

Discretization of Euler's equations for incompressible fluids through semi-discrete optimal transport.

Quentin Mérigot

Joint works with Jean-Marie Mirebeau et Thomas Gallouët

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(... and borrowing many ideas from Yann...)

Solutions to Euler's equations as geodesics in \mathcal{SDiff}

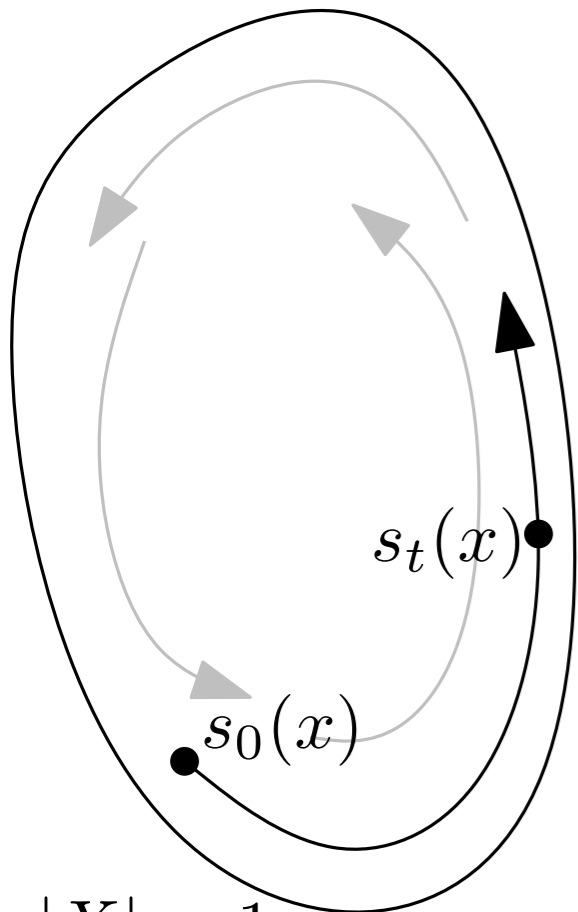
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[Arnold 1966]

$X \subseteq \mathbb{R}^d$ bounded

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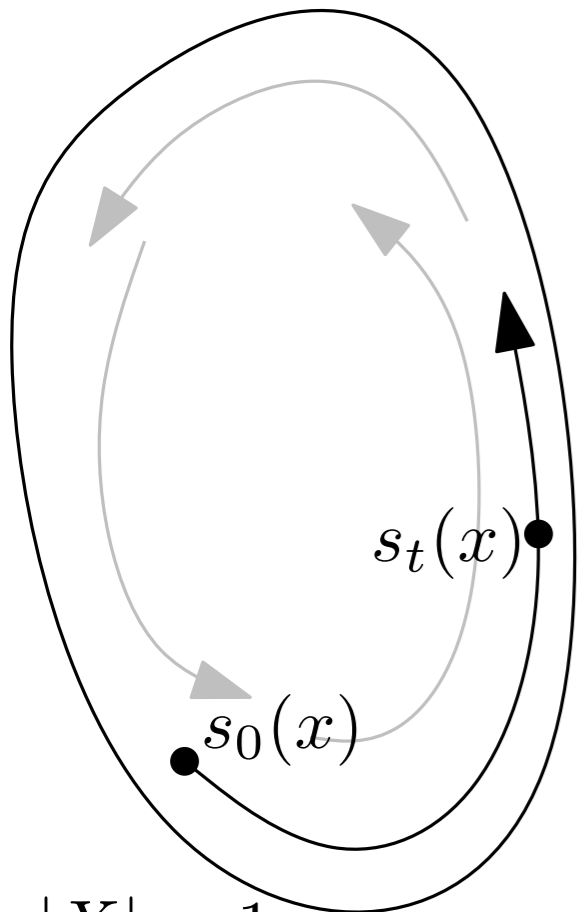
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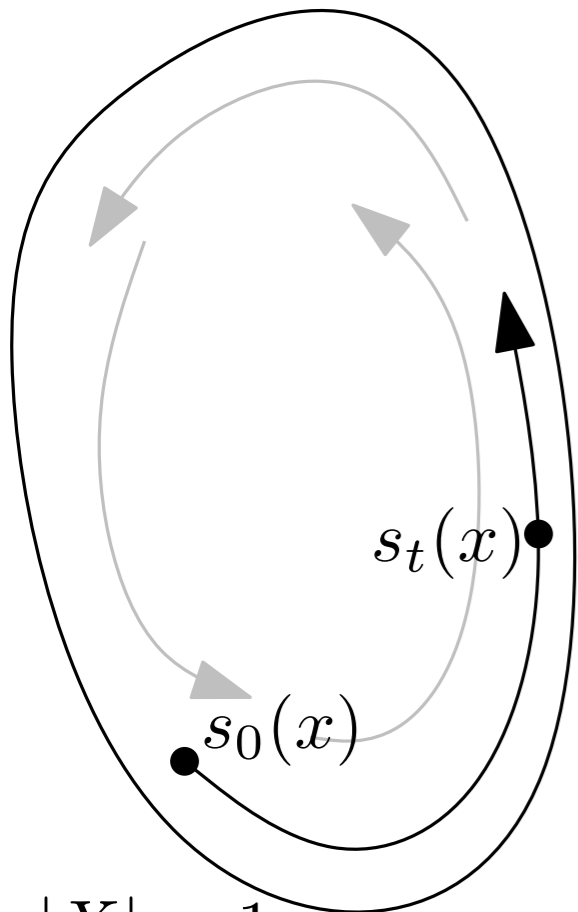
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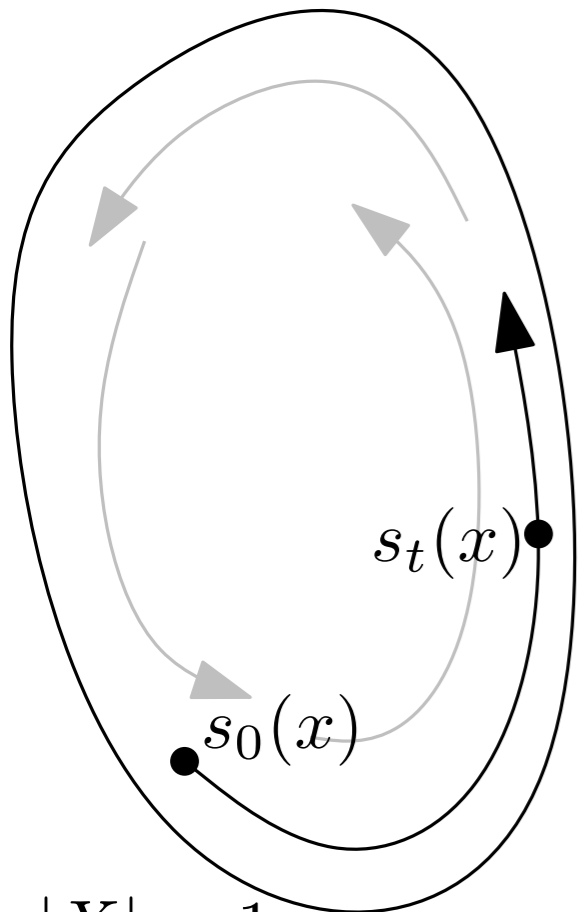
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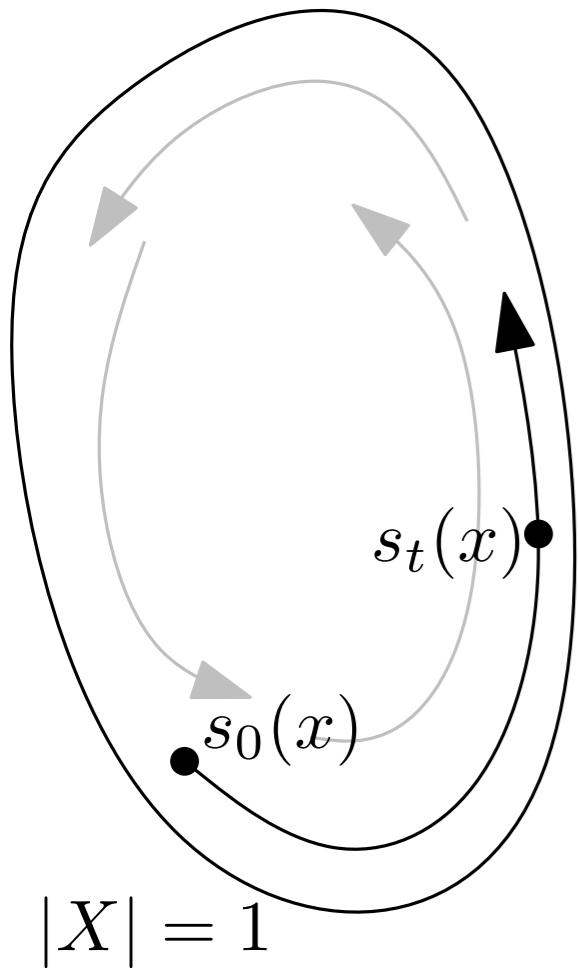
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→ With $u_t := \dot{s}_t \circ s_t^{-1}$ (= velocity in Eulerian coordinates), one recovers **Euler's equations** for incompressible fluids:

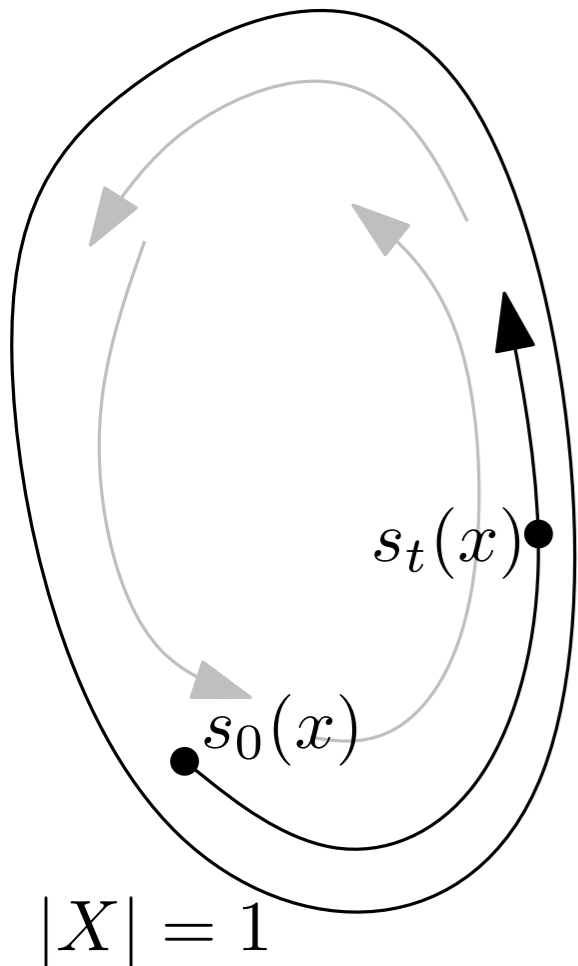
$$\begin{cases} \partial_t u + (u \cdot \nabla) u = -\nabla p & \text{in } X \\ \text{div} u = 0 & \text{in } X \\ u \cdot n = 0 & \text{on } \partial X \end{cases}$$

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This talk: Using this formulation for numerical computations (following Brenier):

- Minimizing geodesics (with Jean-Marie Mirebeau, 2015)
- Cauchy problem (with Thomas Gallouet, 2016).

1. Discretization of the Cauchy problem

Joint work with Thomas Gallouët

Brenier's approximation of geodesics I

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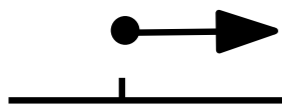
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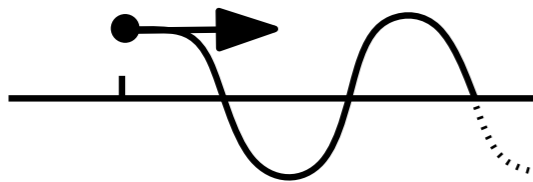
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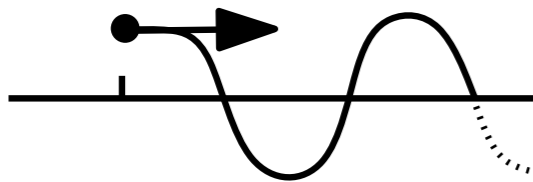
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→ \mathcal{C}^1 convergence towards the geodesic requires $\frac{h}{\varepsilon} \longrightarrow 0$.

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Leb = restriction of Lebesgue measure to a compact domain X

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(NB: in \mathbb{R}^d ($d \geq 2$), each iterations costs N^3 ... → different approach needed)

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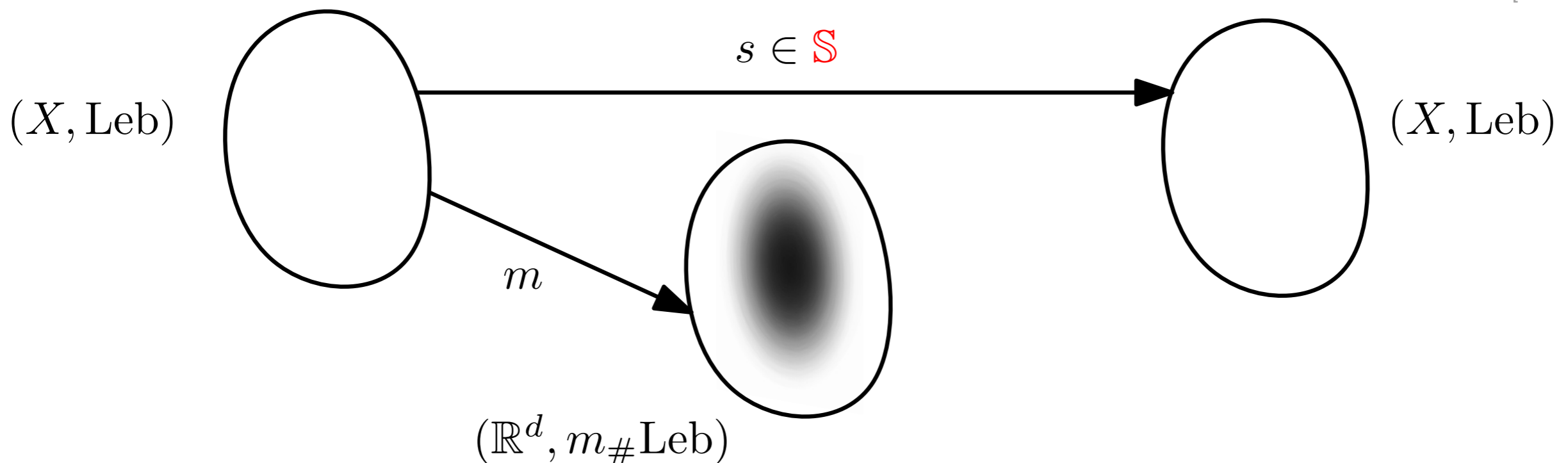
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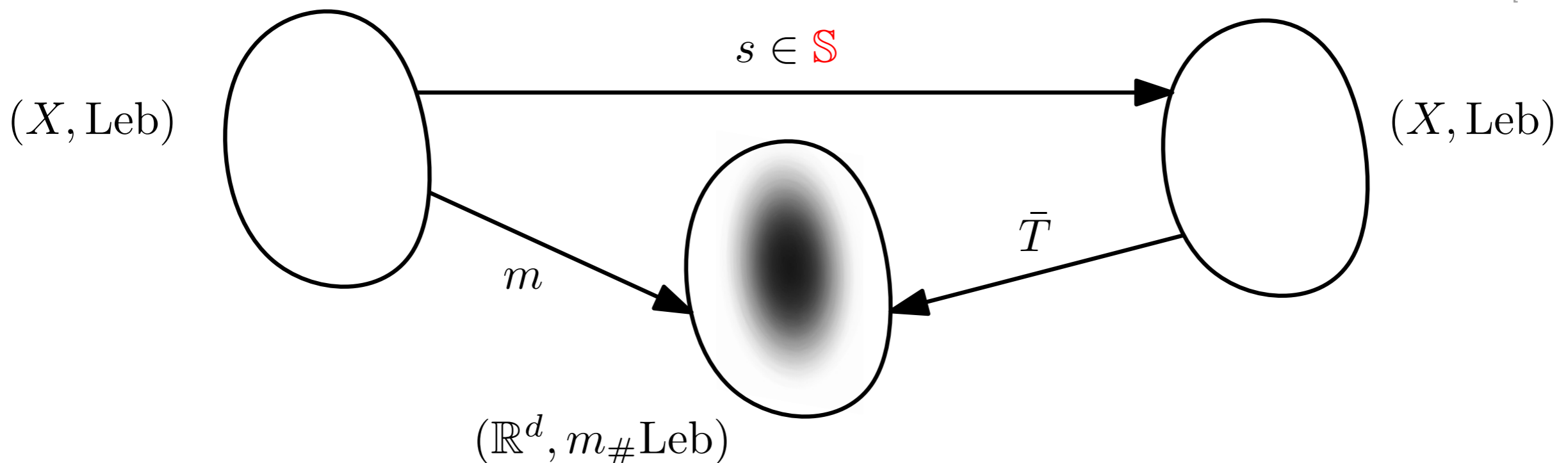
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Let \bar{T} be the quadratic optimal transport map between Leb and $m_{\#}\text{Leb}$. Then,

$$\Pi_{\mathcal{S}}(m) = \{\bar{s} \in \mathcal{S} \mid \bar{T} \circ \bar{s} = m\}$$

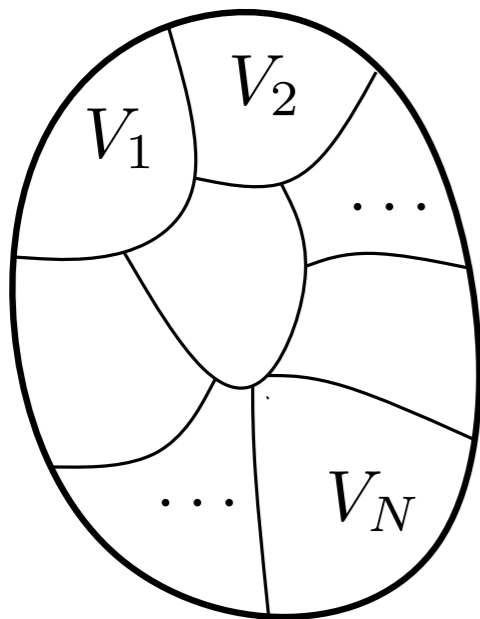
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Objective: Constructing a finite-dimensional subspace of \mathbb{M} and computing $\Pi_{\mathcal{S}}$

