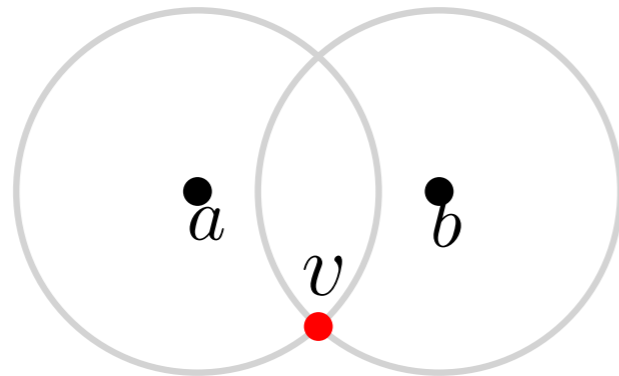
which translations change reachability in fsd?



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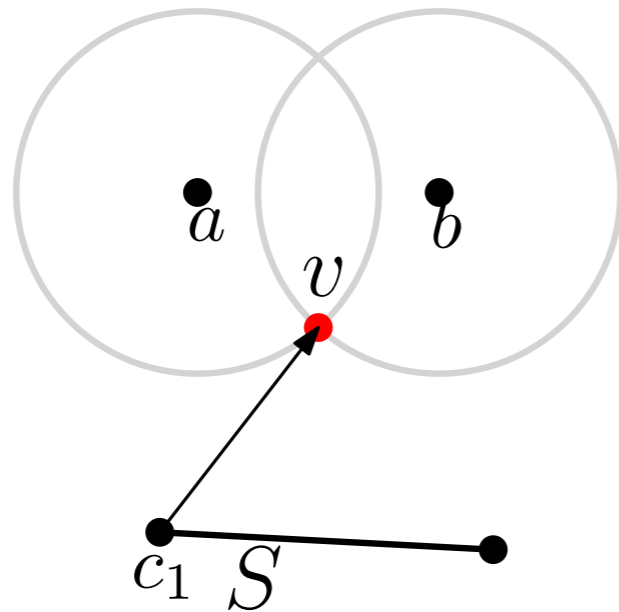
Geometric space

Translation space

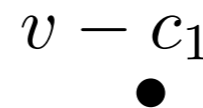
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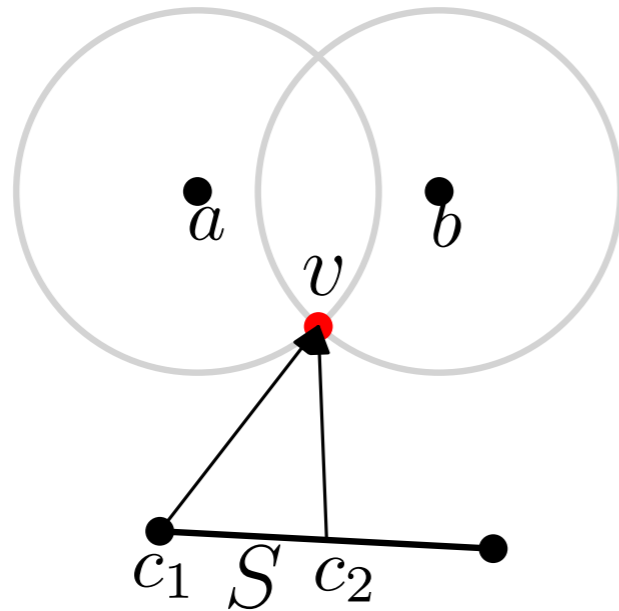


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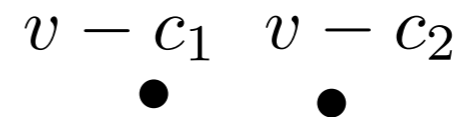
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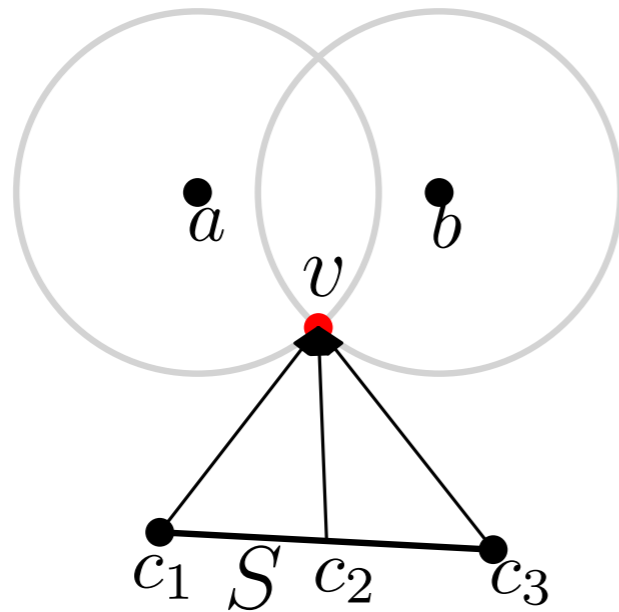


Translation space

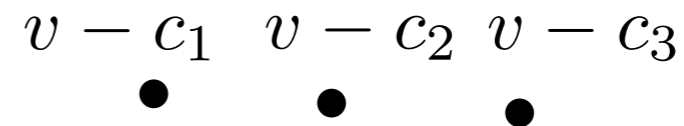
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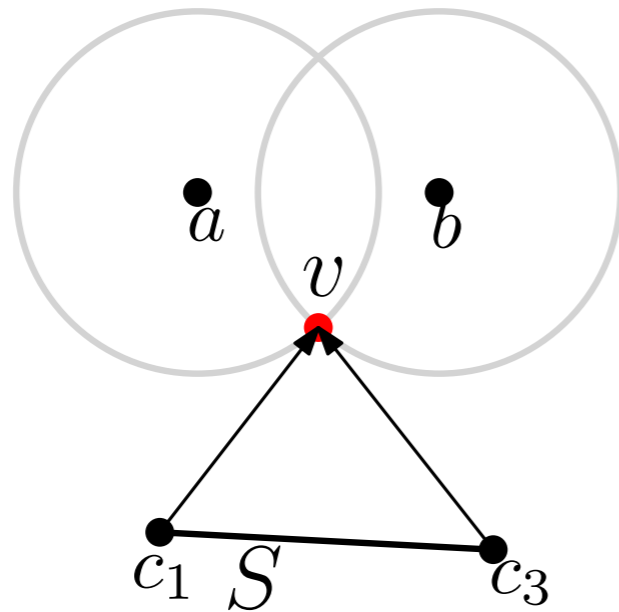


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which translations change reachability in fsd?



Geometric space

- $O(n^3)$ translations in total
- $O(n^2)$ pairs of vertices on P
- $O(n)$ segments on Q

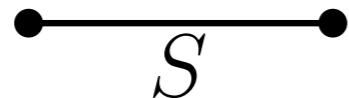
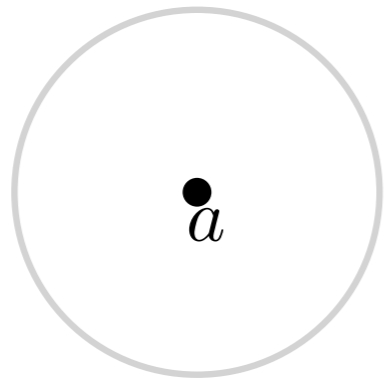


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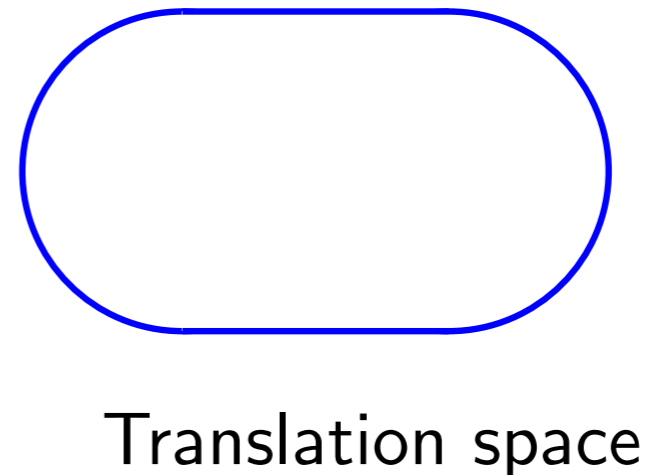
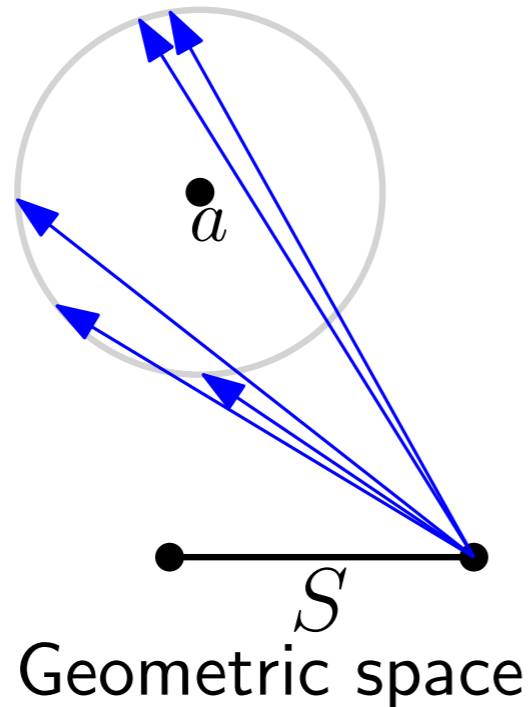
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Translation space

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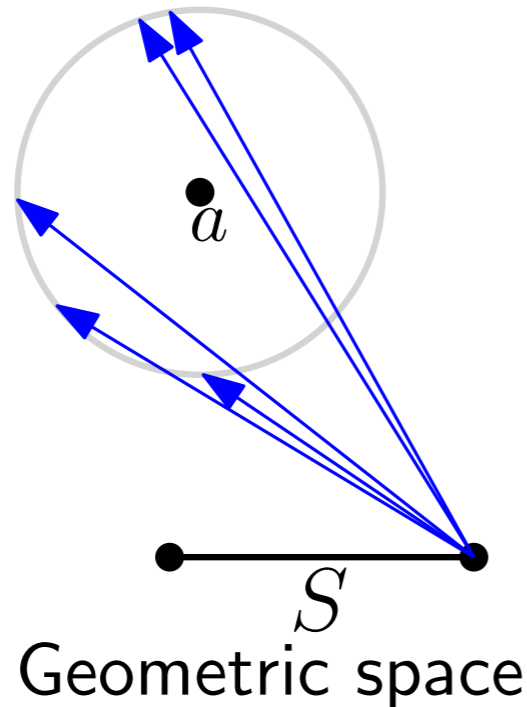
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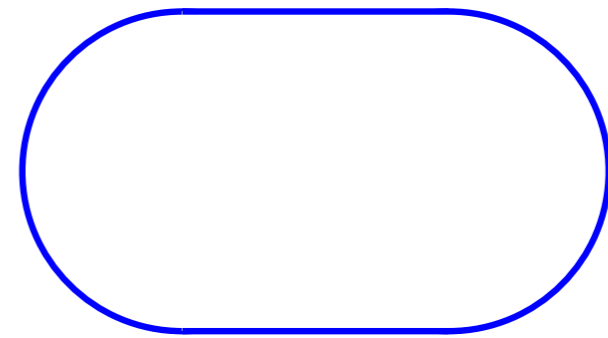
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which translations change reachability in fsd?



- $O(n^2)$ racetrack translations in total
- $O(n)$ vertices on Q
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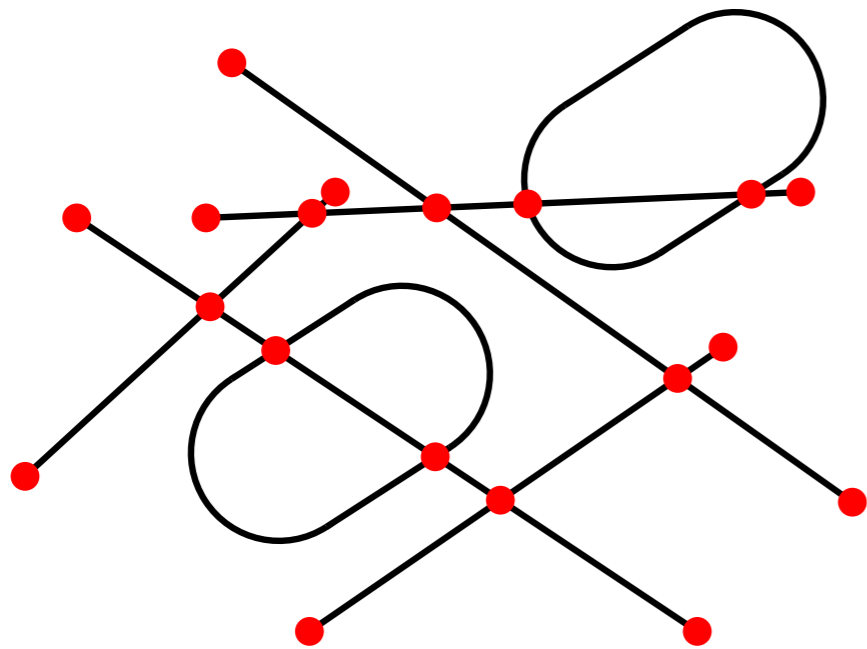