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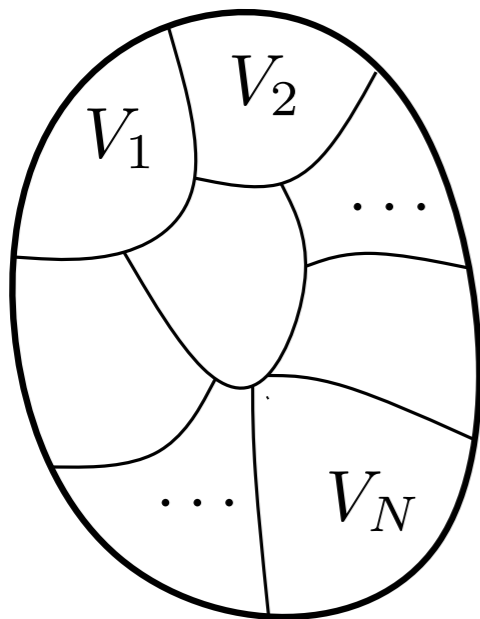
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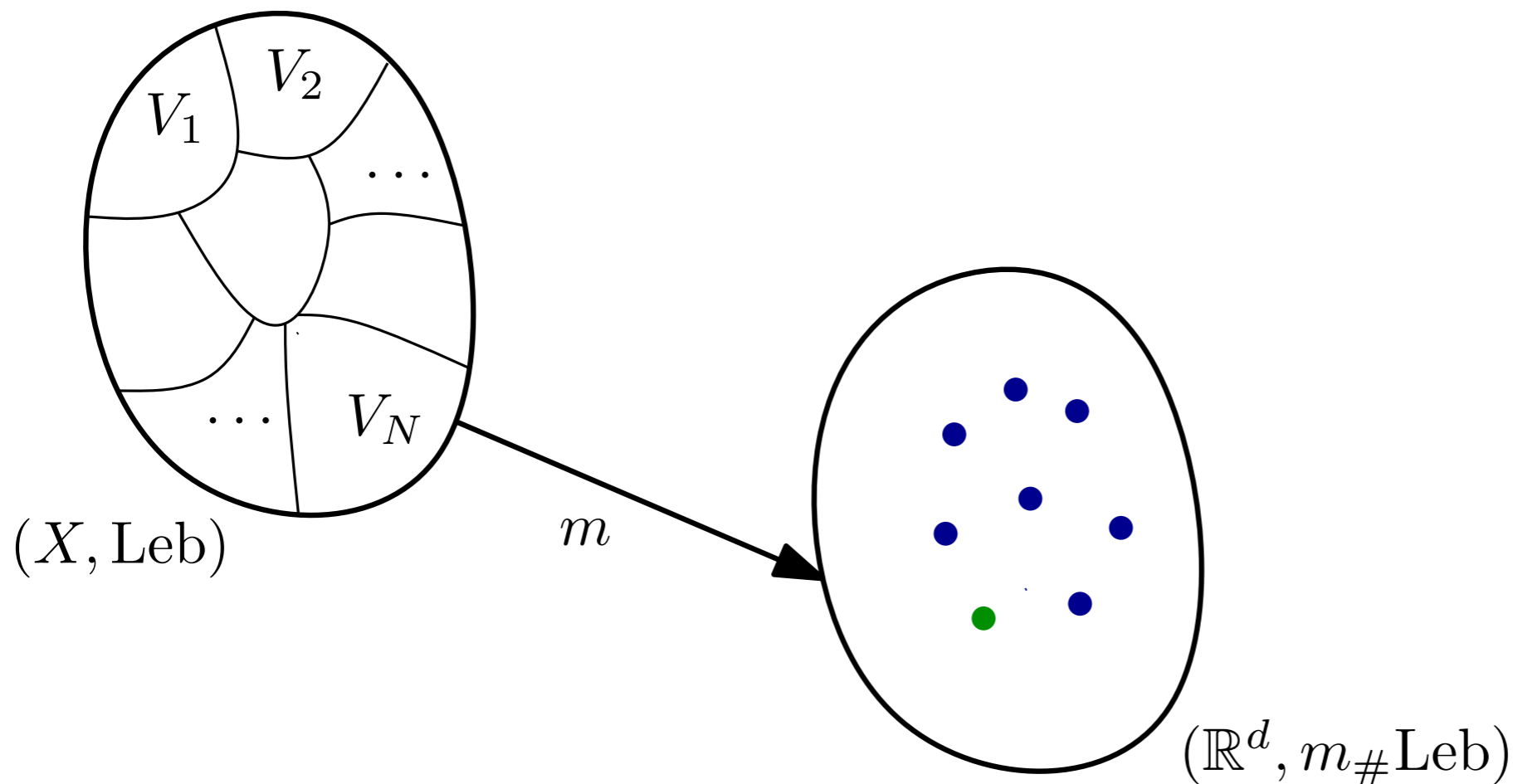
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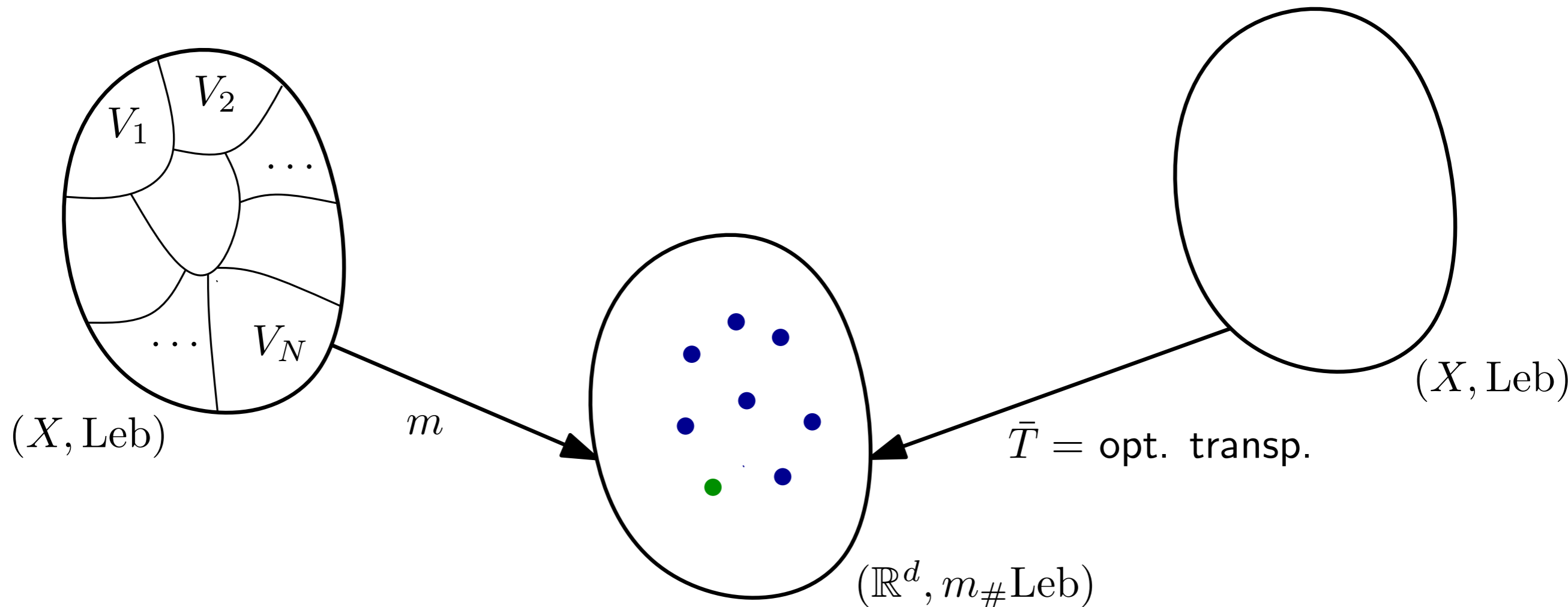
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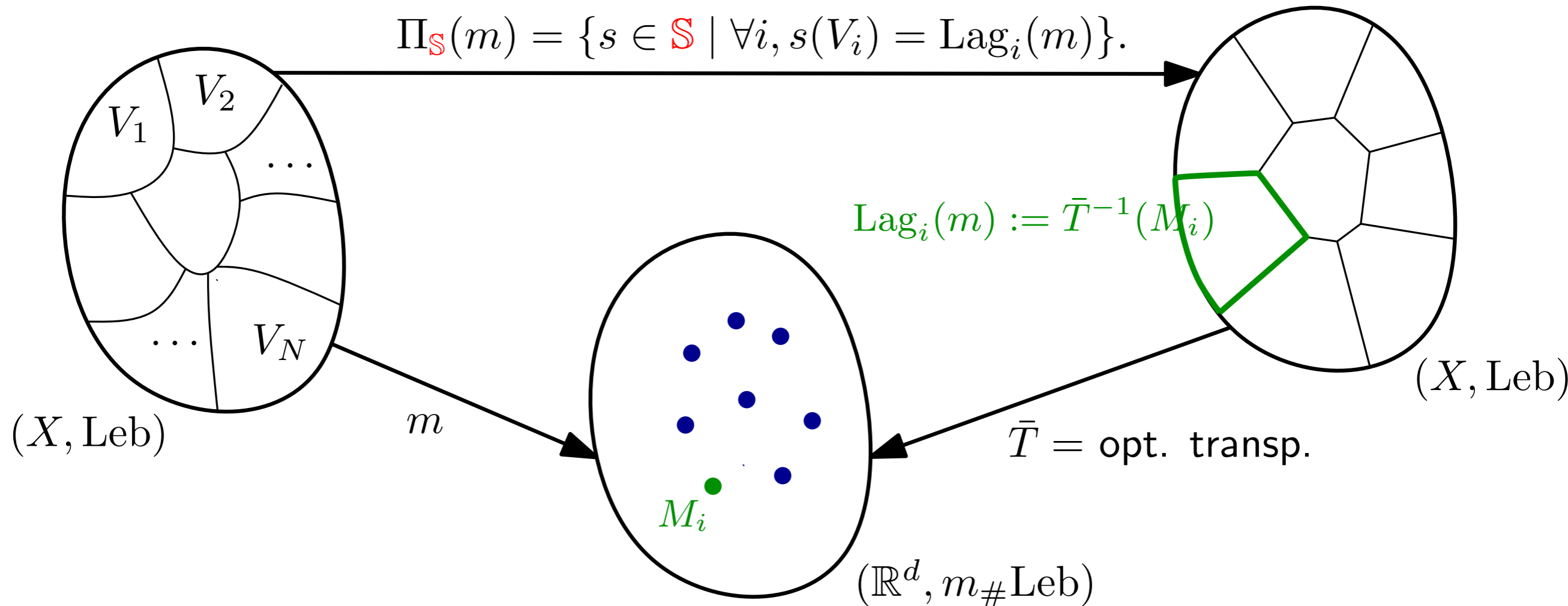
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Convergence of the space-discretization

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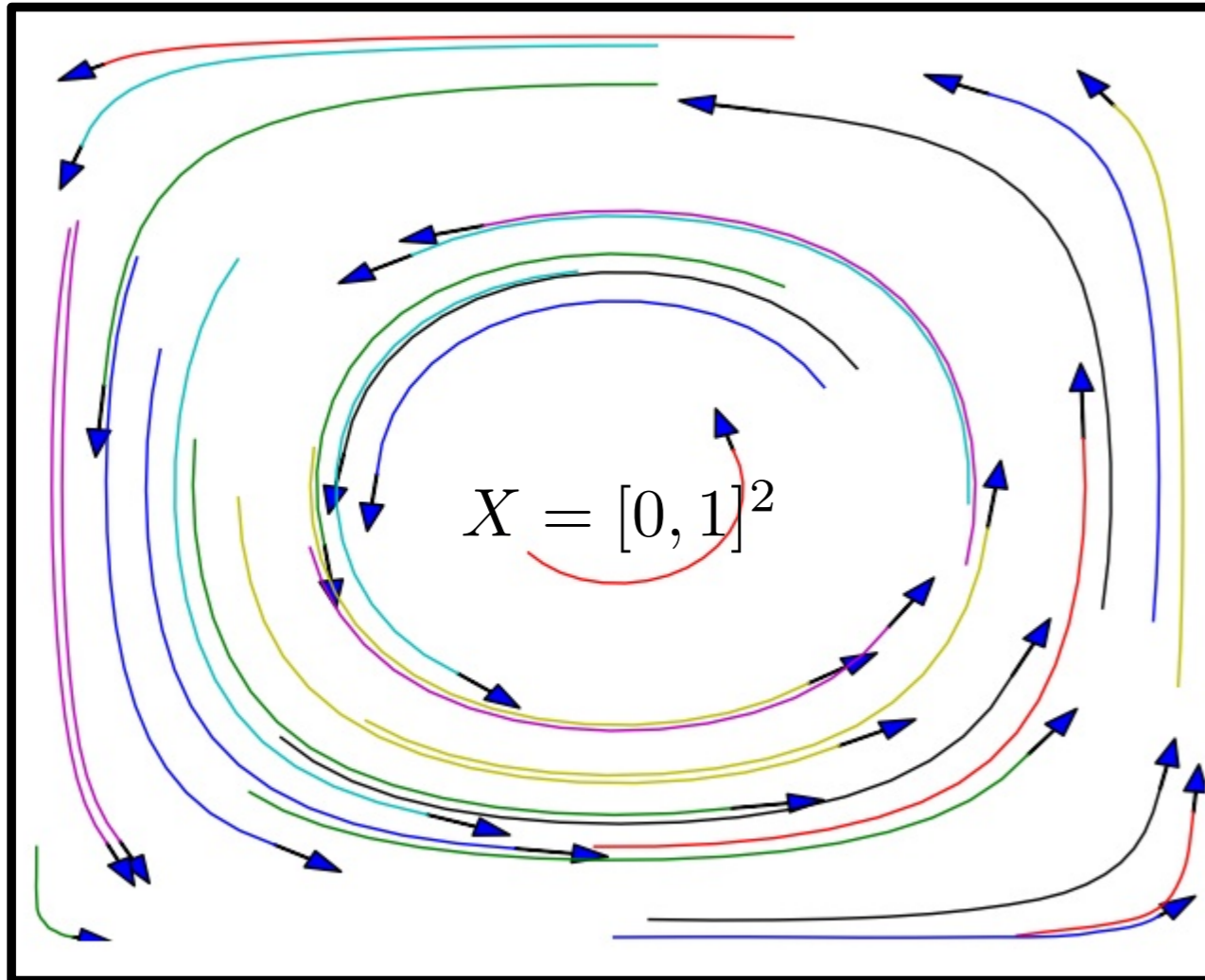
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→ Convergence of a time-discretization using the symplectic Euler scheme.

Numerical result: Stationary flow on $[0, 1]^2$

Stationary flow on $[0, 1]^2$: speed: $u(\mathbf{x}) = (\cos(\pi x_1) \sin(\pi x_2), \sin(\pi x_1) \cos(\pi x_2))$
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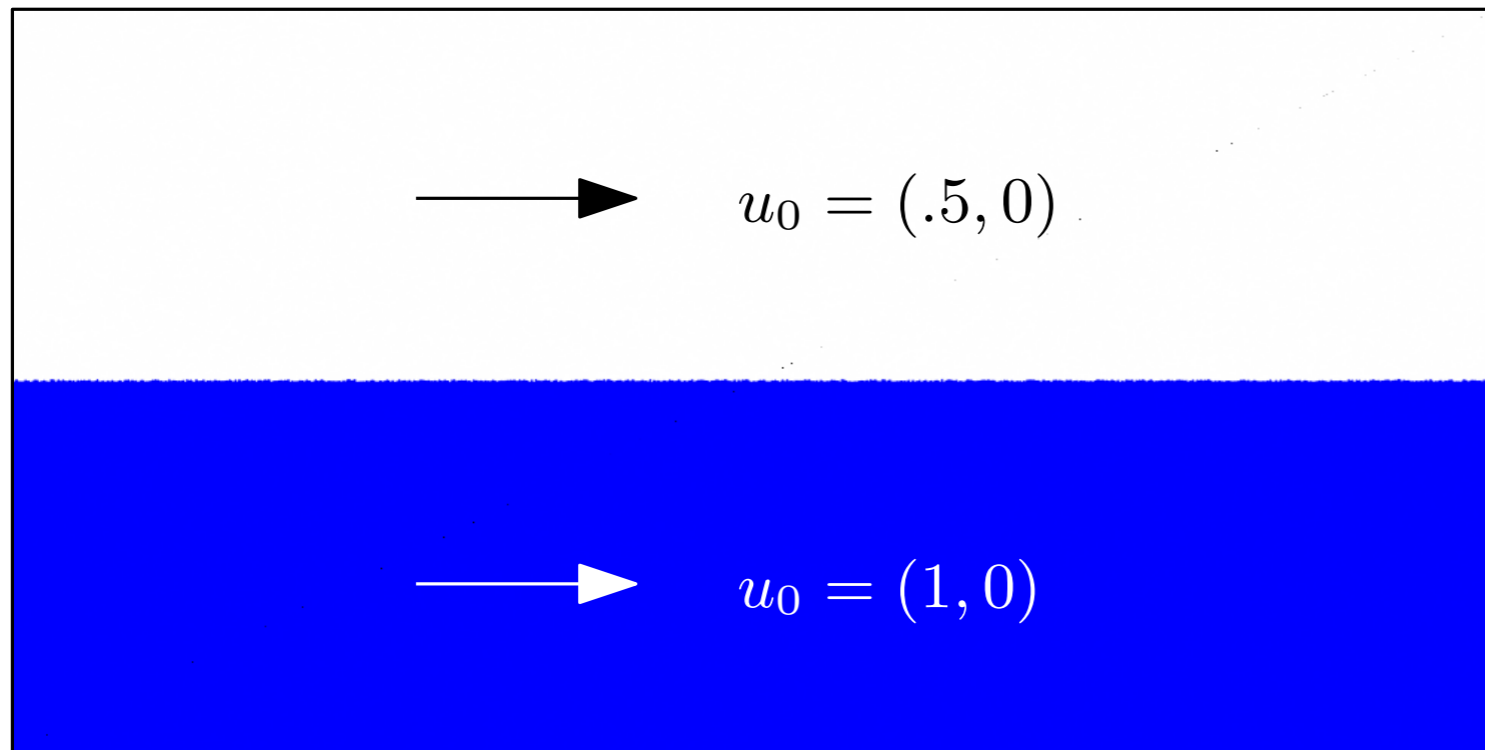
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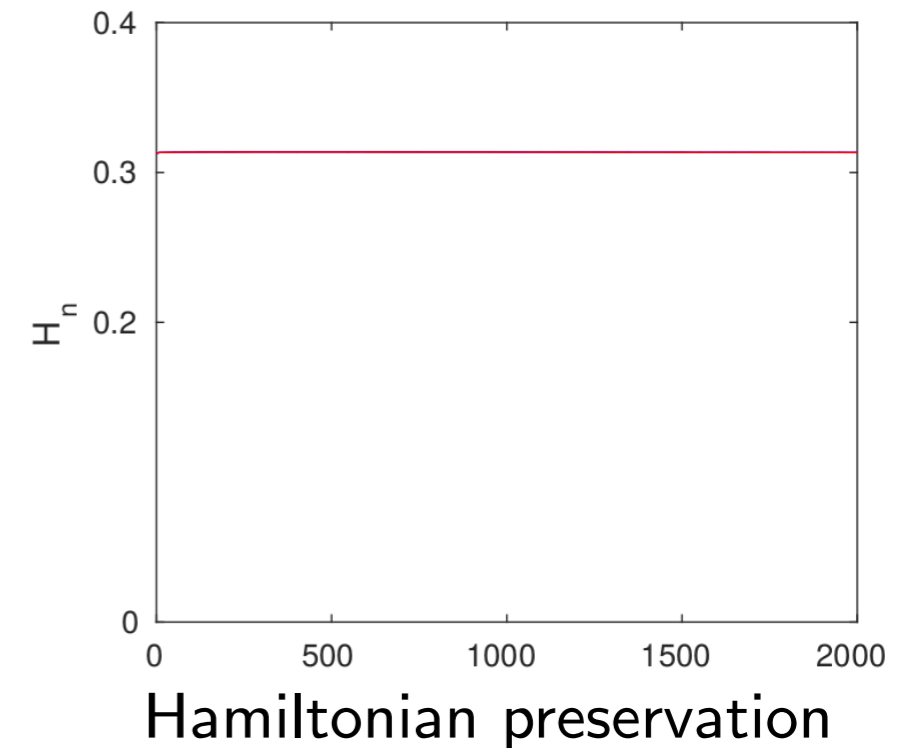
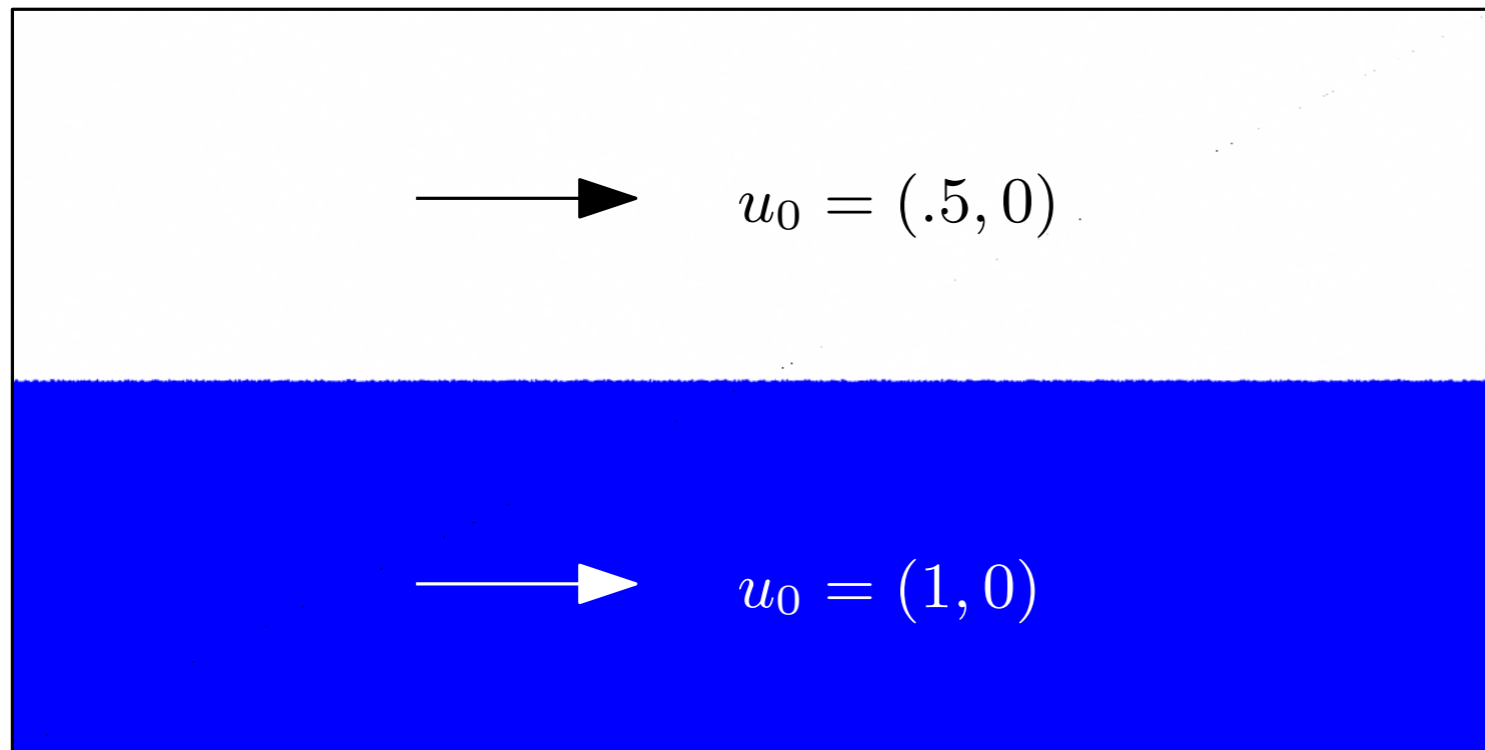
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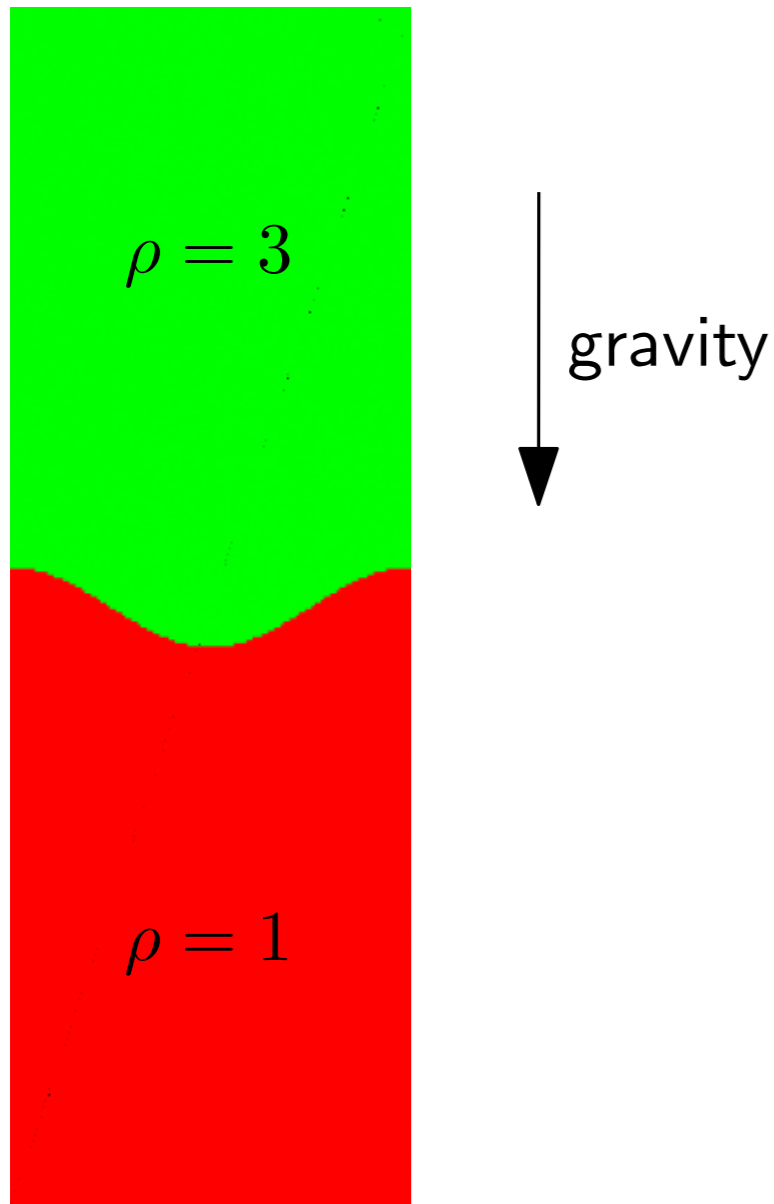
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B. Rayleigh-Taylor instability (Inhomogeneous fluid)