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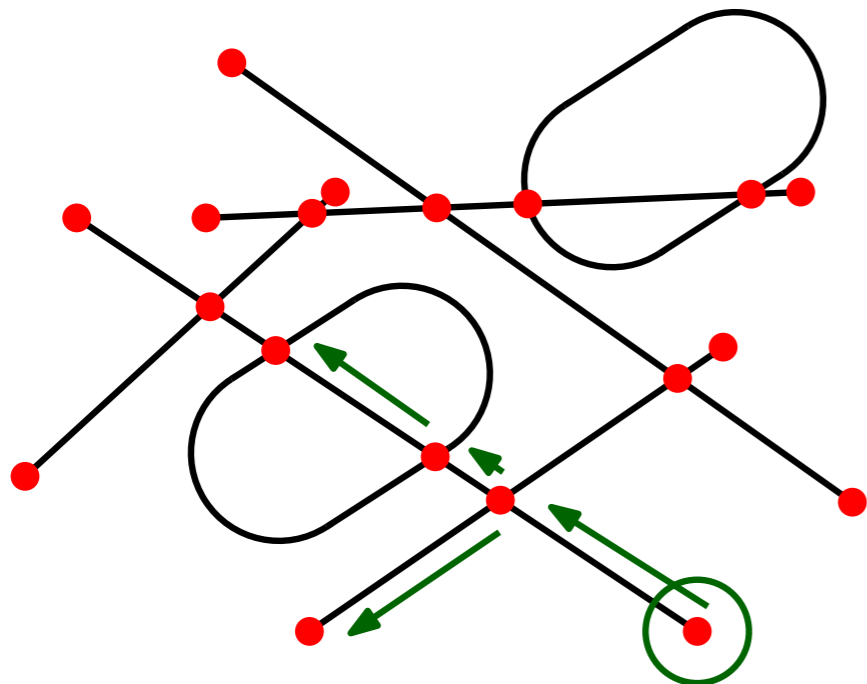


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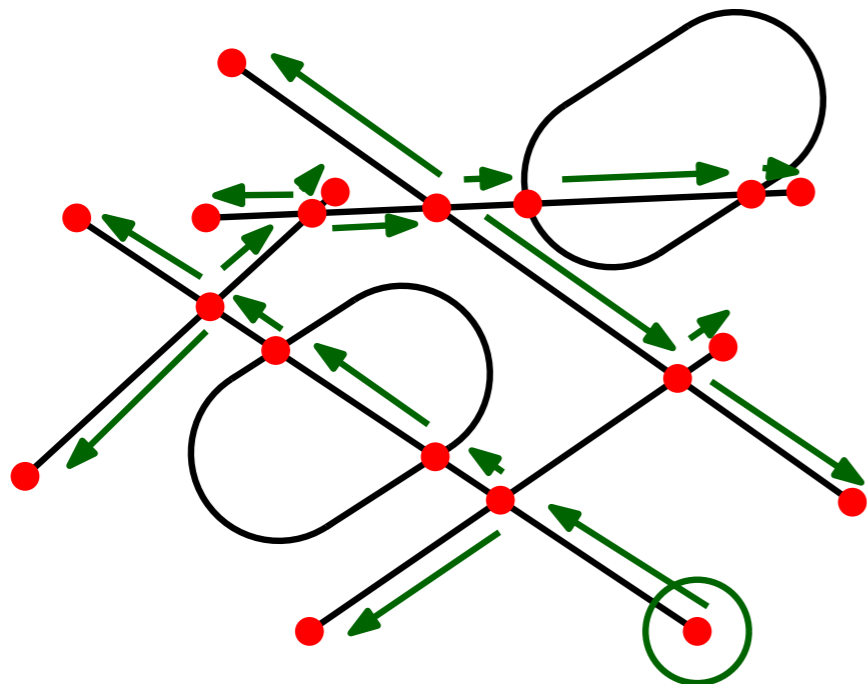
Translation space

1. Pick a starting intersection, and explore the arrangement of size $O(n^6)$ using DFS
2. For each intersection, update the freespace diagram and query reachability in $O(n^{\frac{4}{3}})$ time.

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Translation space

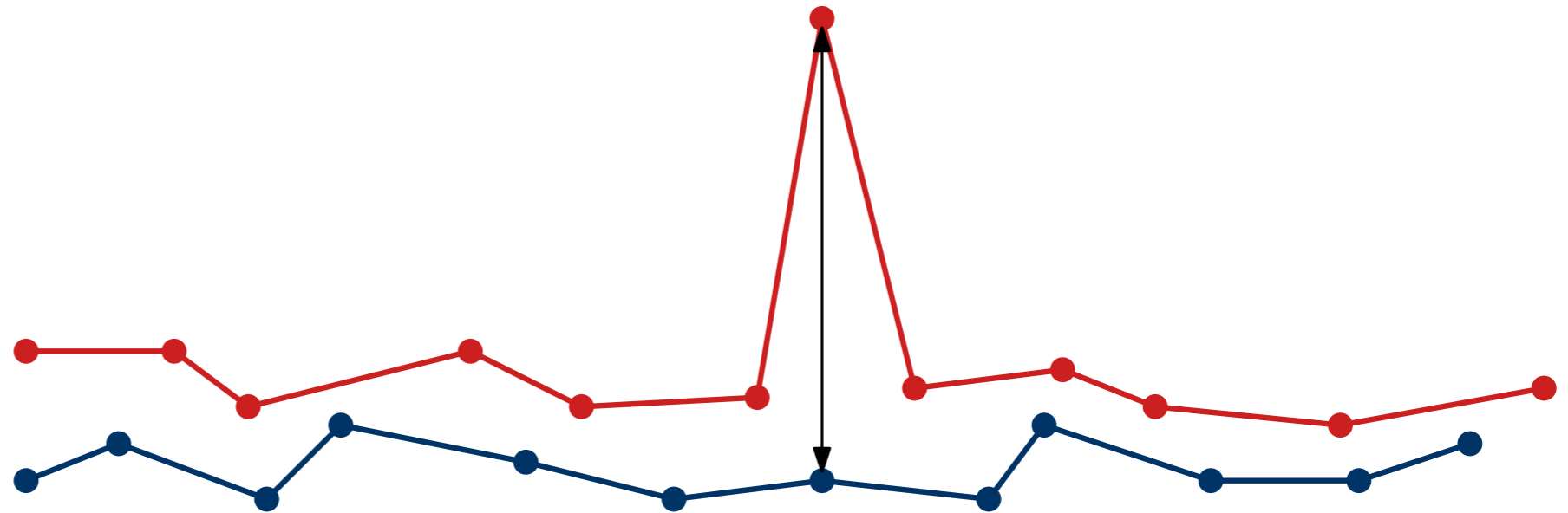
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Less Sensitive to Outliers

Fréchet distance is a bottleneck measure

How to handle this?

- ignore parts of the curve
- take an average

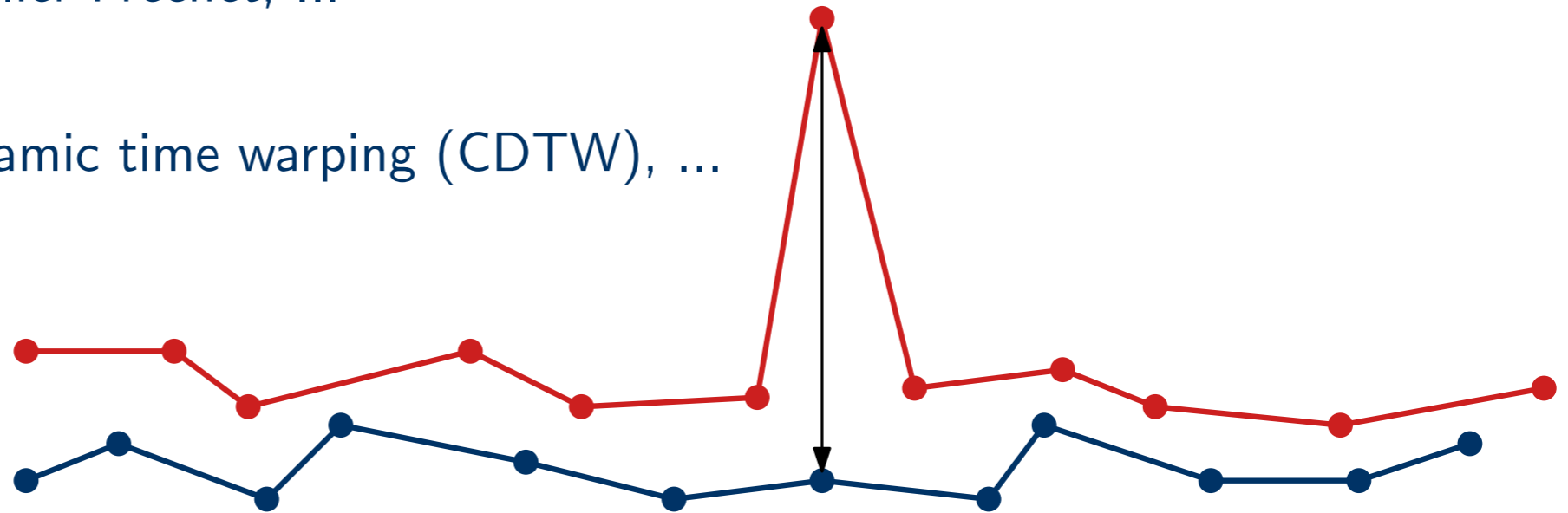


Less Sensitive to Outliers

Fréchet distance is a bottleneck measure

How to handle this?

- ignore parts of the curve
partial, k-shortcut, k-outlier Fréchet, ...
- take an average
average, continuous dynamic time warping (CDTW), ...



Variants

How can the Fréchet distance be adapted?



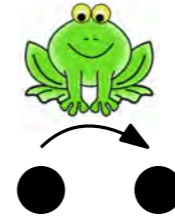
Variants

- weak [1992, Alt & Godau]



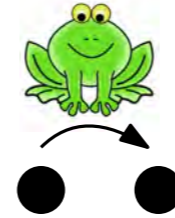
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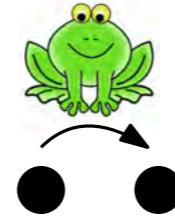
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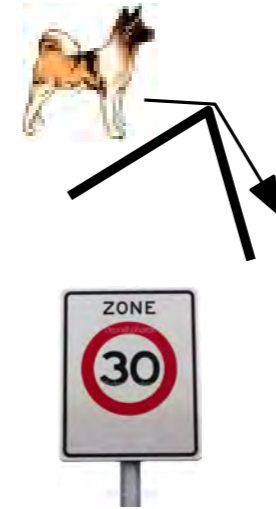
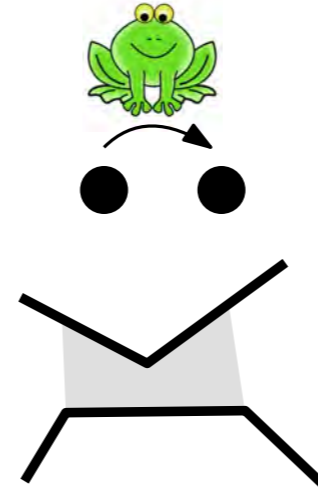
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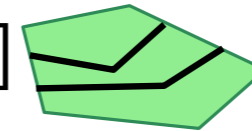
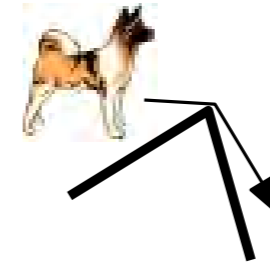
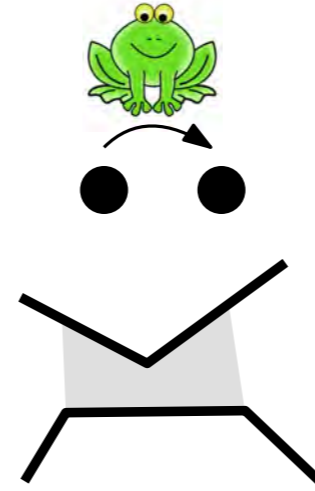
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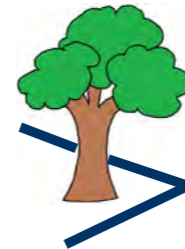
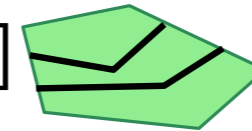
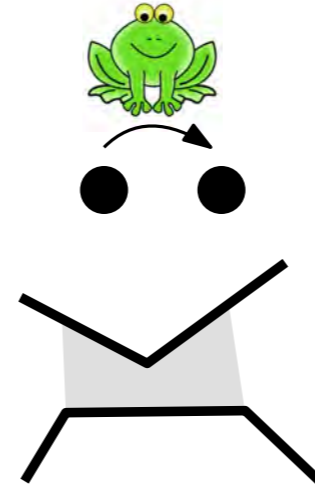
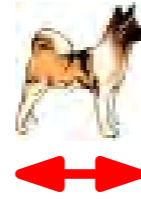
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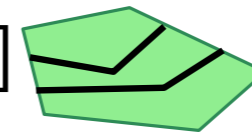
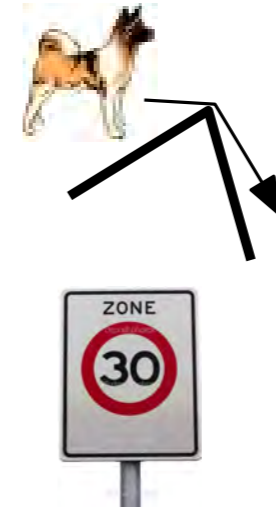
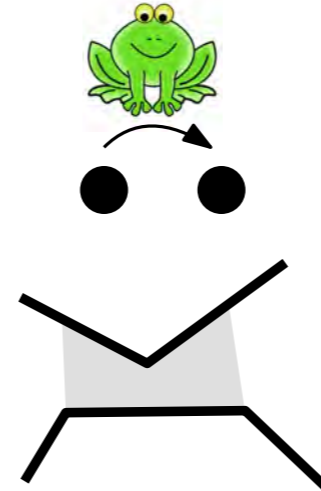
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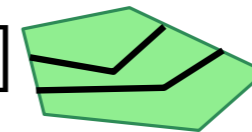
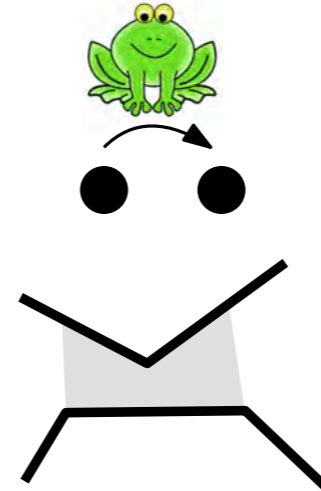
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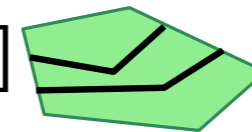
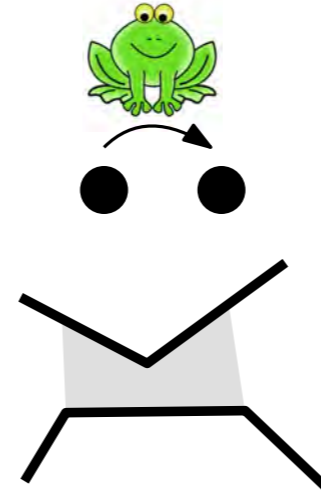
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- ... 14 more years of variants ..