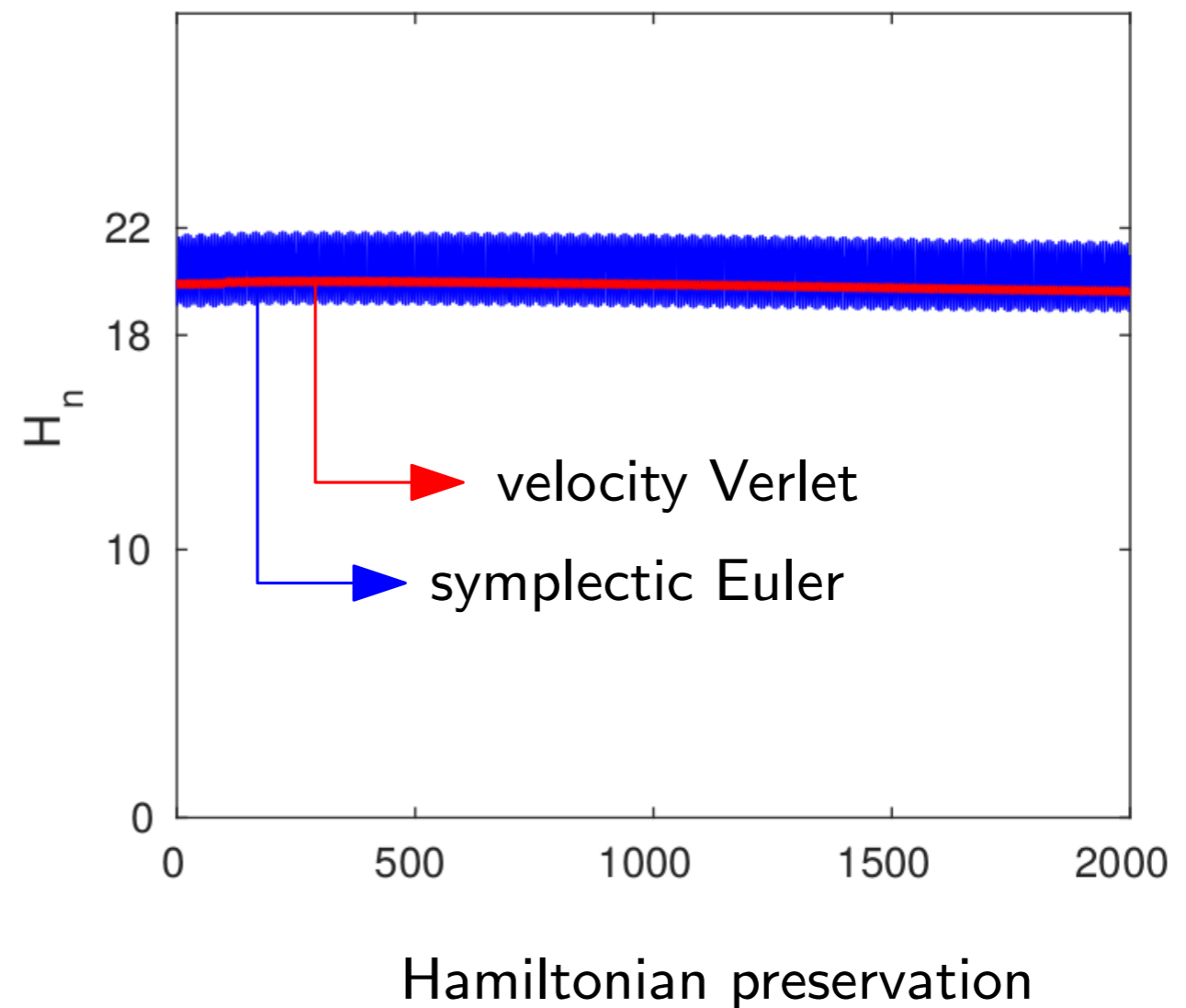
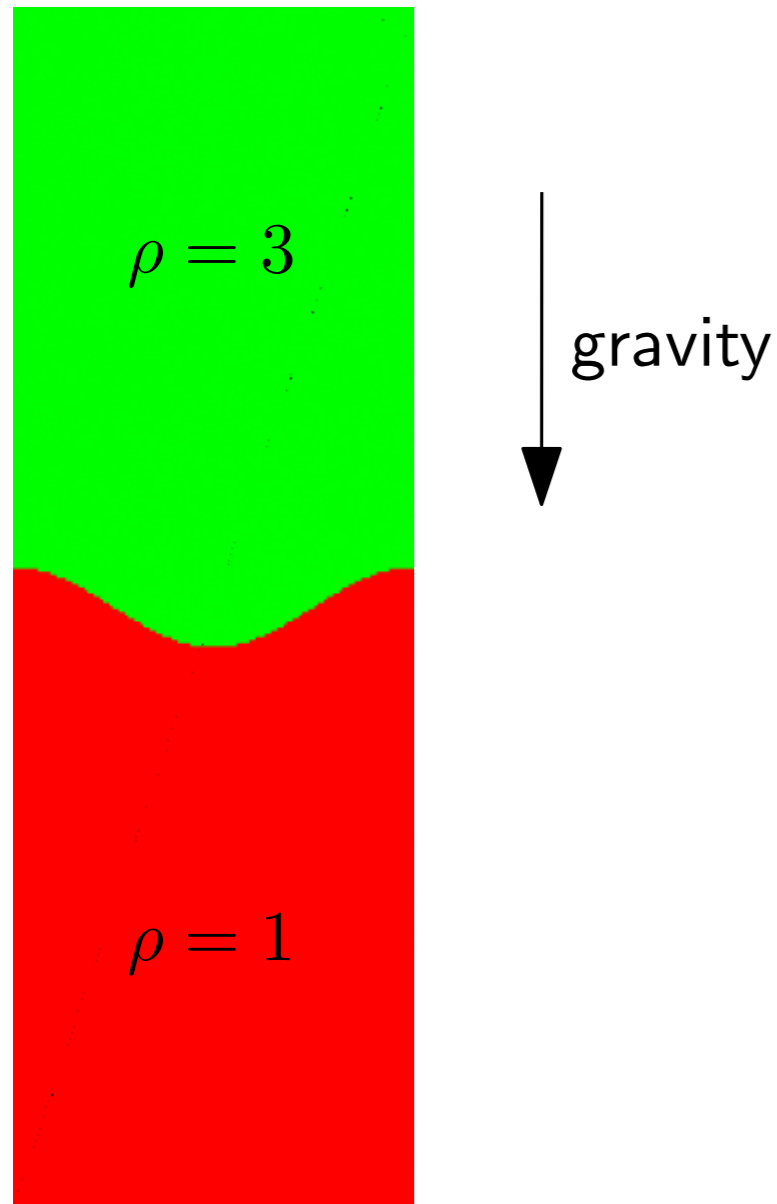
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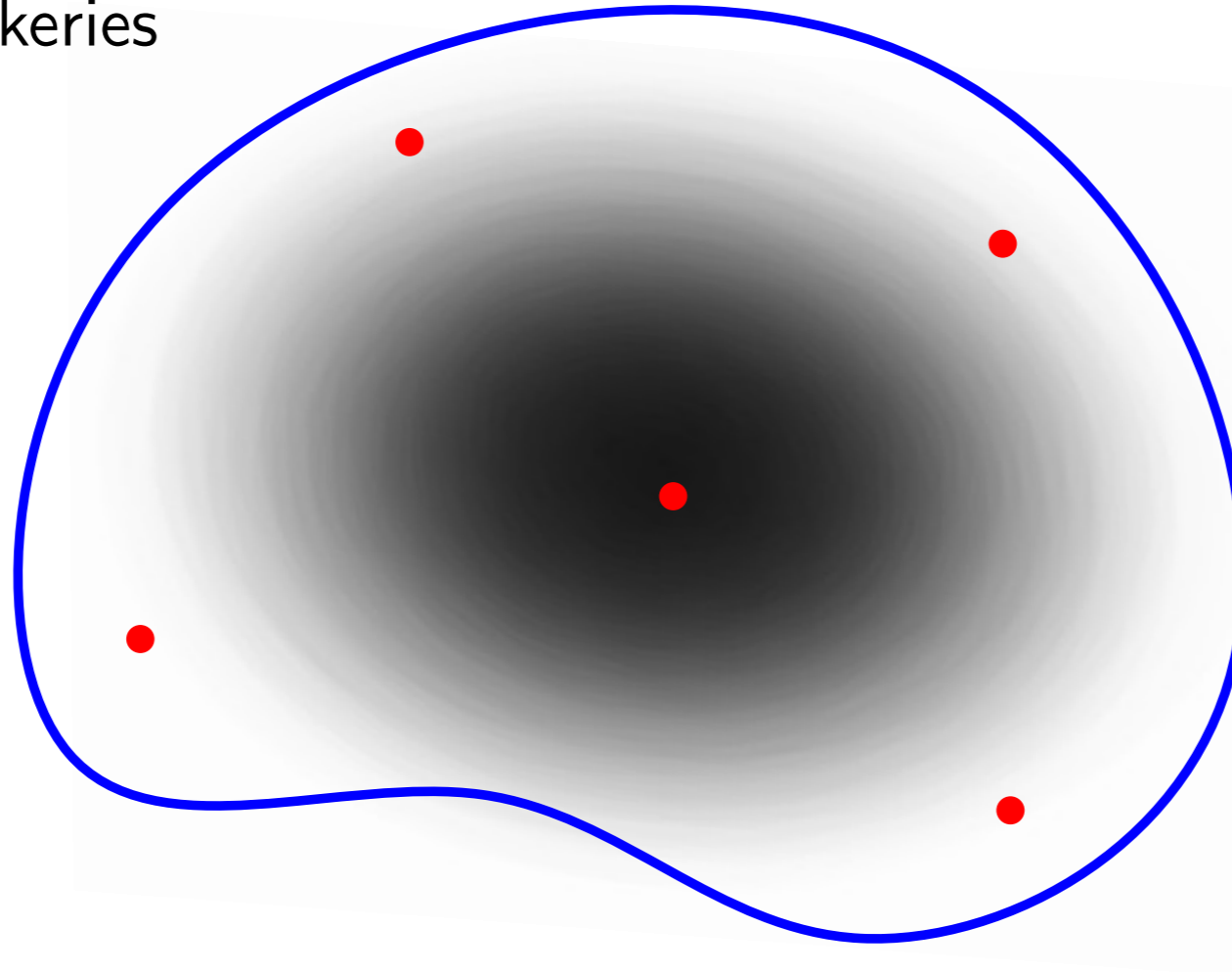
2. Semi-discrete optimal transport