Tasks using similarity

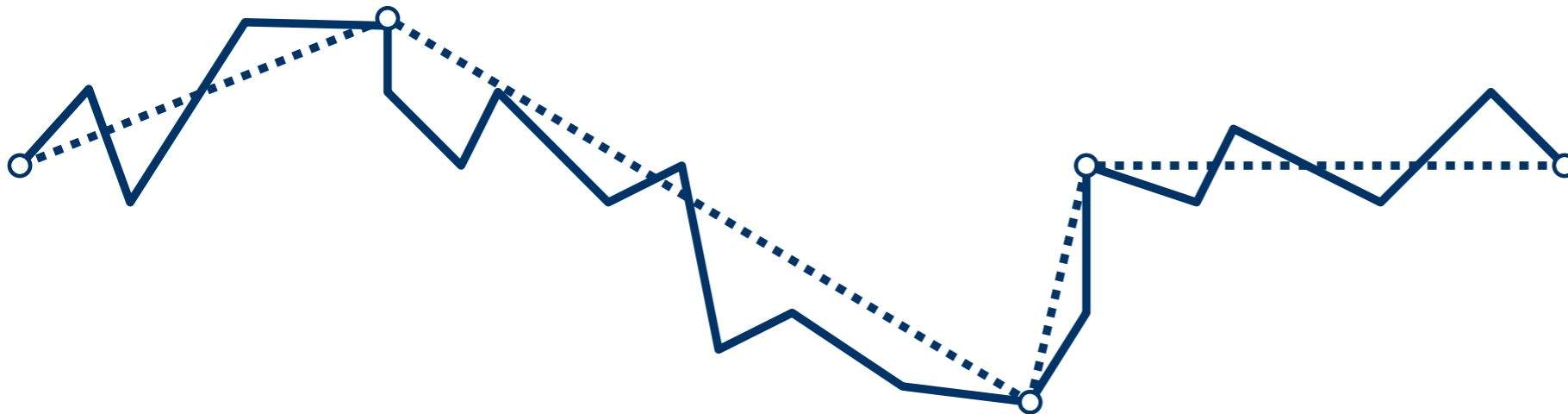
Similarity used in many other tasks for example



Tasks using similarity

Similarity used in many other tasks for example

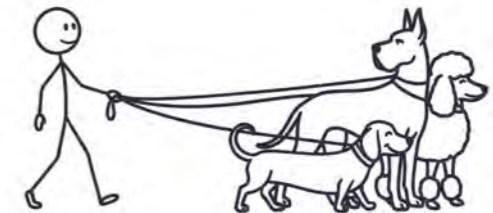
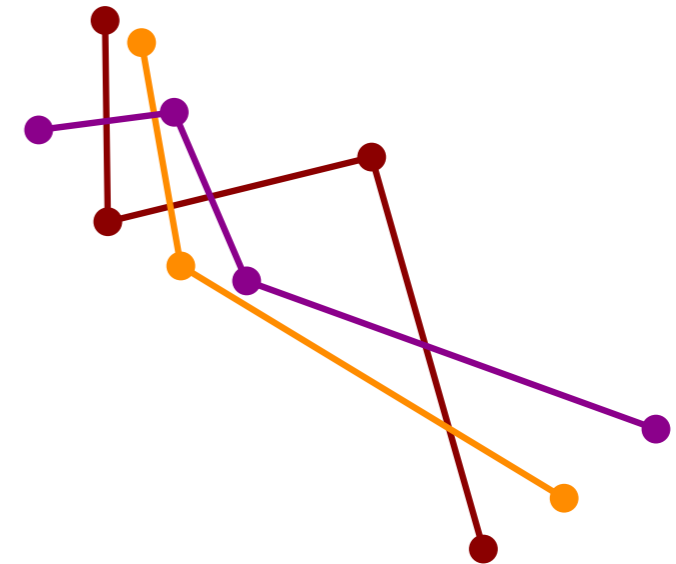
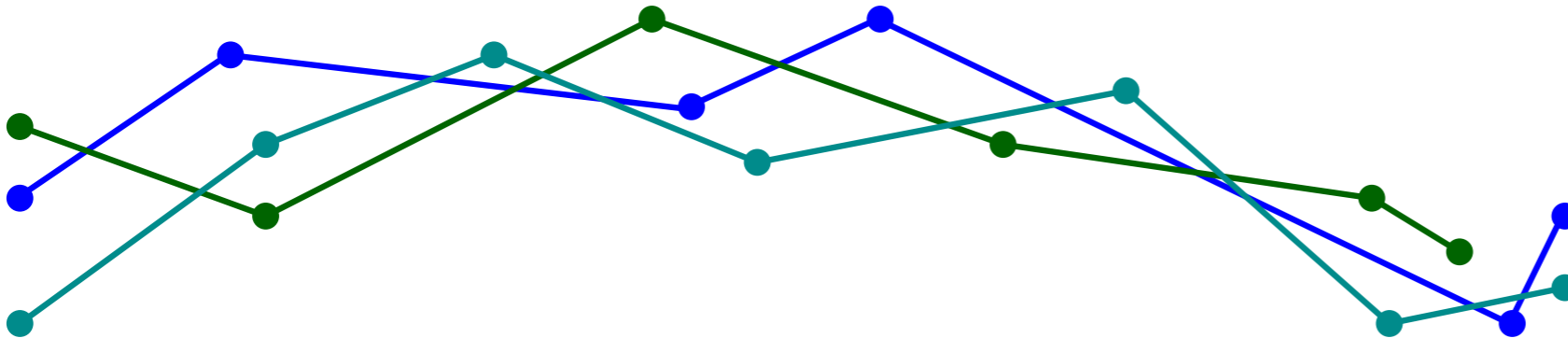
- **simplification:** reduce complexity with high similarity



Tasks using similarity

Similarity used in many other tasks for example

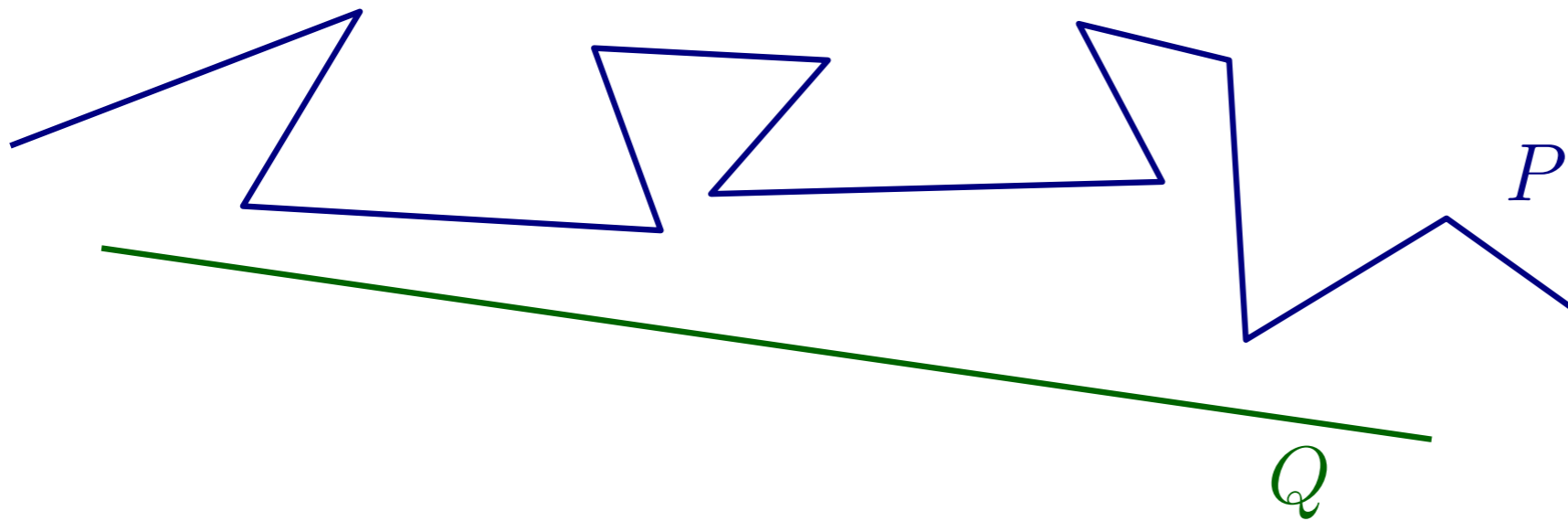
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Tasks using similarity

Similarity used in many other tasks for example

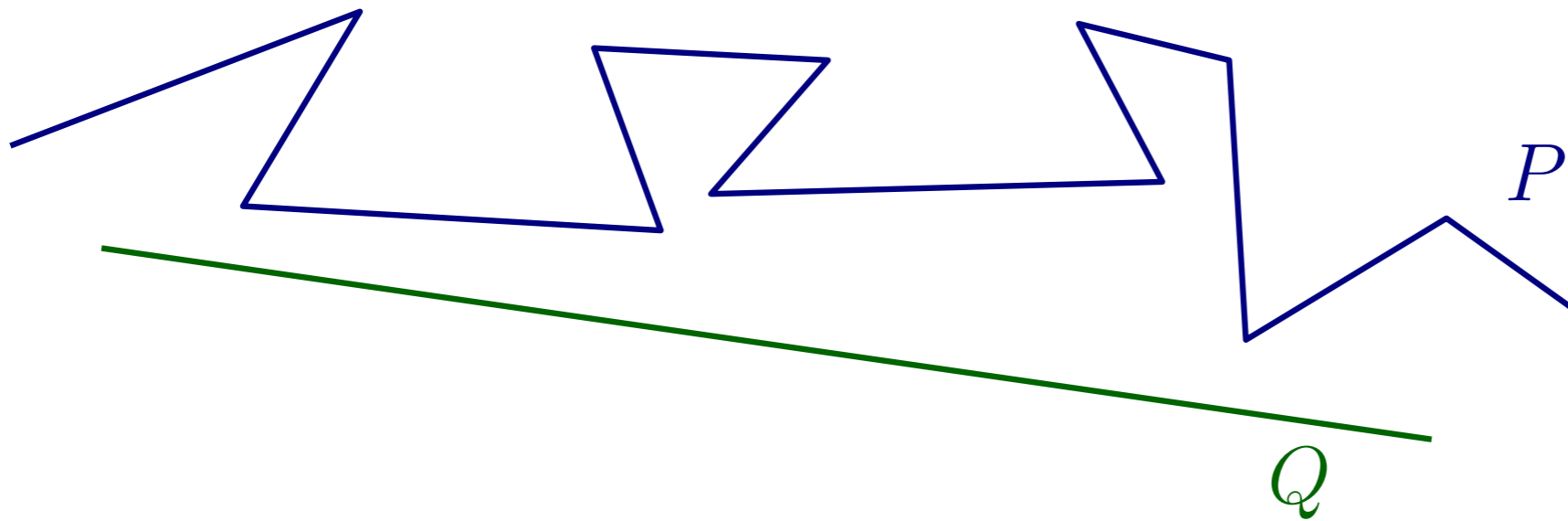
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Tasks using similarity

Similarity used in many other tasks for example

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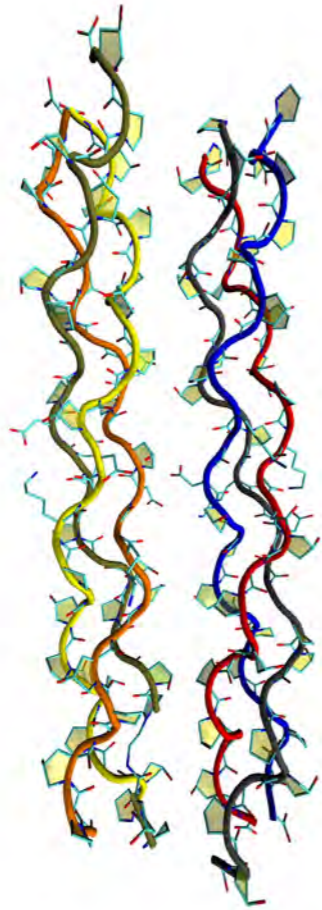