An economic metaphor

$\rho : X \rightarrow \mathbb{R}$ density of population

$c(x, y) := \|x - y\|^2$ cost of walking from x to y

Y = location of bakeries

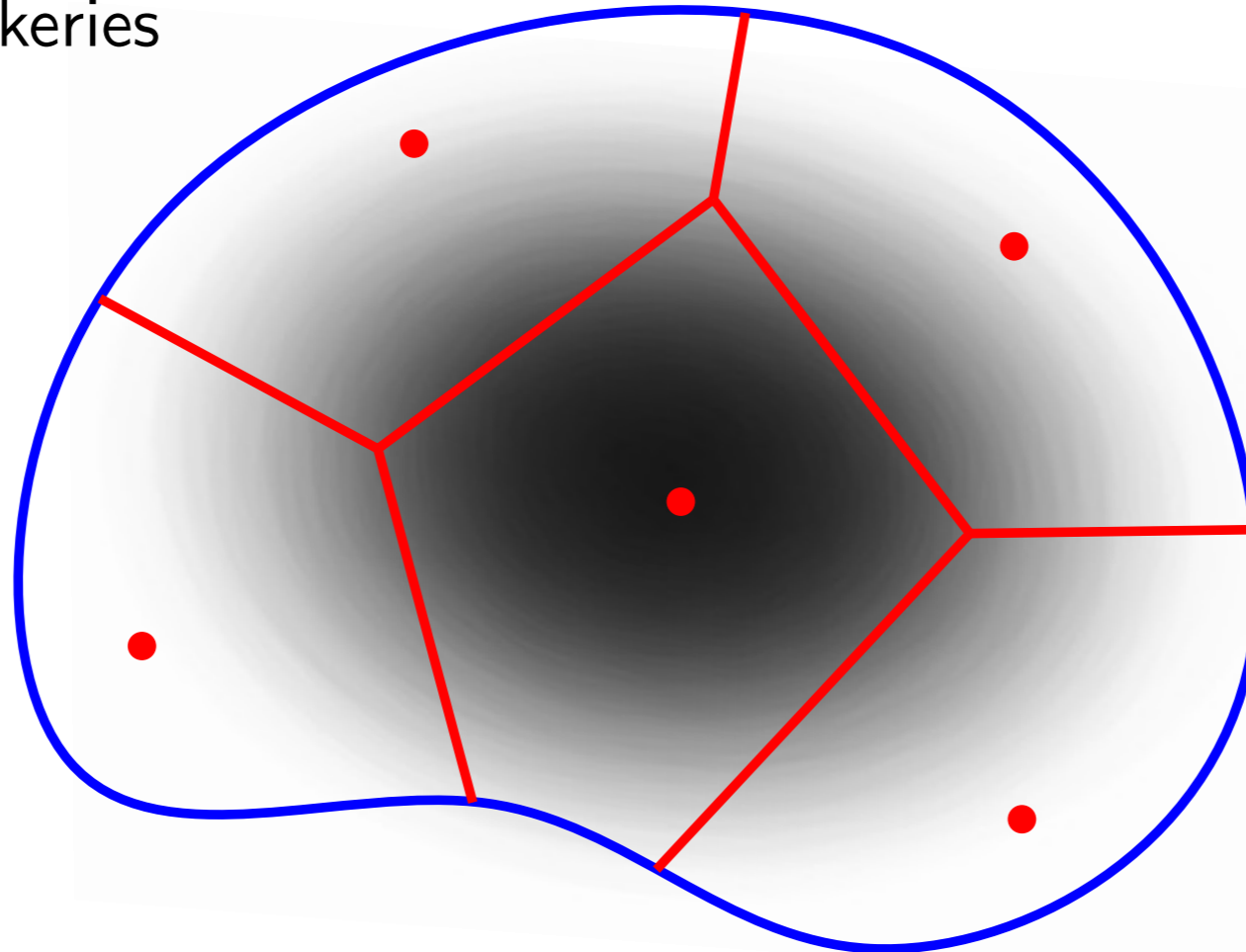


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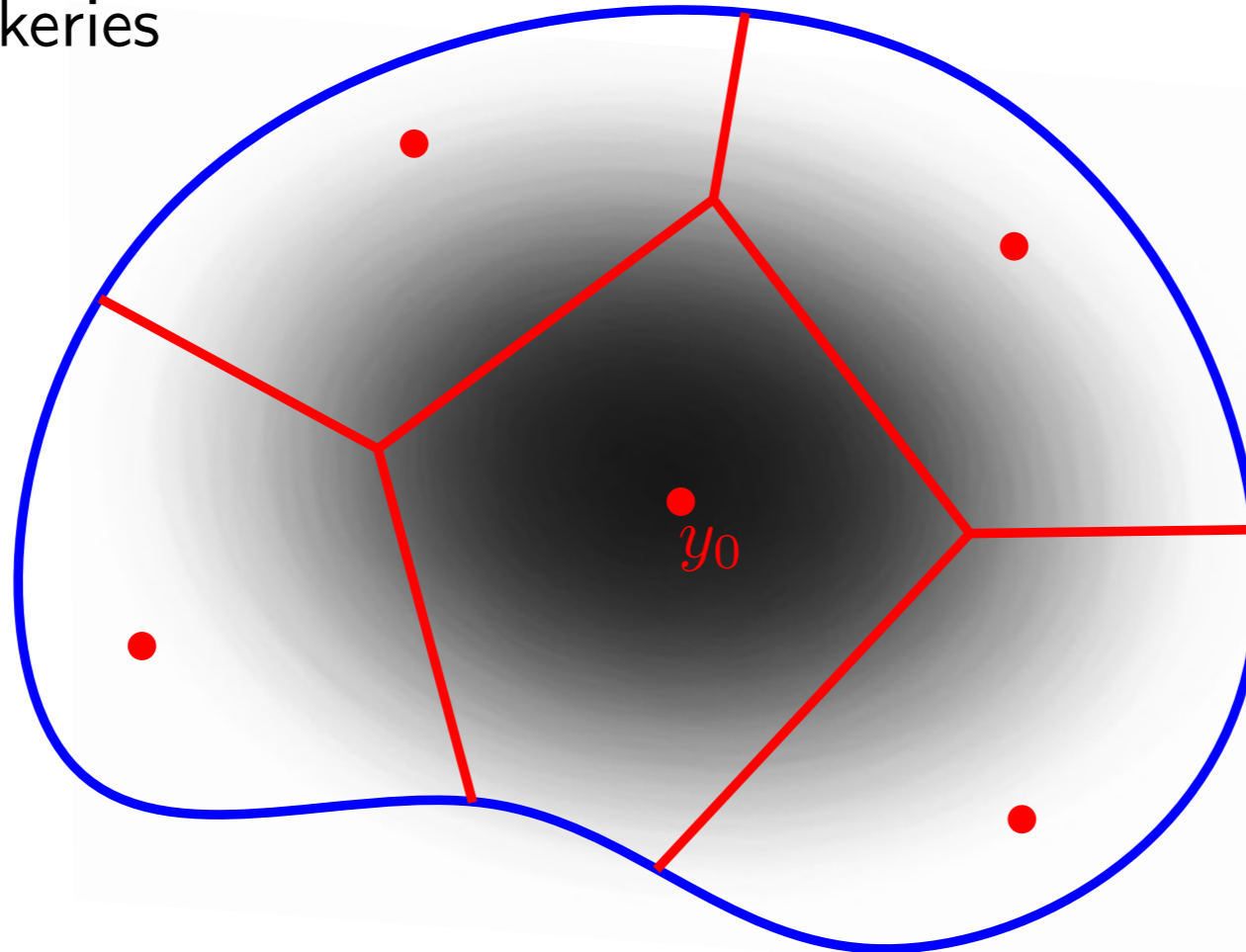
$$\text{Vor}(y) = \{x \in X; \forall z \in Y, c(x, y) \leq c(x, z)\}$$

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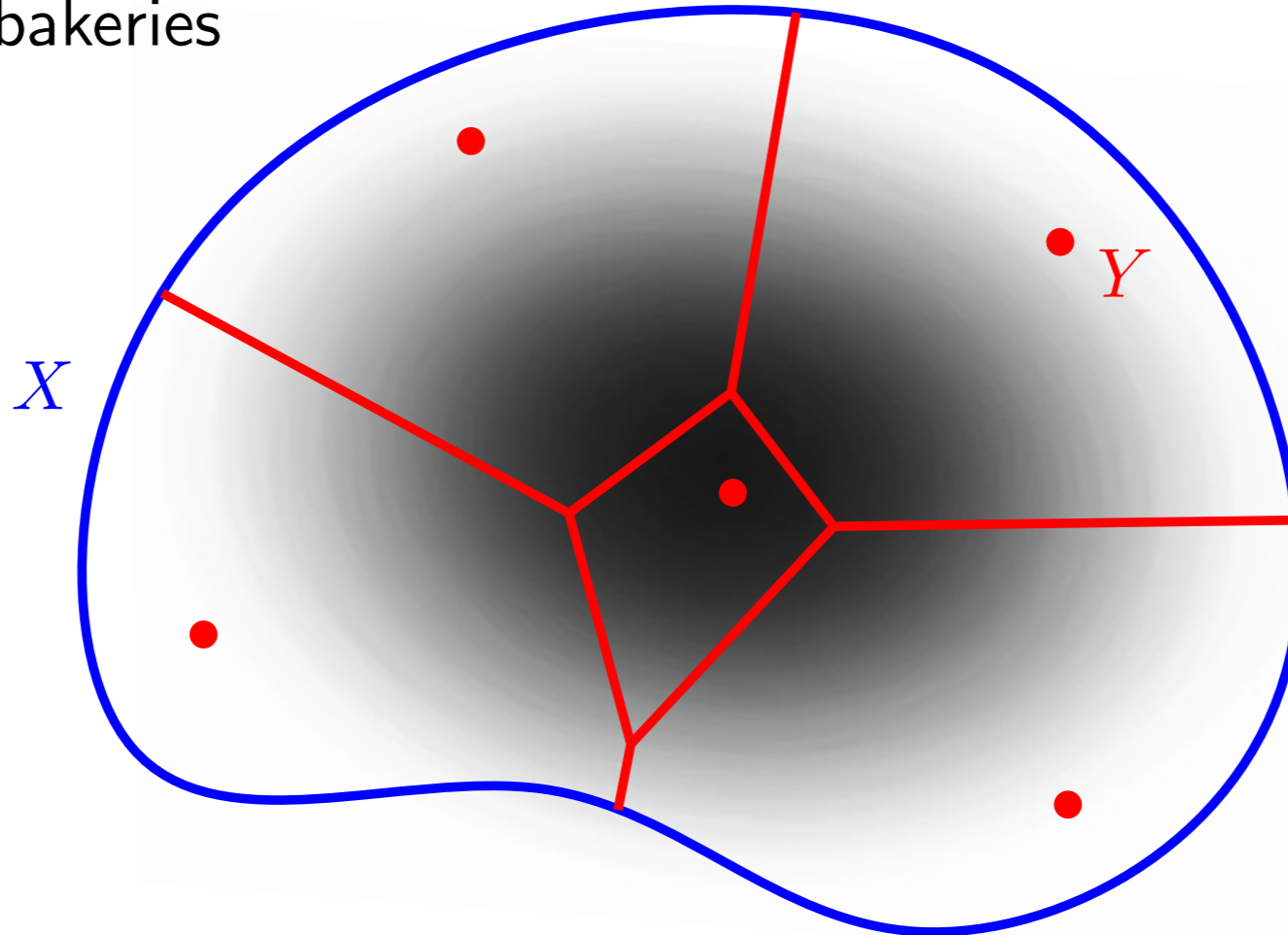
Minimizes total distance walked ... **but** might exceed the capacity of bakery y_0 !

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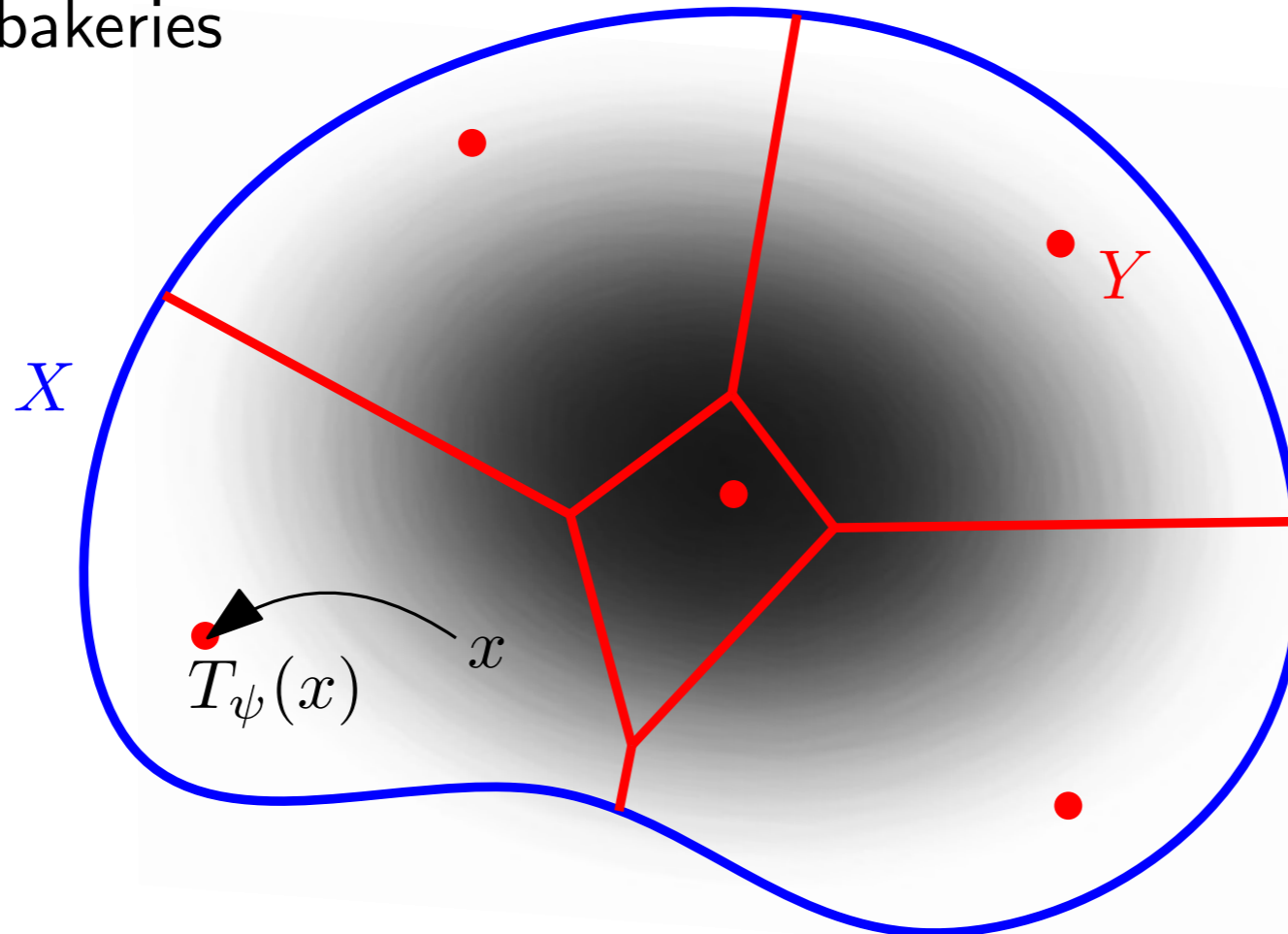
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Lemma: The map T_ψ induced by this decomposition is a c -optimal transport between ρ and $\nu_\psi := T_{\psi\#}\nu = \sum_{y \in Y} \rho(\text{Lag}_y(\psi))\delta_y$.

SD-OT as Concave Maximization

Theorem: Finding an **optimal transport** between ρ and $\nu = \sum_Y \nu_y \delta_y$

\iff finding **prices** ψ on Y such that $\nu_\psi = \nu$

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► Byproduct of Kantorovich duality.

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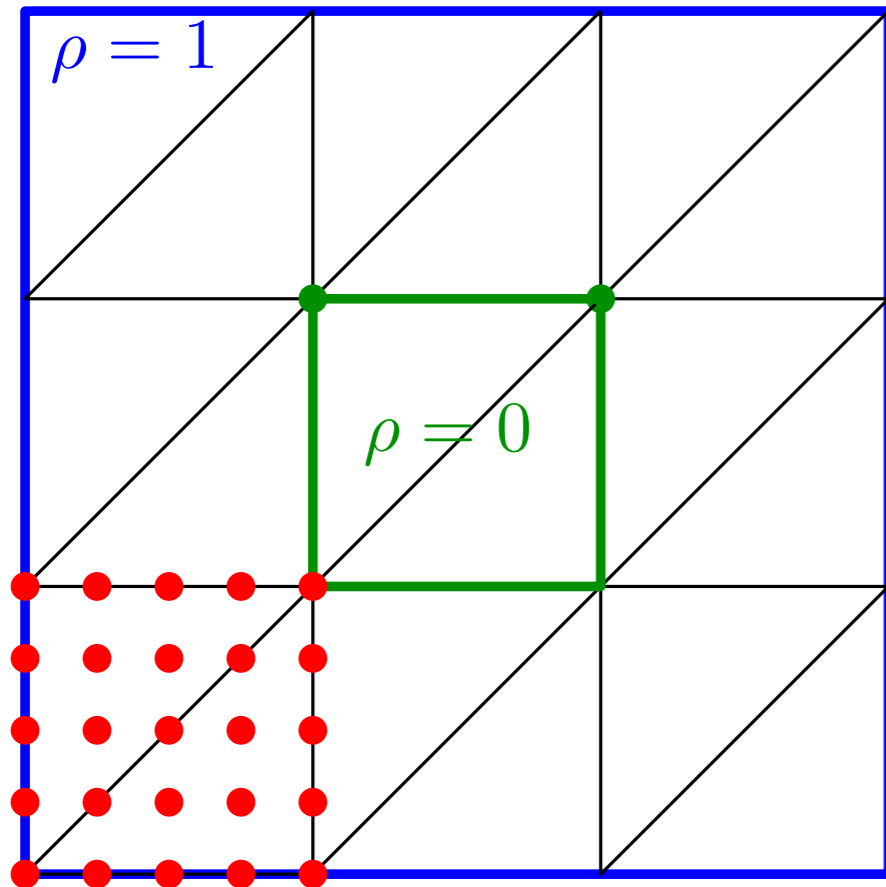
In the simulations, we use a (damped) Newton's algorithm, solving a sequence of linearized **discrete** Monge-Ampère equations.

Numerical example

- ▶ Simple damped Newton's algorithm, with global linear convergence, [Mirebeau 15]
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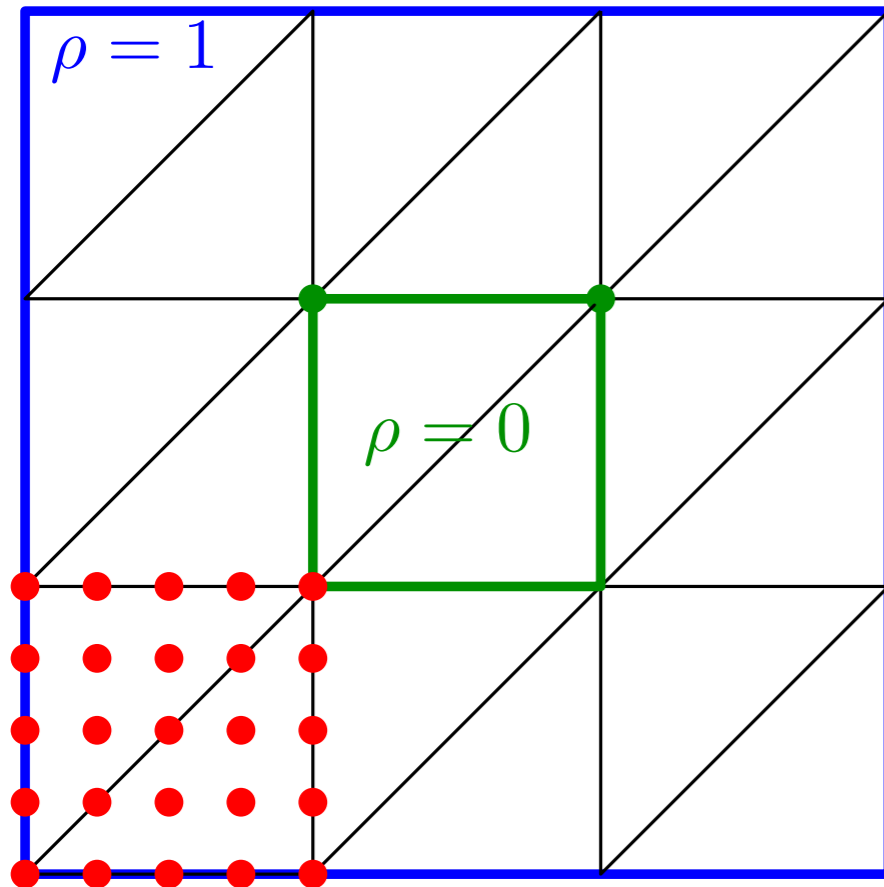


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Target: Uniform grid Y in $[0, 1]^2$.

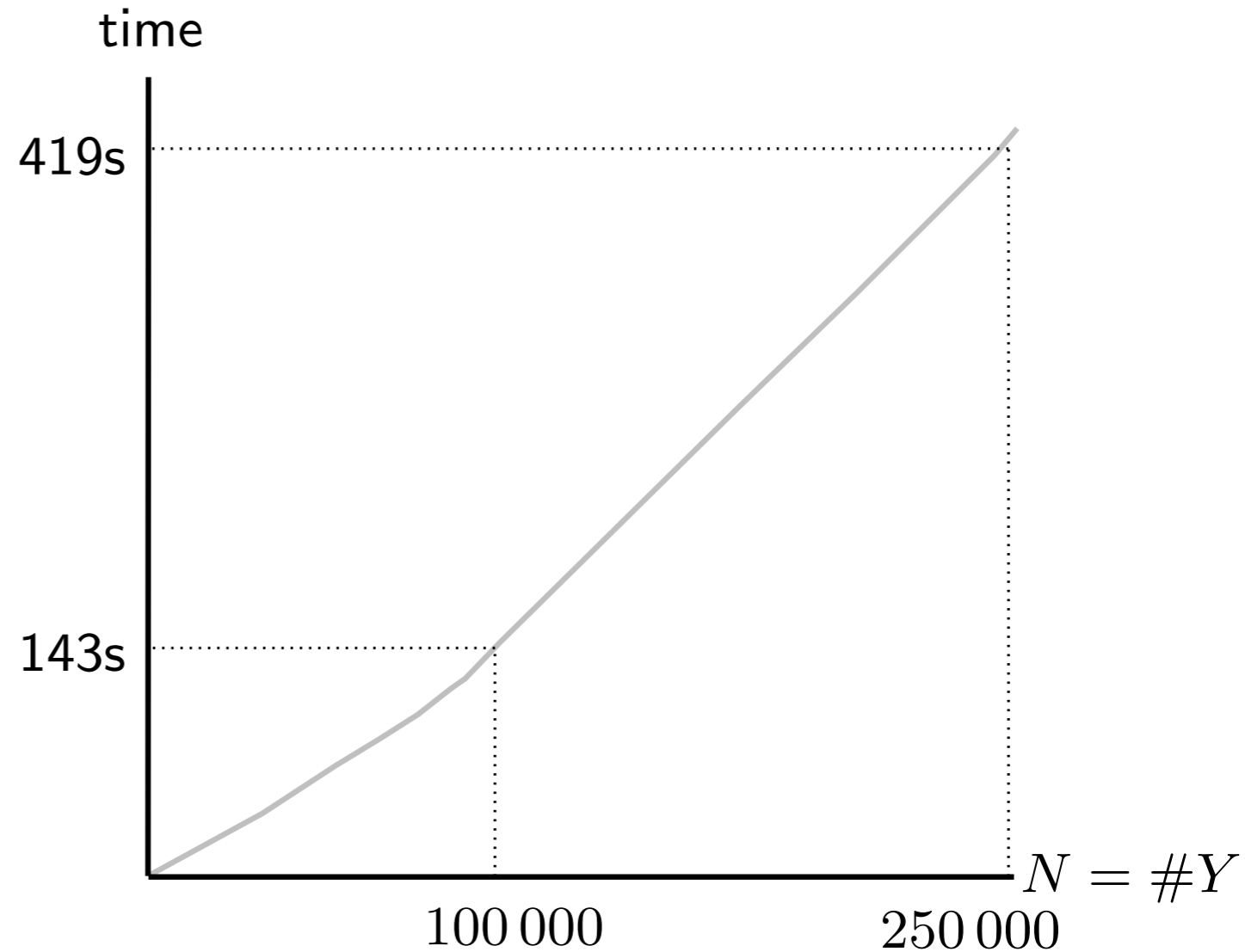
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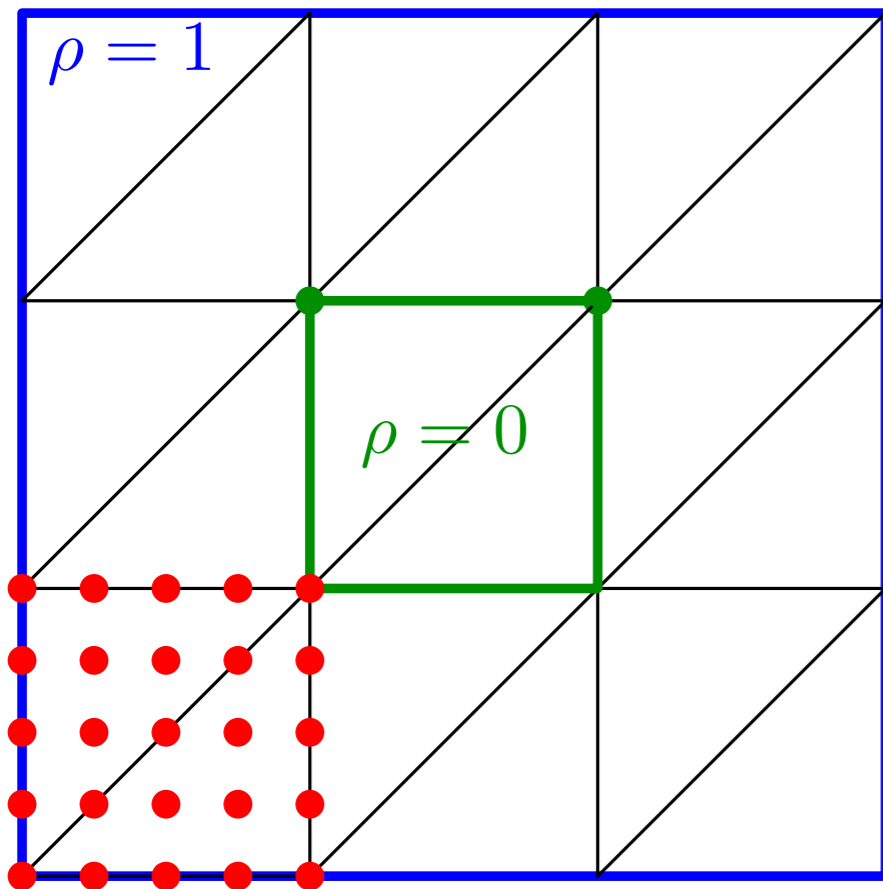
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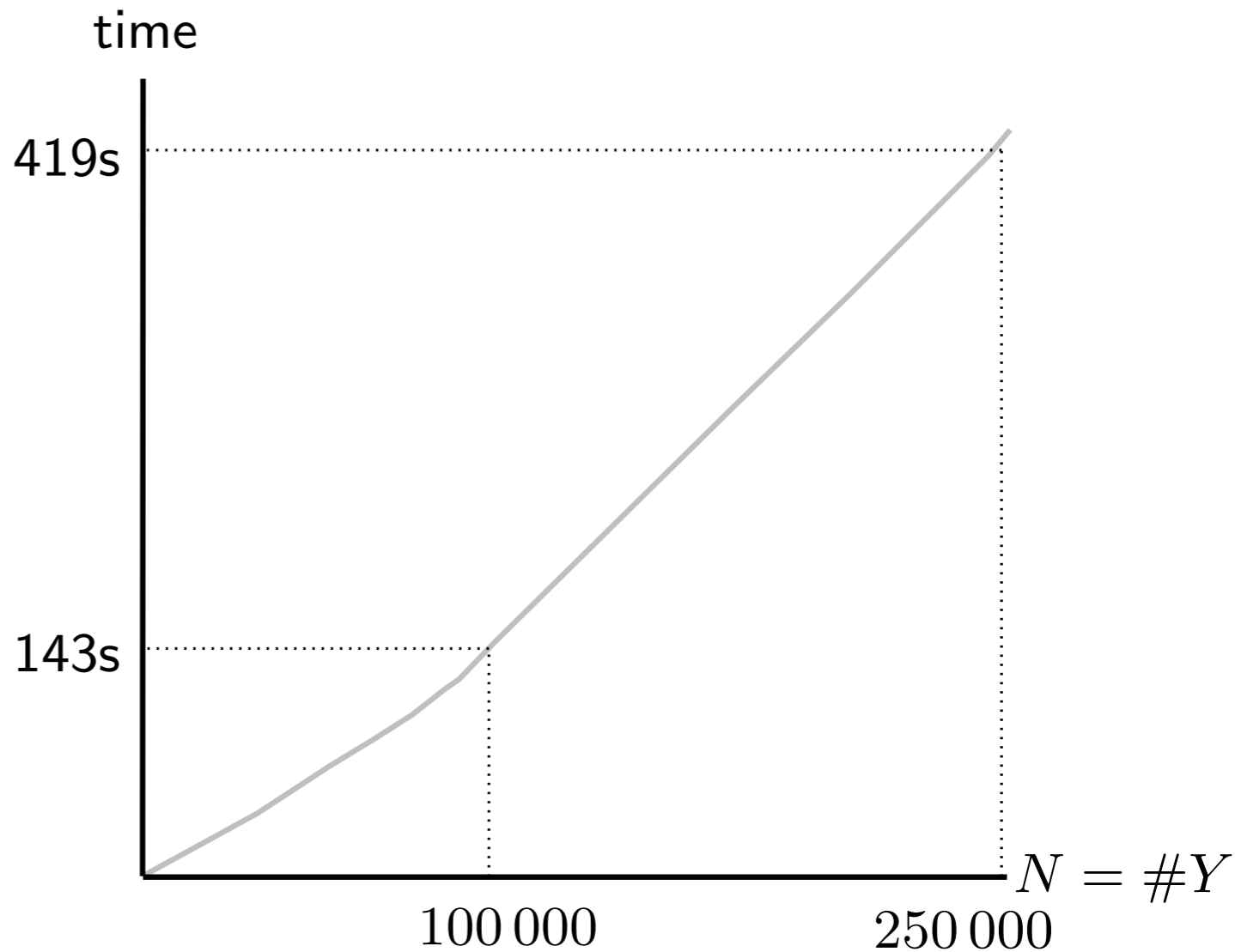
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Near- $\mathcal{O}(N)$ vs $\mathcal{O}(N^3)$ complexity for fully discrete (combinatorial) OT.

3. Minimizing geodesics in \mathcal{SDiff}

Joint work with Jean-Marie Mirebeau

Finite-dimensional example

Let S be a submanifold in \mathbb{R}^d , whose minimizing geodesics need to be approximated.

► **Minimizing geodesics:** $\min_{s:[0,1] \rightarrow \mathbb{R}^d} \frac{1}{2} \int_0^1 \|\dot{s}_t\|^2 dt$ where $\begin{cases} \forall t \in [0, 1], s_t \in S \\ s_0 = s_*, s_1 = s^* \end{cases}$

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► **Time-discretization:** Given a number of timesteps $T \in \mathbb{N}$, consider

$$\min_{m_1, \dots, m_T \in \mathbb{R}^d} \frac{T}{2} \sum_{i=0}^{T-1} \|m_{i+1} - m_i\|^2 + \lambda \left(\sum_{i=1}^{T-1} d_S^2(m_i) + \|m_0 - s_*\|^2 + \|m_T - s^*\|^2 \right).$$

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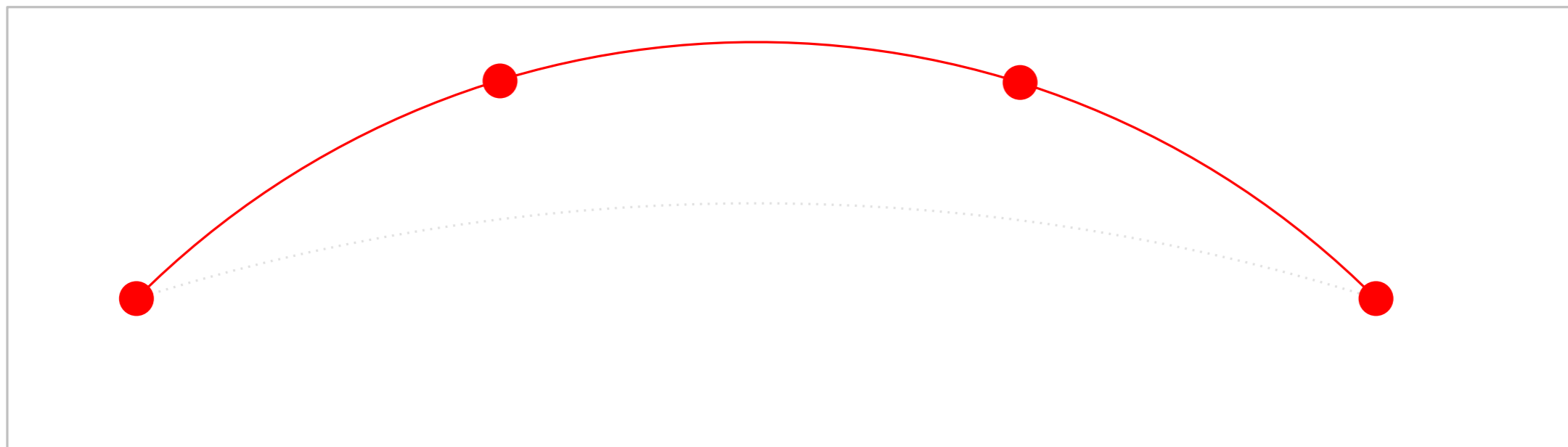
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Imagine now that only a finite sample $S_K \subseteq S$ is known, with $\text{Card}(S_K) = K$.

→ How should $\lambda = \lambda(T, K)$ be chosen ?

Finite-dimensional example

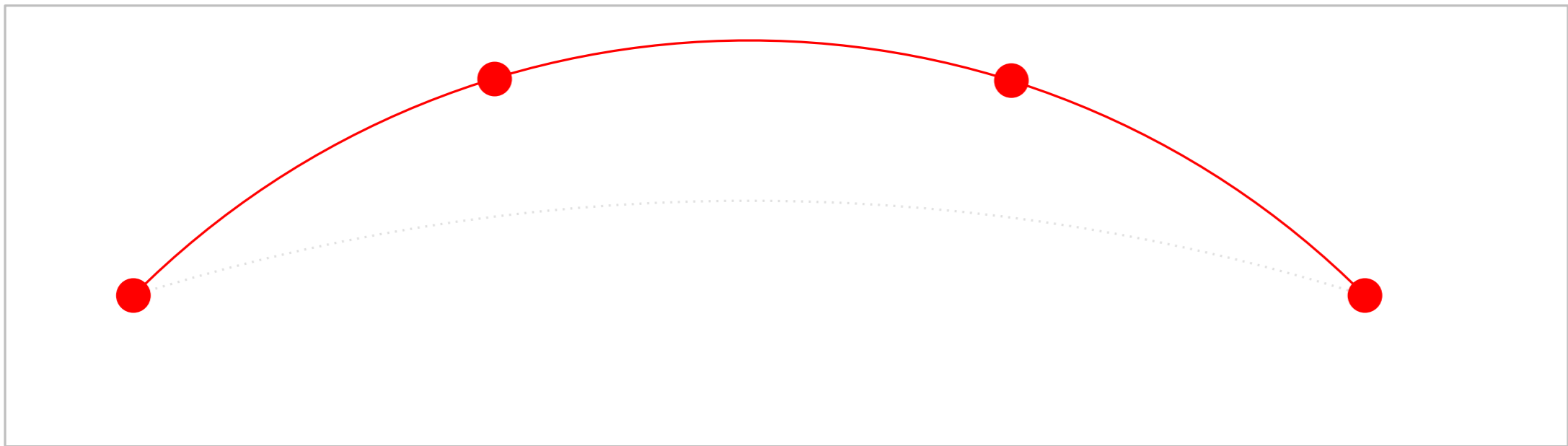
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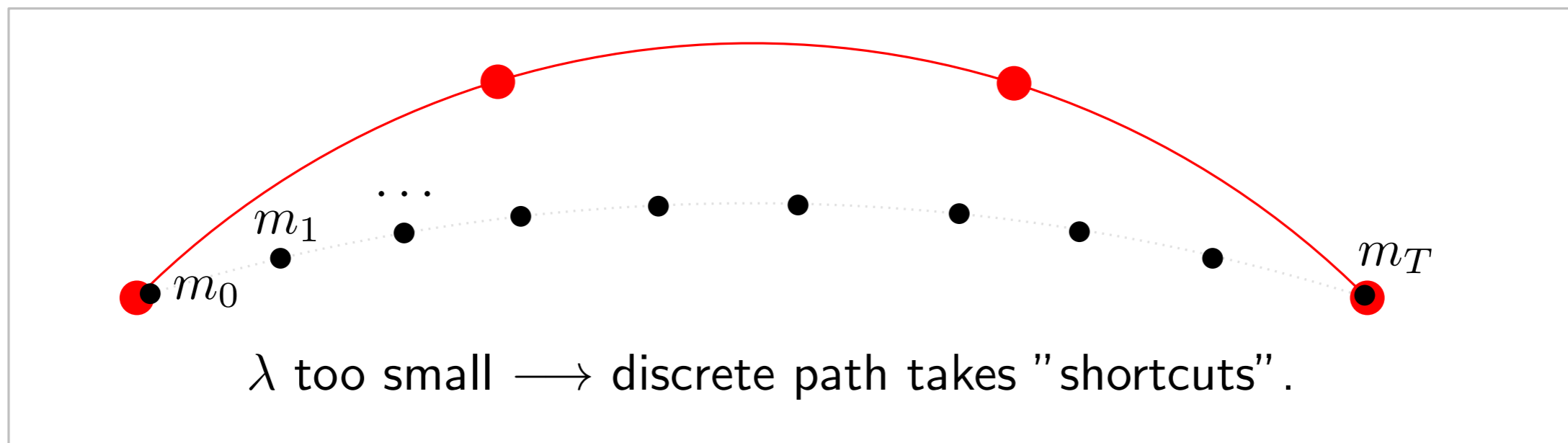
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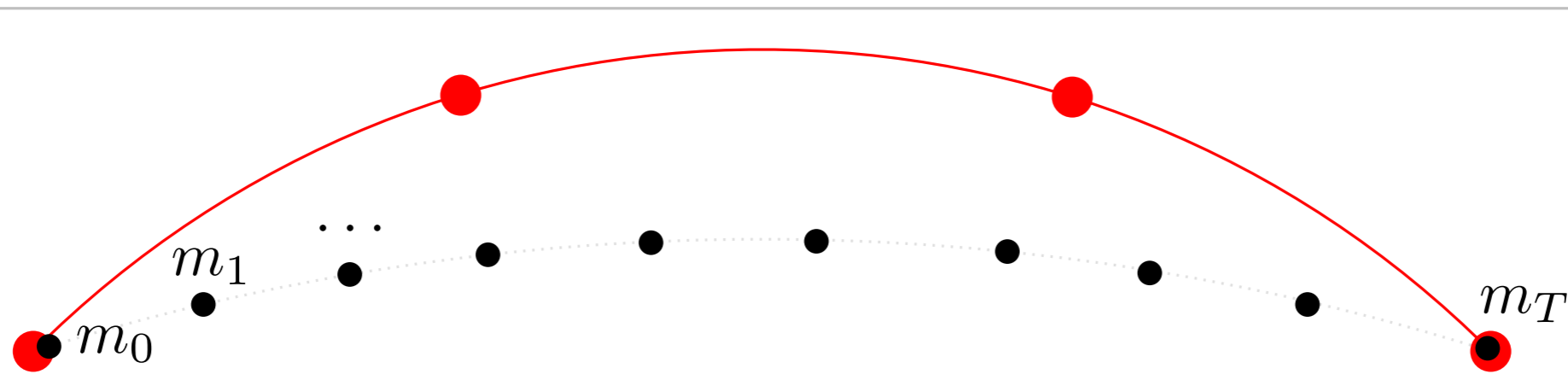
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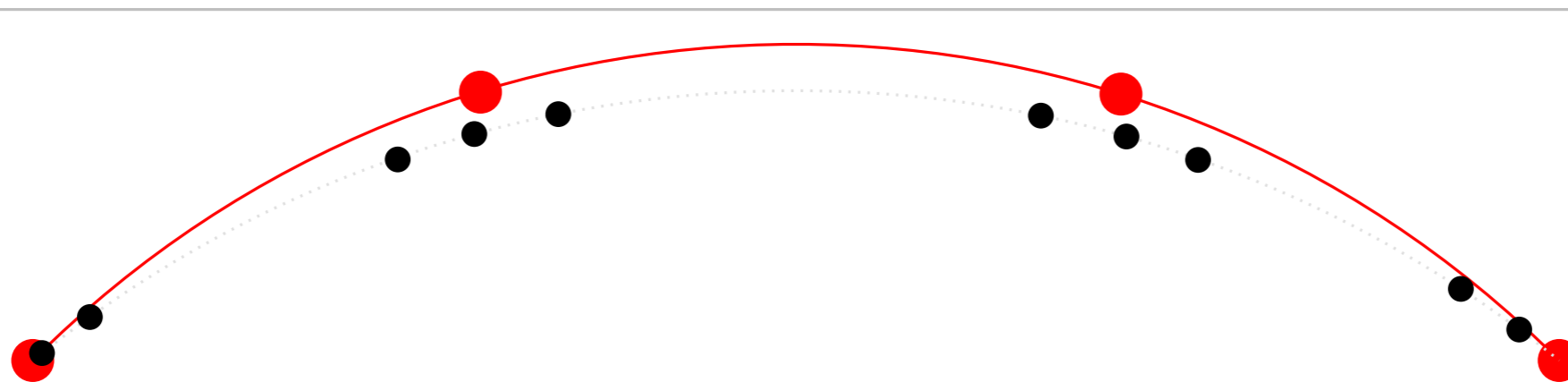
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λ too small \longrightarrow discrete path takes "shortcuts".