many interesting settings and results

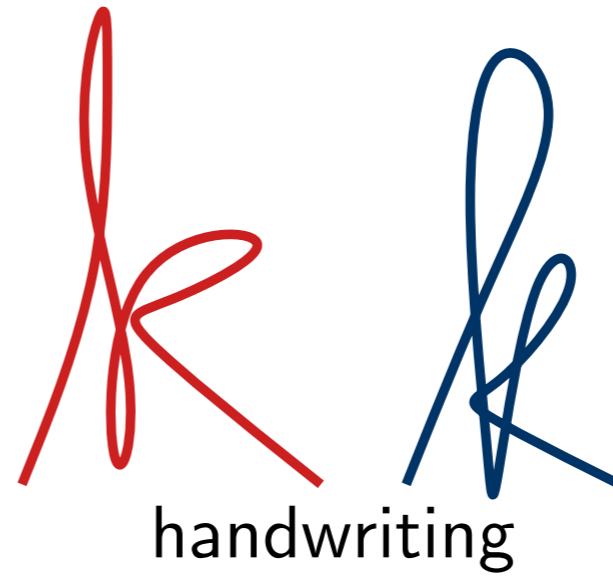


Applications

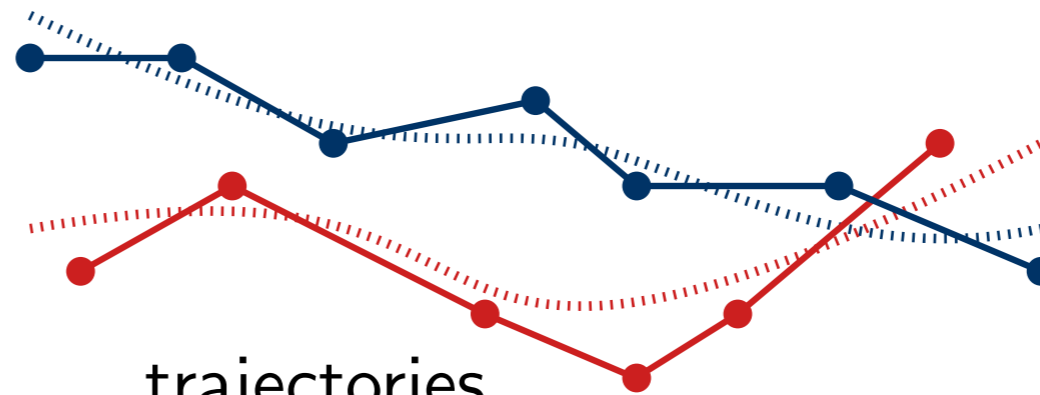
many applications, for instance..



protein backbones



handwriting



trajectories

Applications: Road Network Analysis

many entities moving
on a network



Applications: Road Network Analysis

many entities moving
on a network

Questions

- which path was taken?