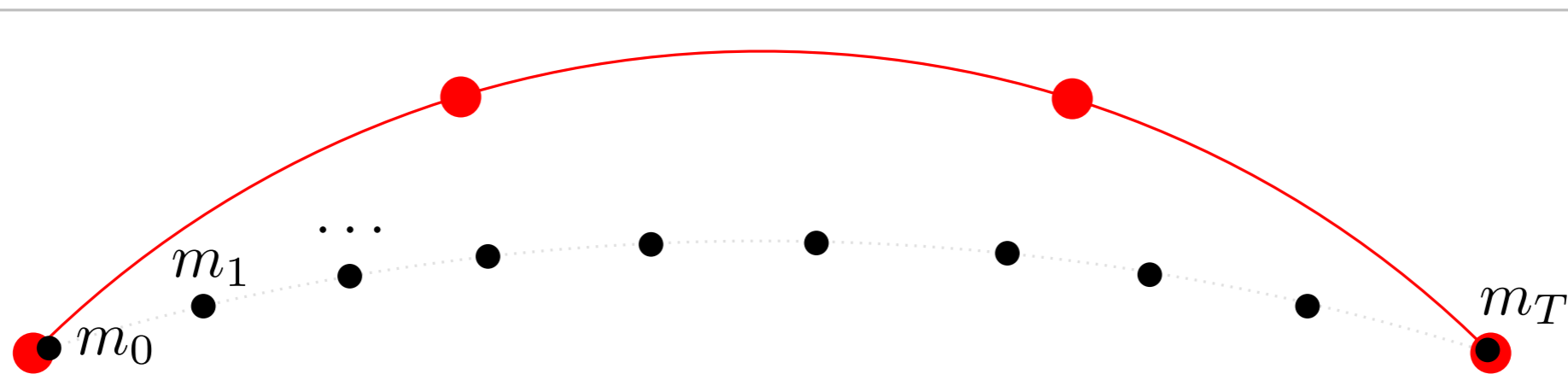


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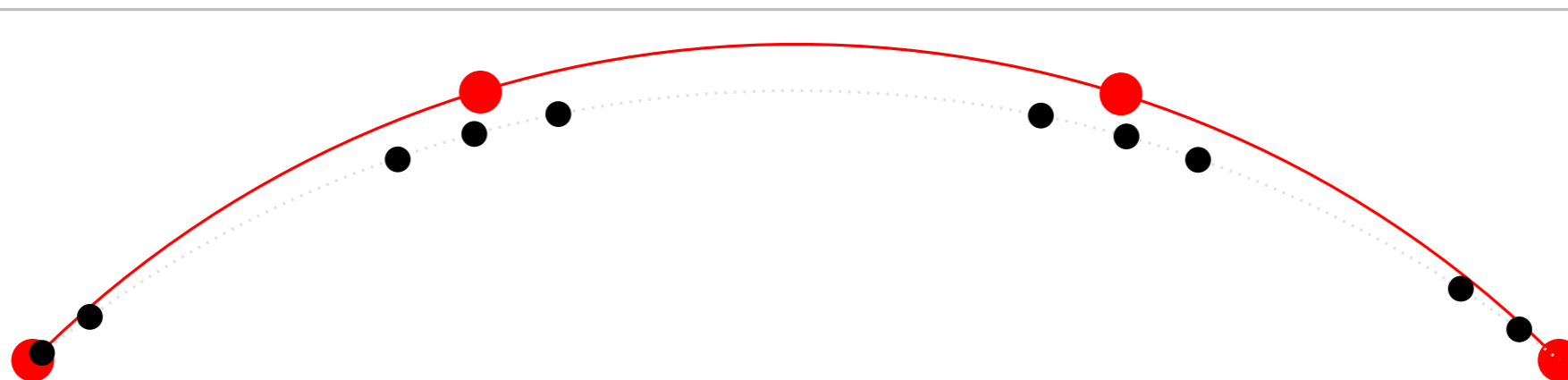
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\longrightarrow combinatorial optimization pb (when $\lambda = +\infty$)

Minimal geodesics in \mathcal{SDiff} and relaxations

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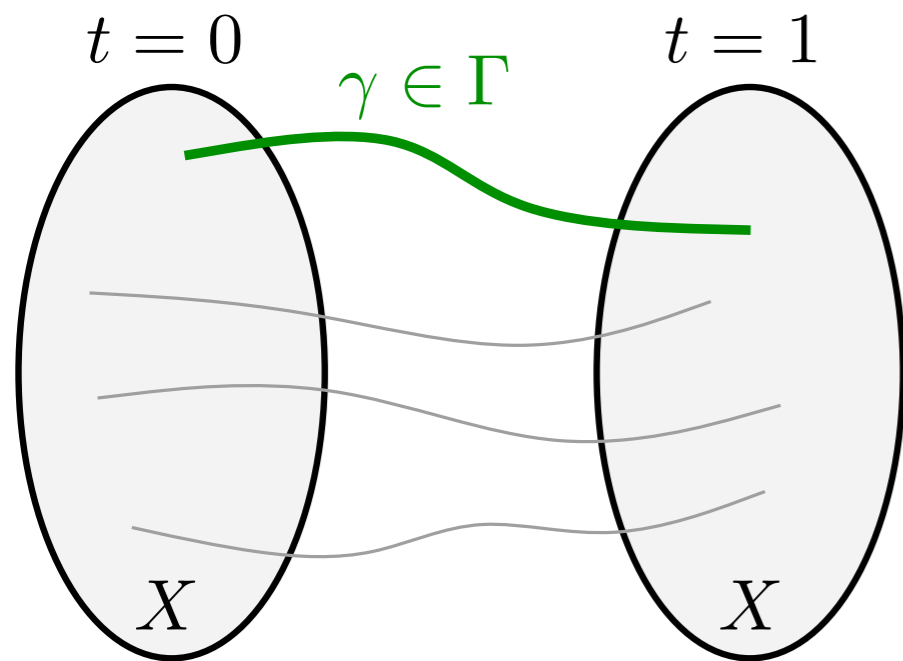
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C. relaxation involving measures over the set Γ of \mathcal{C}^0 paths in X . [Brenier '89]
[Schinerman '94]

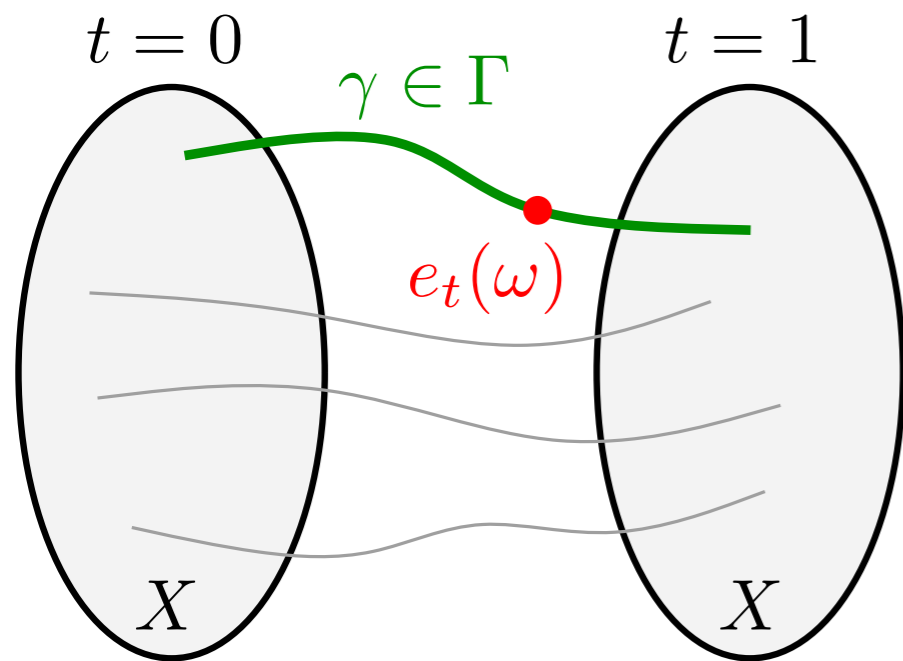
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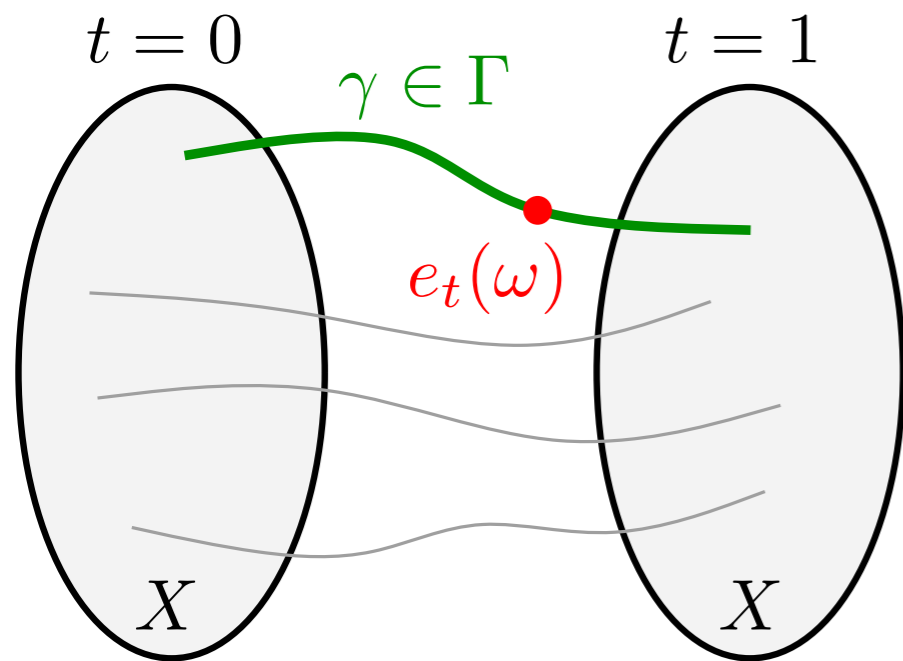
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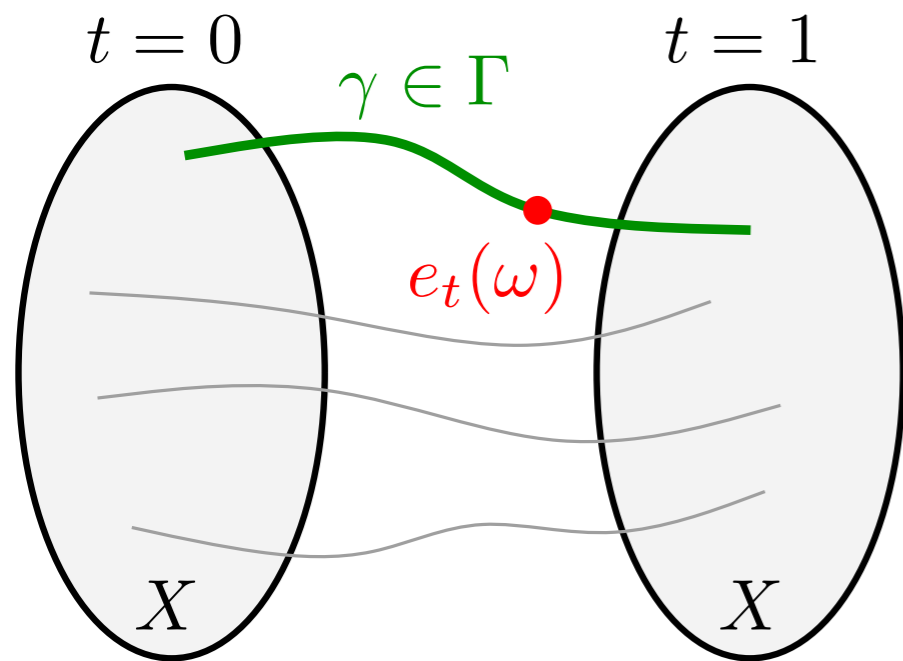


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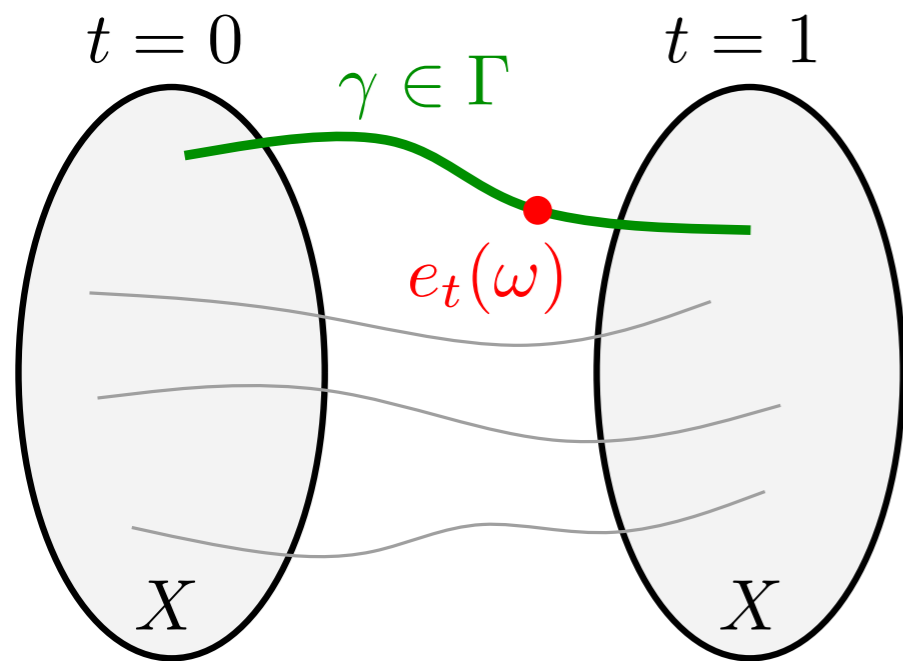
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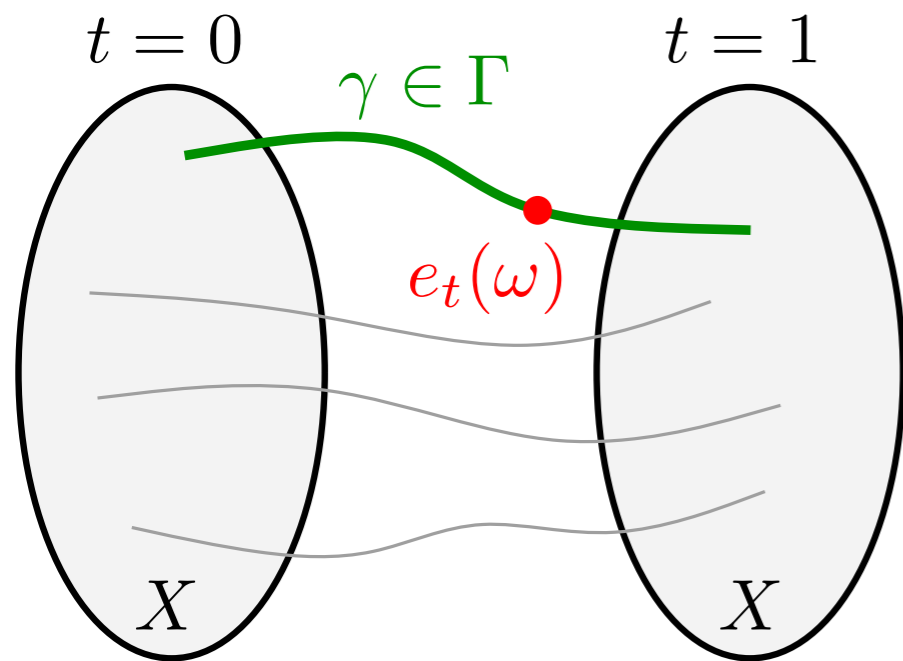
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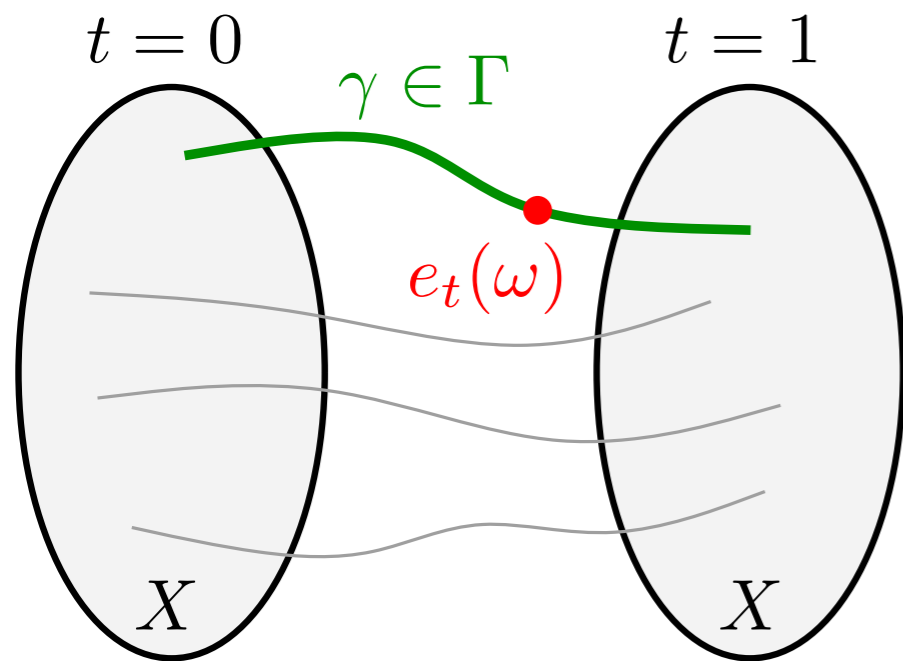
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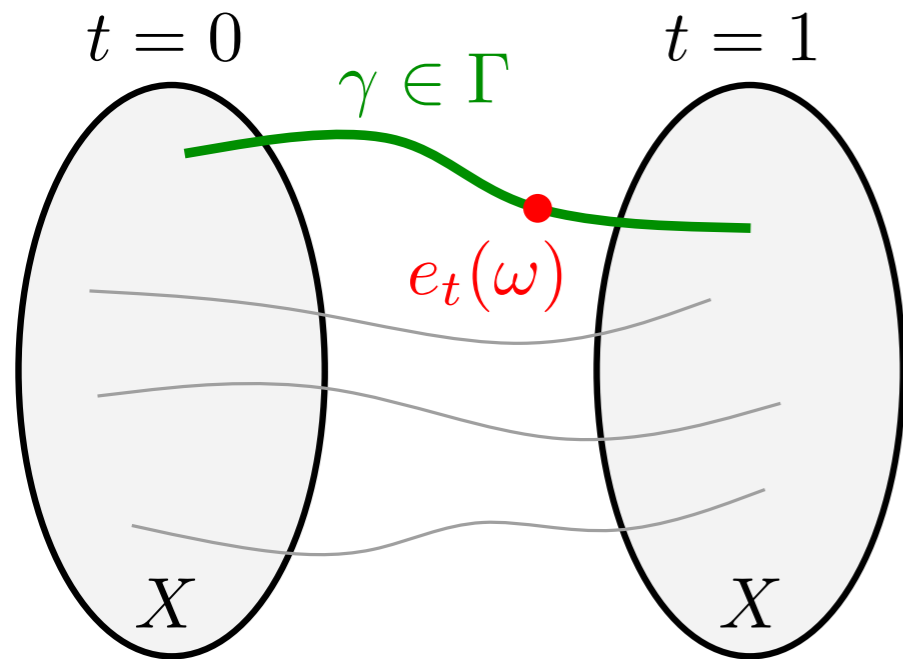
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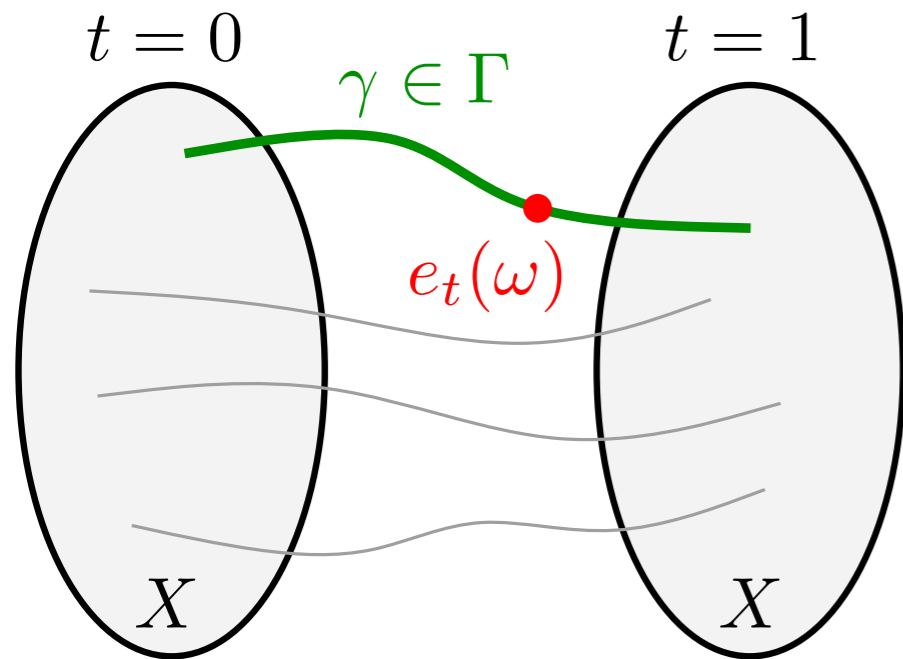
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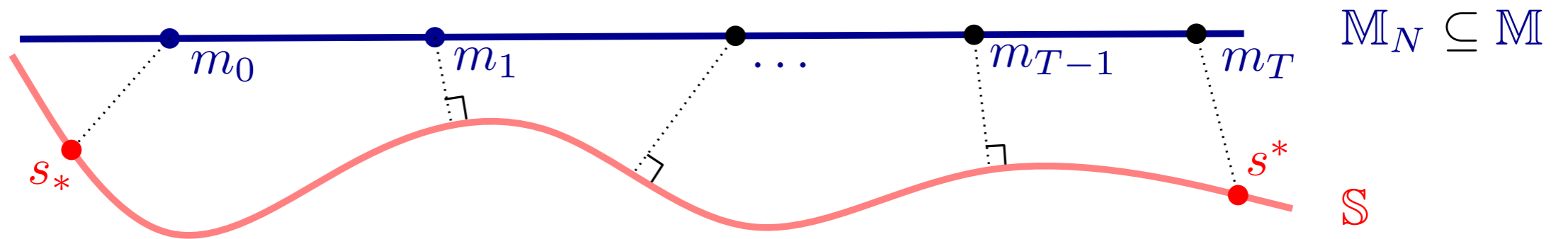
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- ▶ **Numerics:** mostly in 1D using permutations [Brenier '87, Brenier–Roesch '98]

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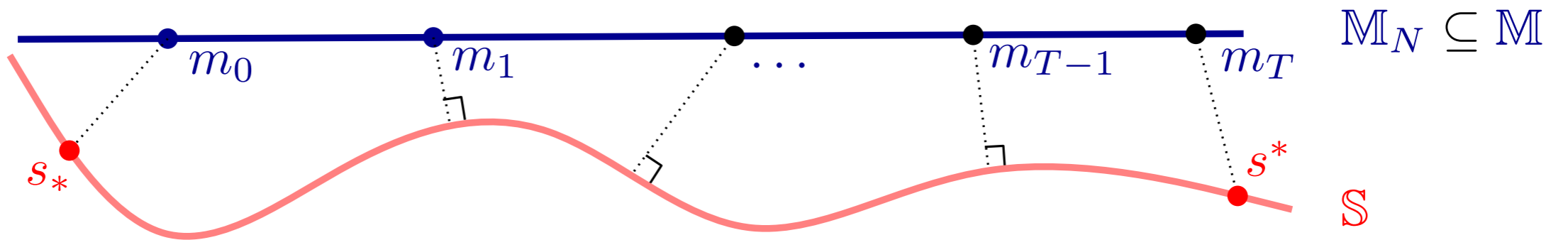
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boundary conditions

incompressibility

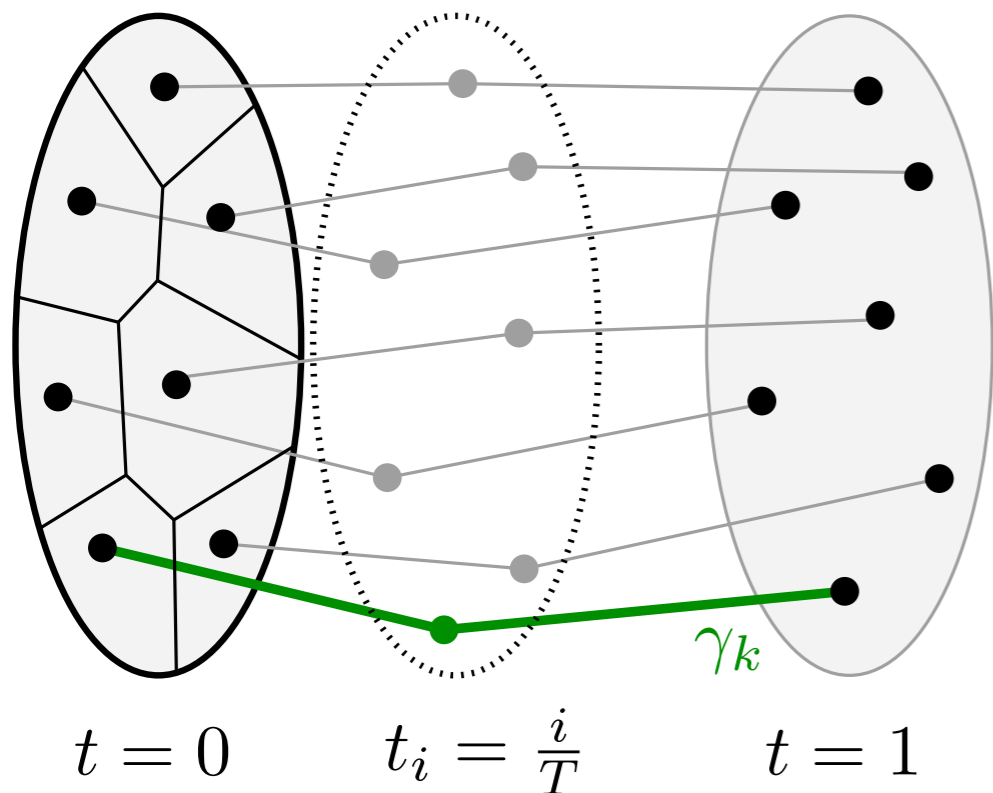
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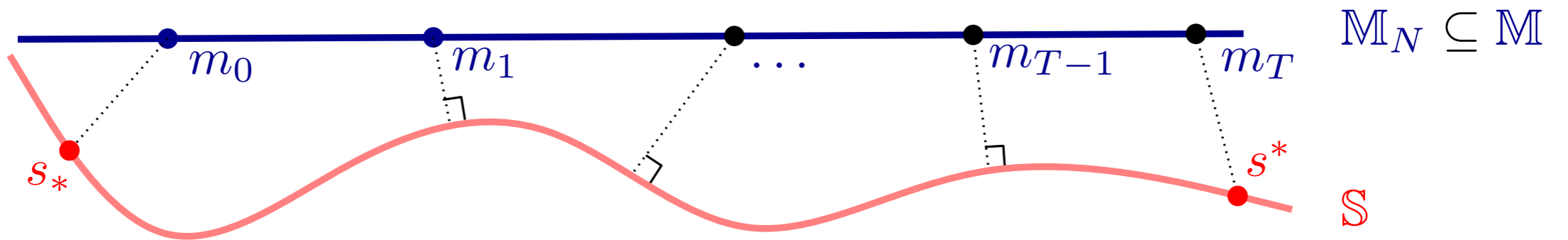
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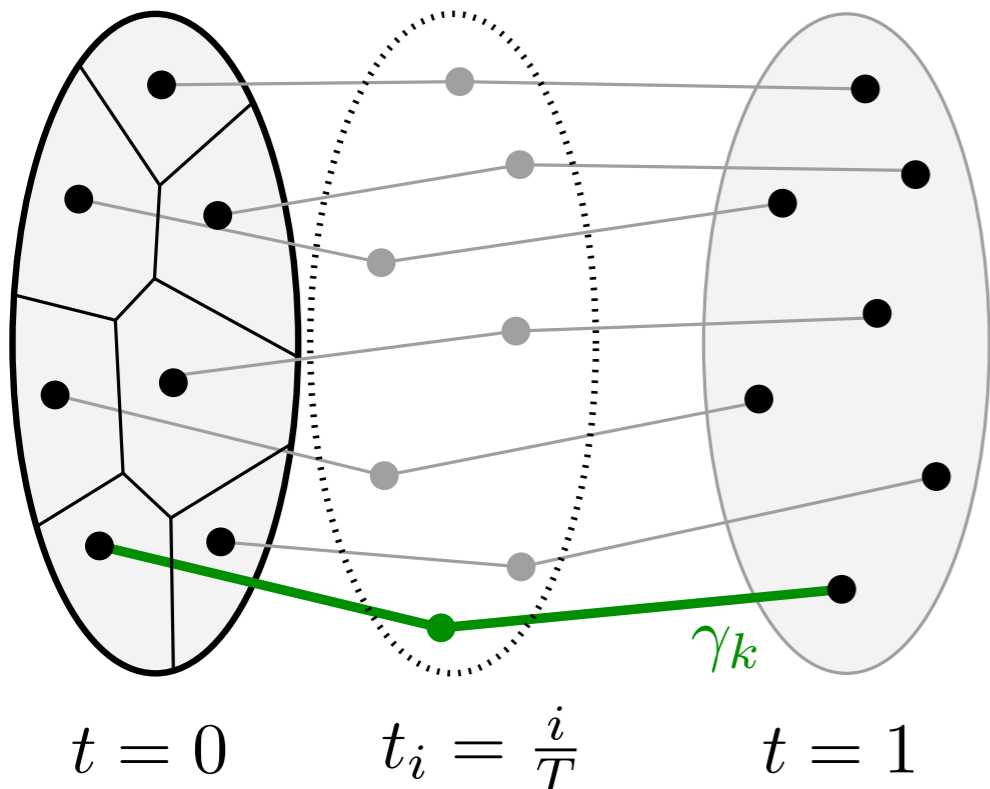
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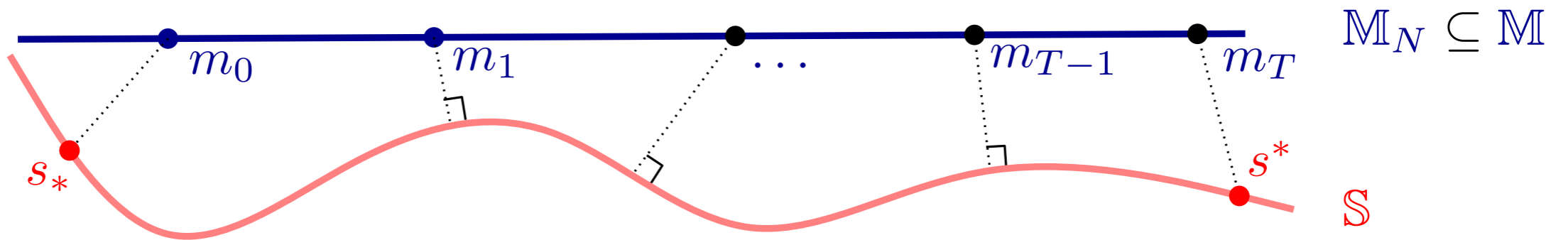
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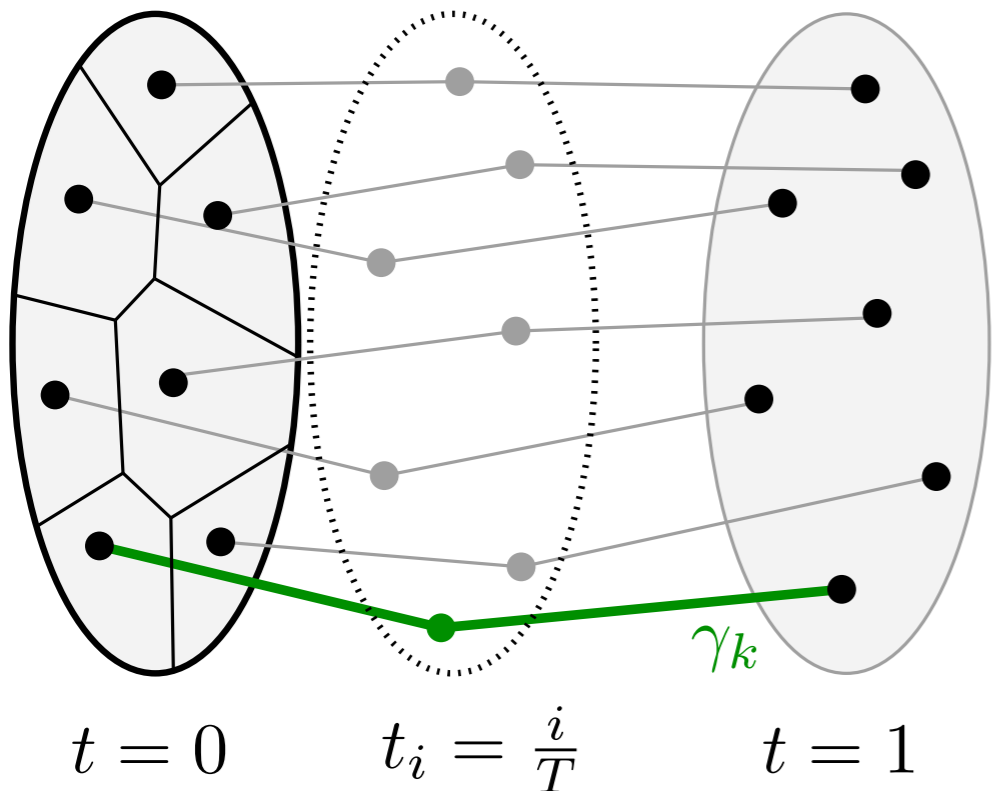
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→ \simeq Common discretization for both relaxations!
 → Choice of penalization parameter?

Convergence theorem

Regular generalized geodesic: a probability measure $\mu \in \text{Prob}(\Gamma)$ s.t.

(Regularity) $\exists p$ with Lipschitz gradient s.t. $\forall \gamma \in \text{spt}(\mu), \ddot{\gamma} = -\nabla p \circ \gamma,$

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Then, up to subsequences, $\mu_{m_N} \in \text{Prob}(\Gamma)$ converges weakly to a minimizing generalized geodesic between s_* and s^* .

[Mirebeau–M., 2015]

► Main step: $\limsup_N \mathcal{E}_{N, T_N, \lambda_N}(m_N) \leq \mathcal{E}(\mu^{\text{opt}}).$

more precisely, we need $\min_{m \in \mathbb{M}_N^T} \mathcal{E}_{N, T, \lambda}(m) \leq \mathcal{E}(\mu^{\text{opt}}) + \mathcal{O}(Th_N^2 \lambda)$

for $h_N := N^{-\frac{1}{D}}$, with $D \in \mathbb{N}$ to be determined.

► It turns out that one can take $D := \dim(\text{spt}(\mu^{\text{opt}}))$

→ For a classical solution $s : [0, 1] \rightarrow \mathbb{S}$, $\dim(\text{spt}(\mu^{\text{opt}})) = d. \quad (\lambda_N = N^d)$

→ For a regular generalized solution, $\dim(\text{spt}(\mu^{\text{opt}})) \leq 2d. \quad (\lambda_N = N^{2d})$

Energy estimate for classical solutions

Proposition: Assume that the minimizing geodesic s between s_* and s^* is classical and that $s \in L^\infty([0, 1], H^1(X))$. Then, with $h_N = N^{-1/d}$,

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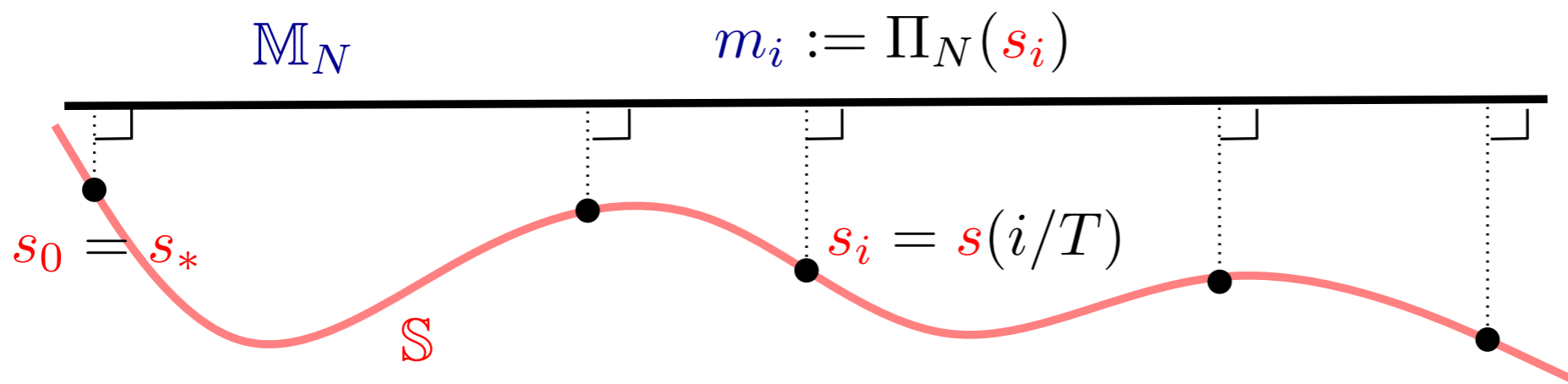
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Energy estimate for classical solutions

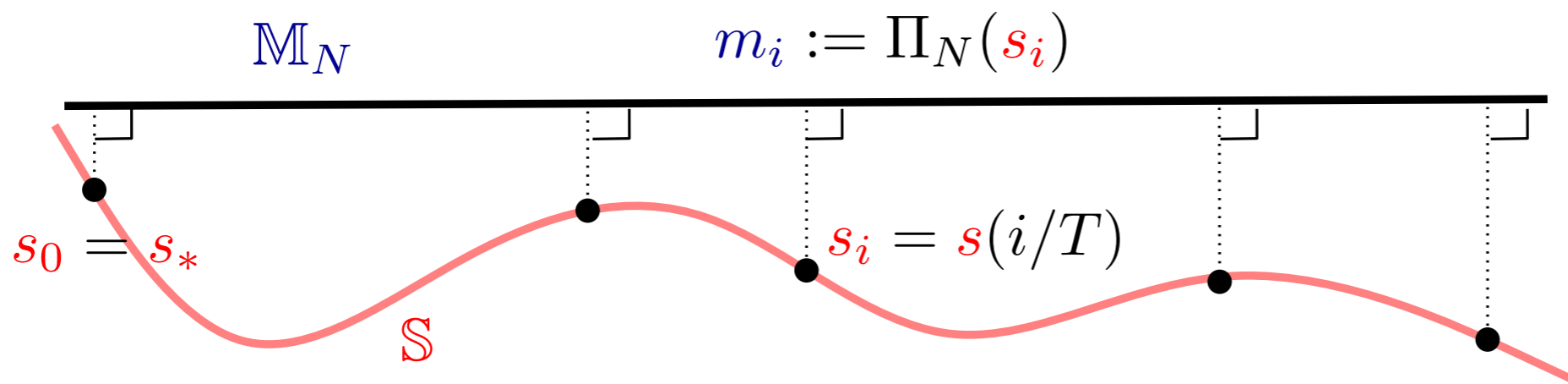
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Then, $\mathcal{E}_{N,T,\lambda}(m)$ is upper bounded using the Poincaré-Wirtinger inequality.