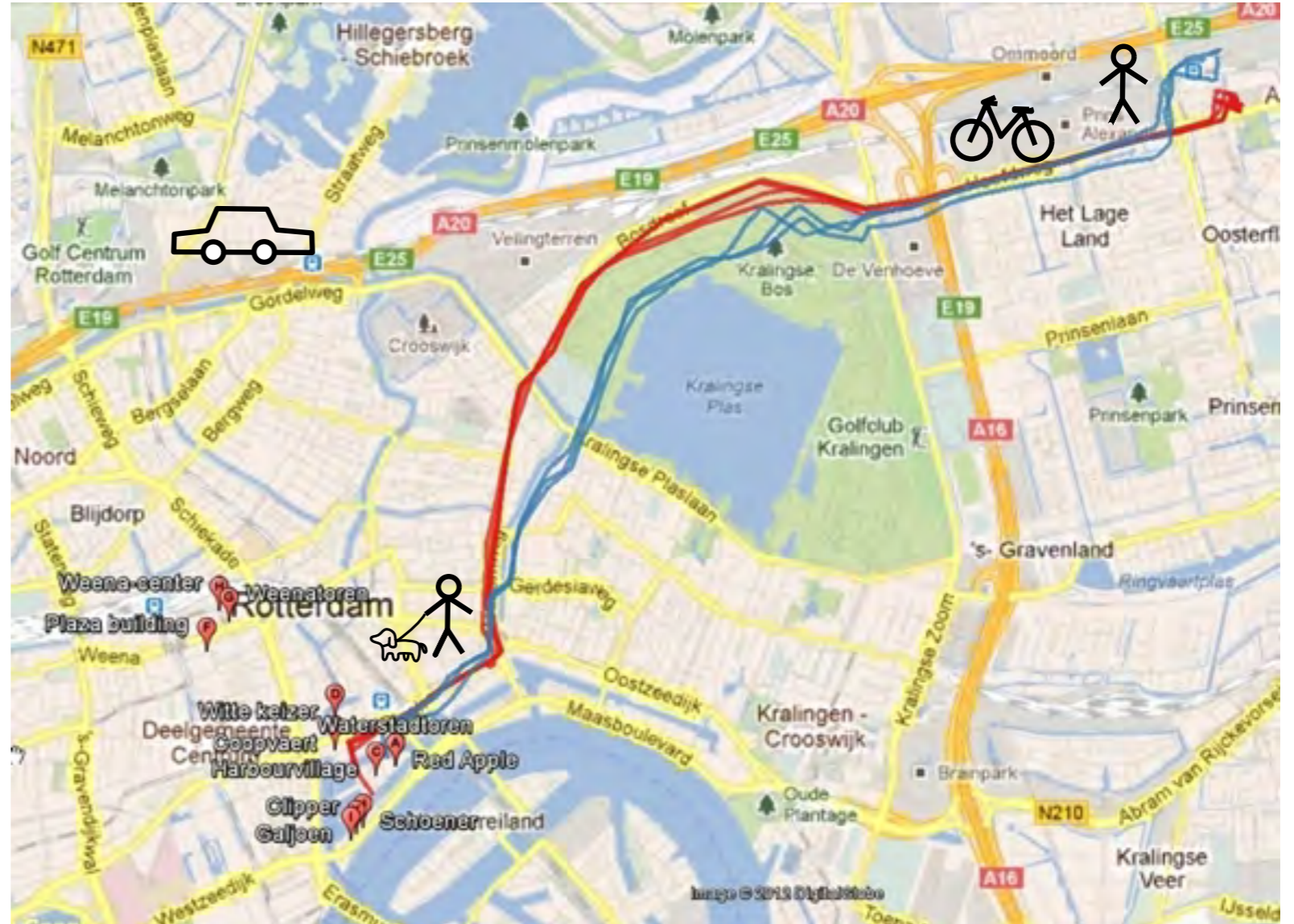


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many entities moving
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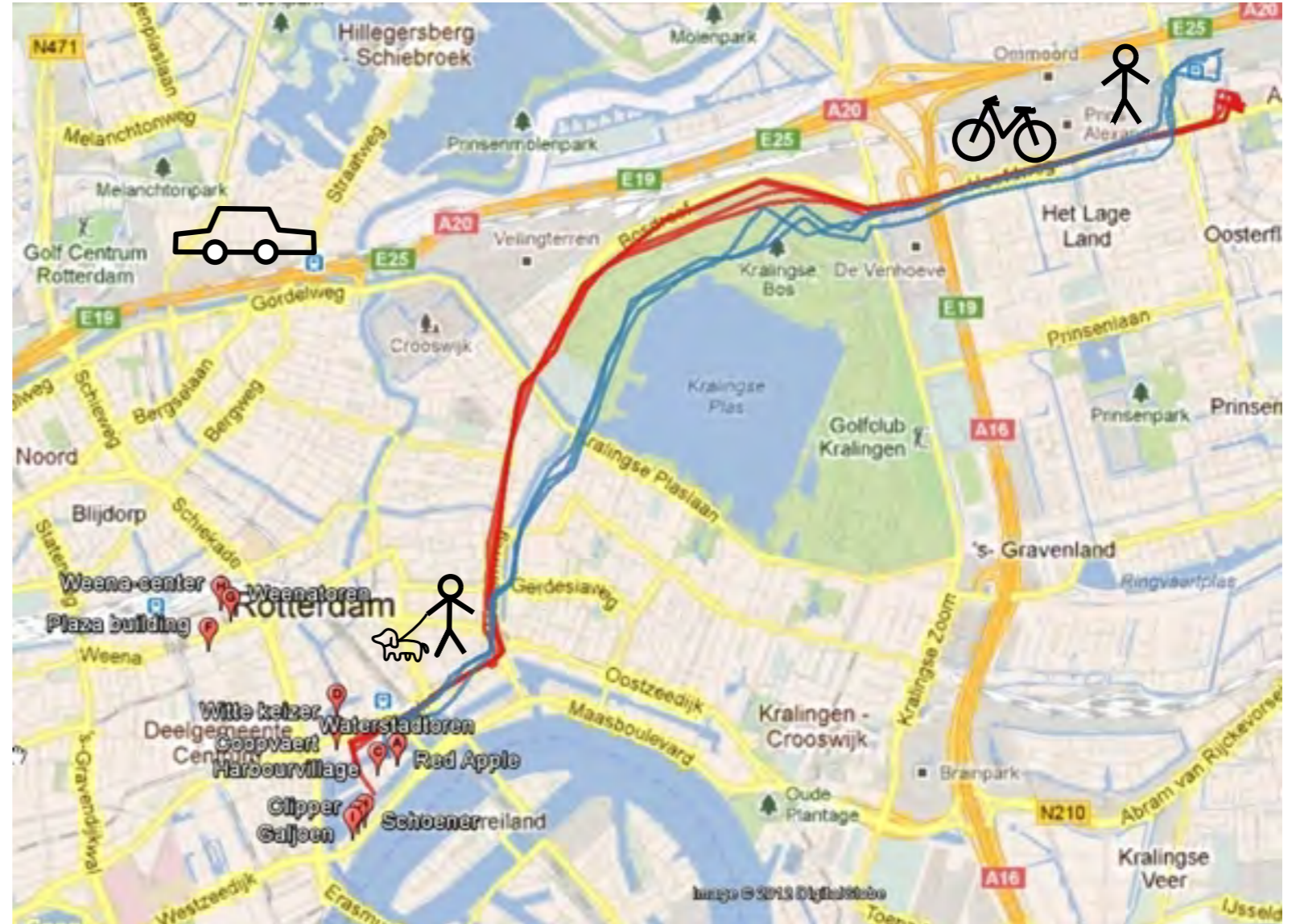


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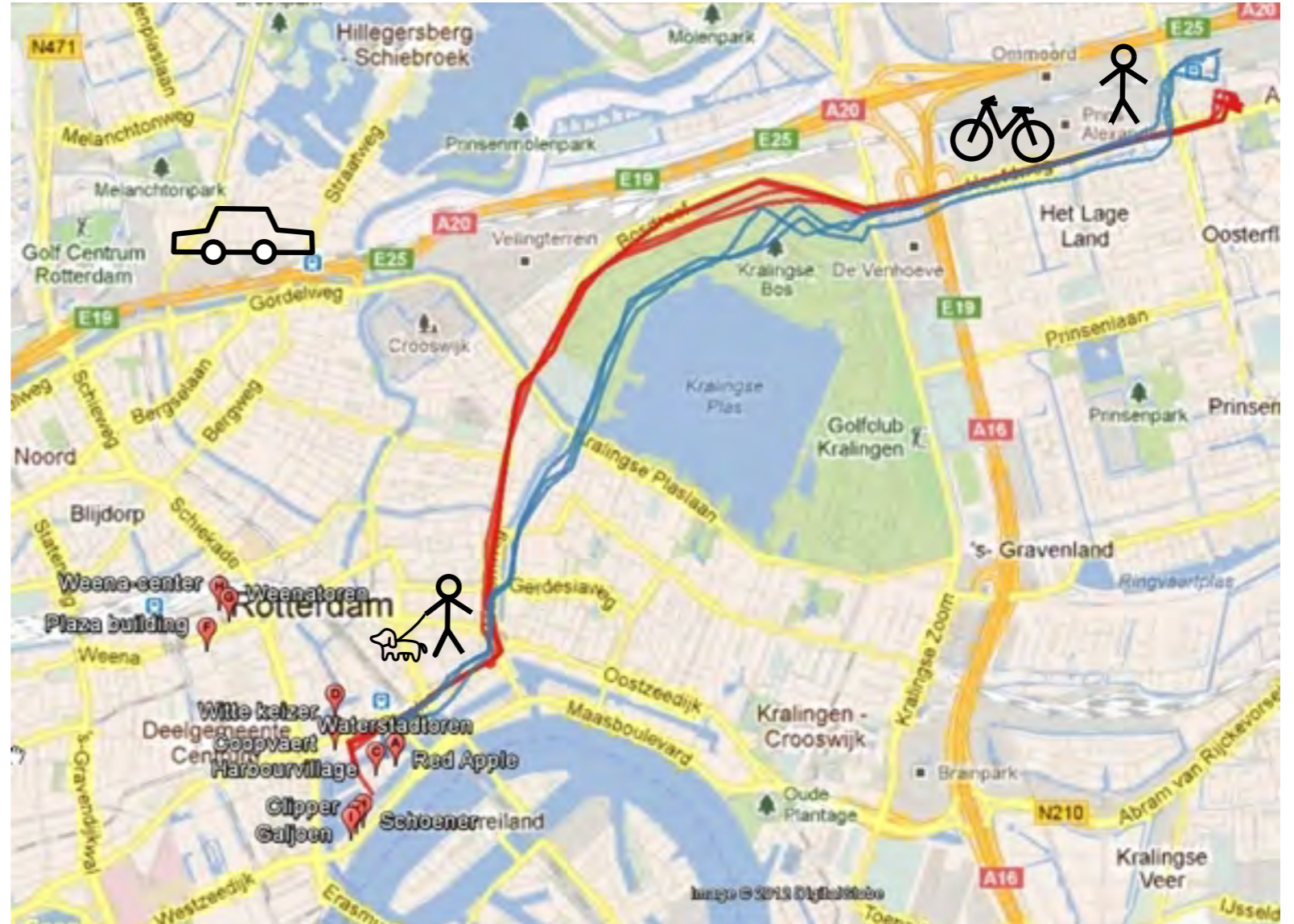


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- can we reconstruct the network?
- how to compare two networks?



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much more research not mentioned here

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