Energy estimate for generalized solutions

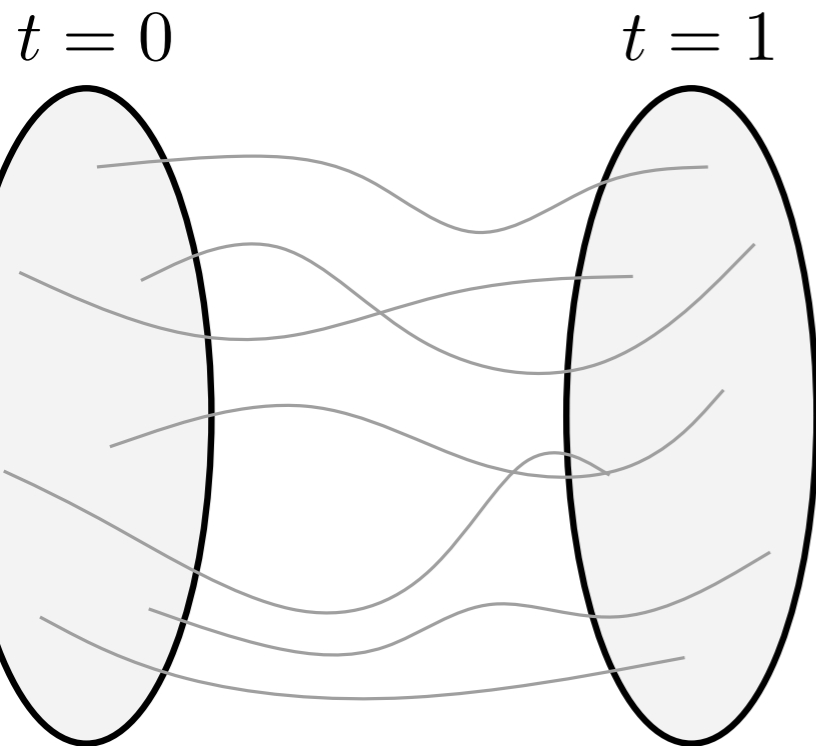
Prop: Assume that the generalized minimizing geodesic in Π is associated to a pressure $p : [0, 1] \times \Omega \rightarrow \mathbb{R}$ with Lipschitz gradient. Then, with $h_N = N^{-1/2d}$,

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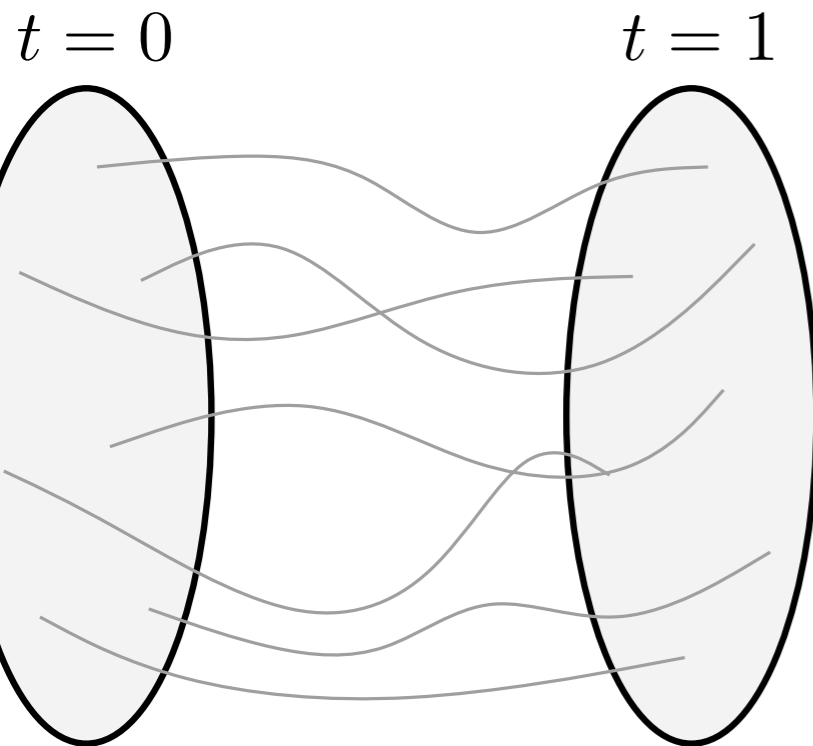


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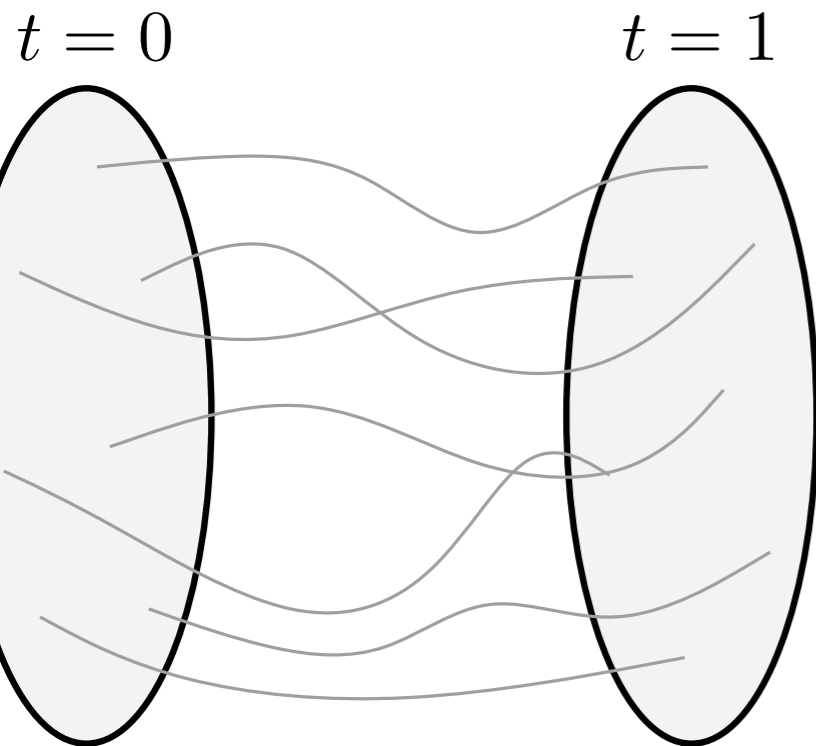
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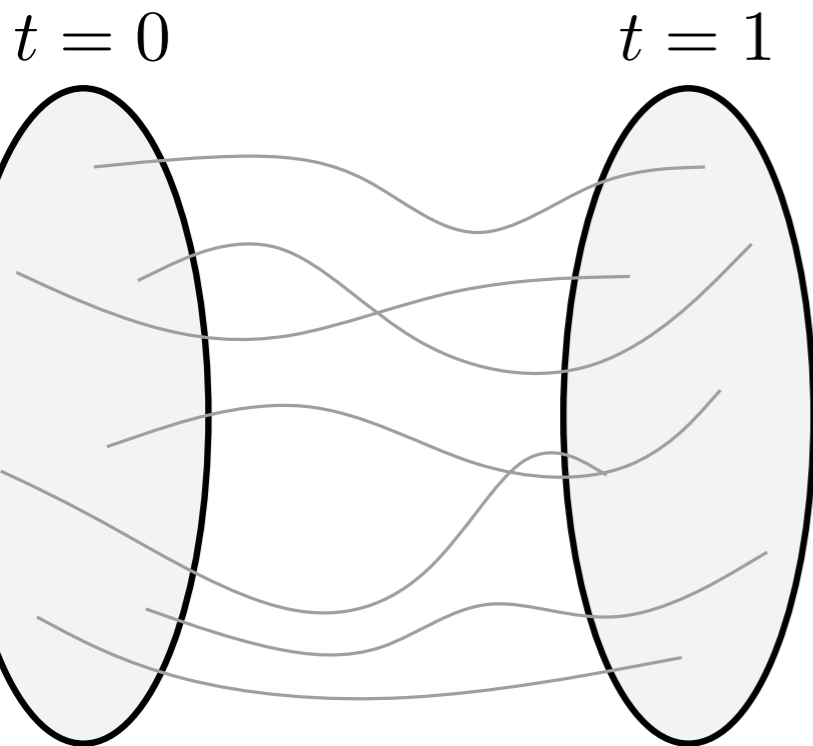
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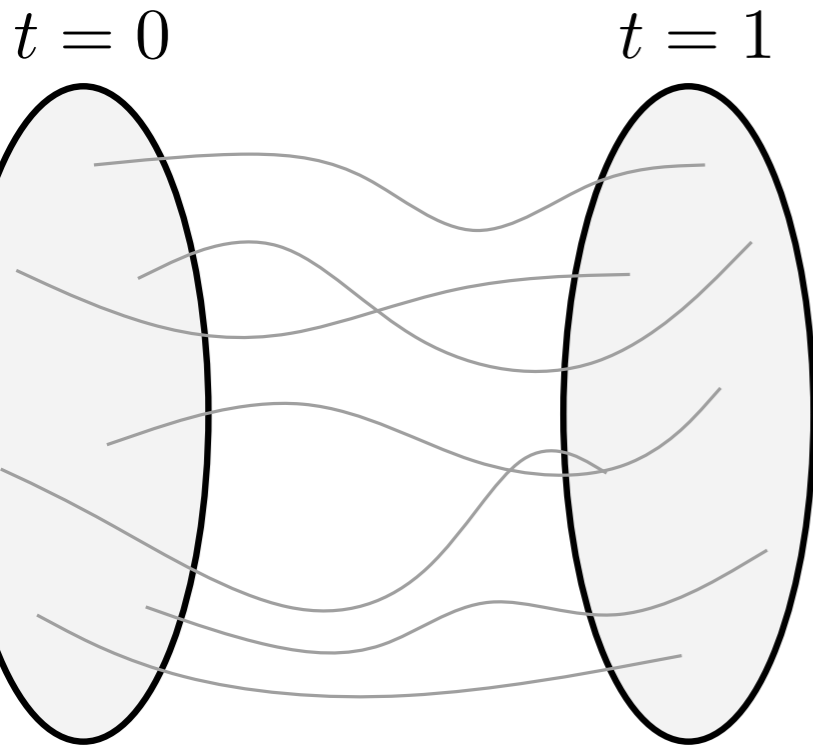
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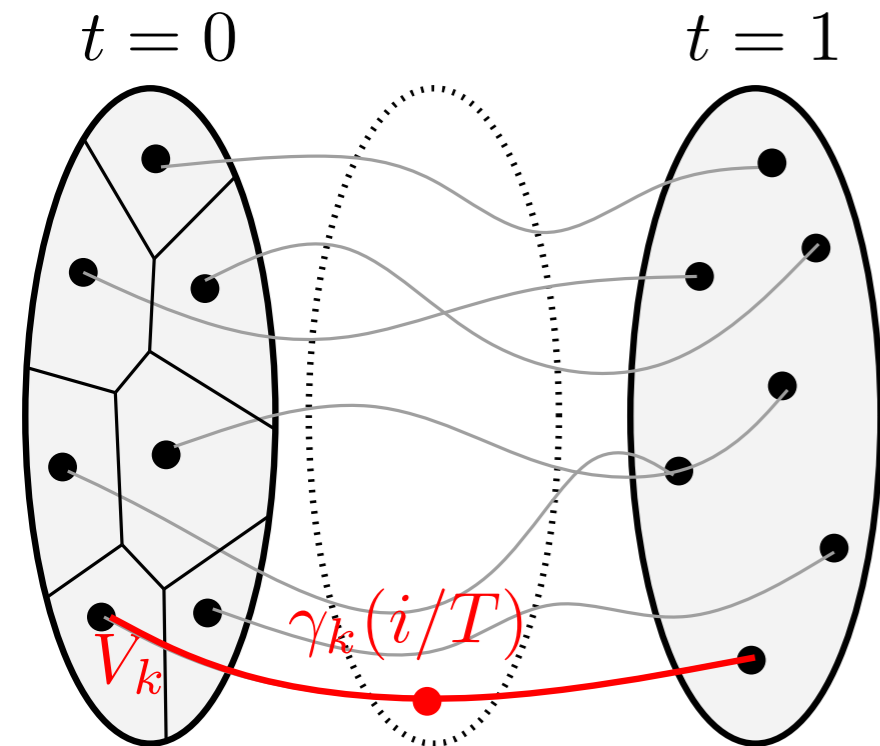
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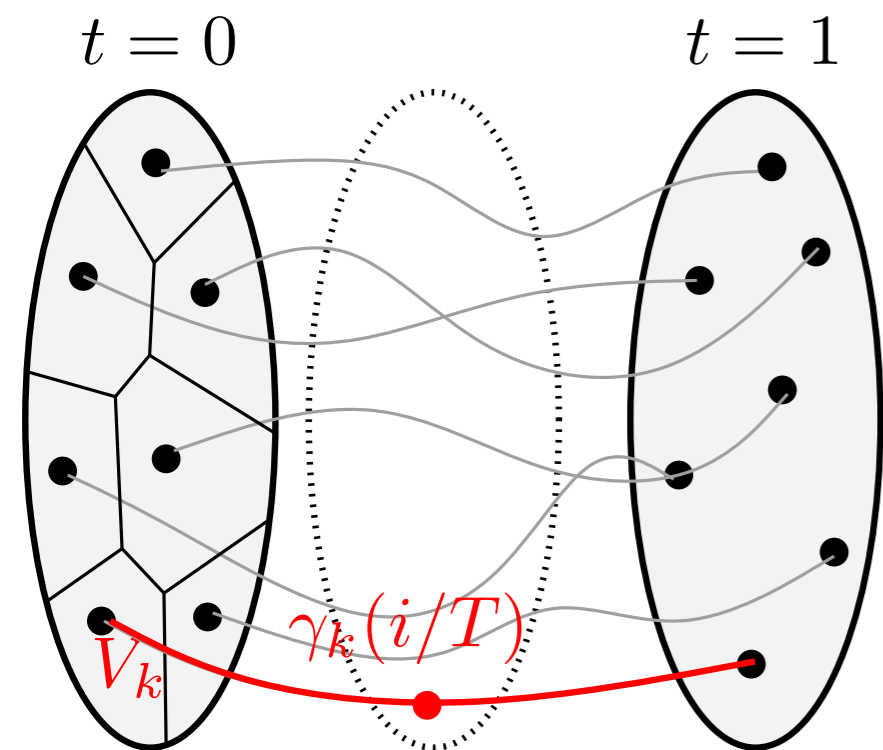
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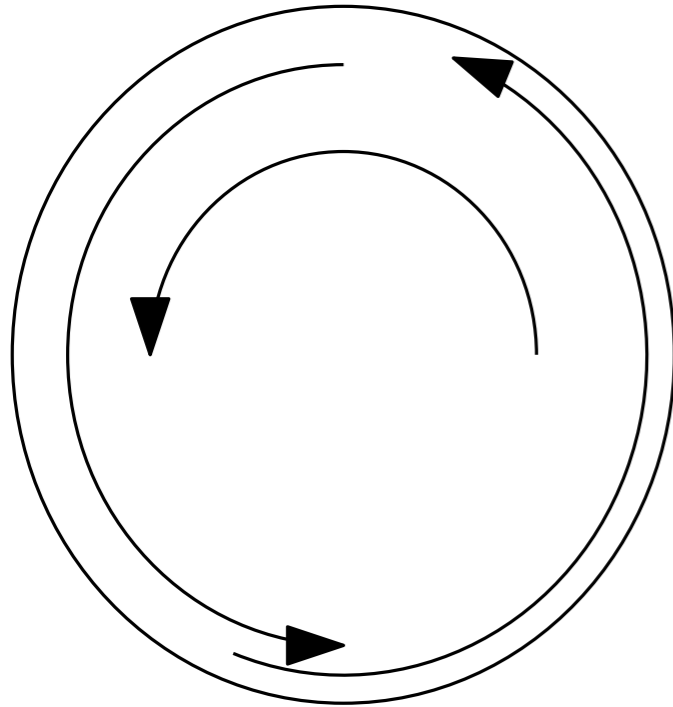
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E. Upper bound $\mathcal{E}_{N,T,\lambda}(m)$ using the quantization estimate.

Numerical result: Inversion of the Disk

$$X = B(0, 1) \subseteq \mathbb{R}^2$$

$$(s_*, s^*) = (\text{id}, -\text{id})$$

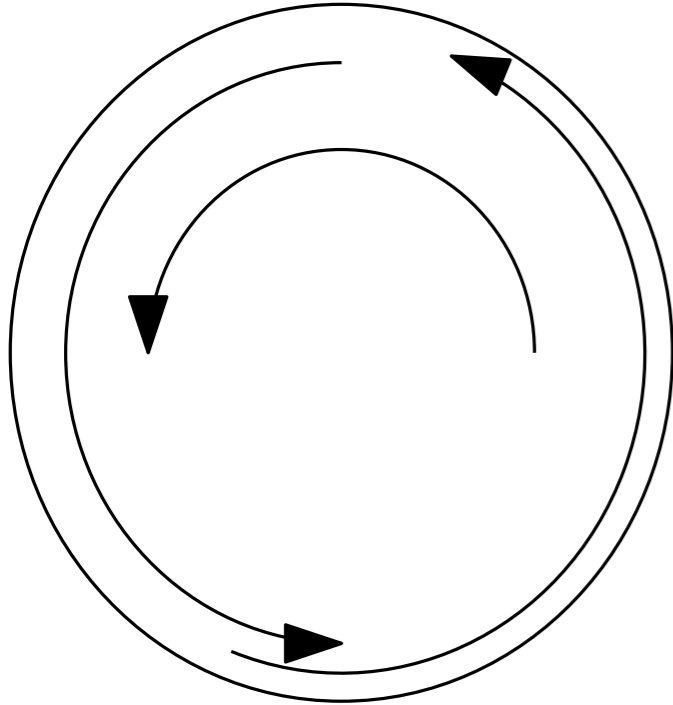


Classical solutions: clockwise/counterclockwise rotations μ_{\pm}

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Examples of generalized solutions:

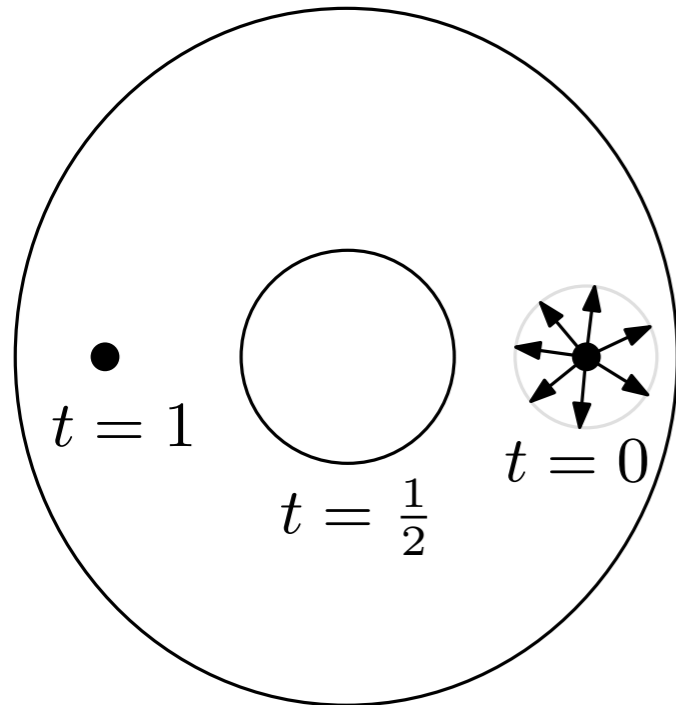
linear combination $\mu_{\frac{1}{2}}$ of μ_{\pm} constructed from rotations

NB: $\dim(\text{spt}(\mu_{\frac{1}{2}})) = 2$

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Brenier's generalized solution: $\mu \in \text{Prob}(\Gamma)$:

$$\text{spt}(\mu) = \left\{ t \mapsto x \cos(\pi t) + v \sin(\pi t) \in \Gamma; \right. \\ \left. (x, v) \in X \times \mathbb{R}^2, \|v\|^2 = 1 - \|x\|^2 \right\}$$

→ non-deterministic solution, $\dim(\text{spt}(\mu)) = 3$

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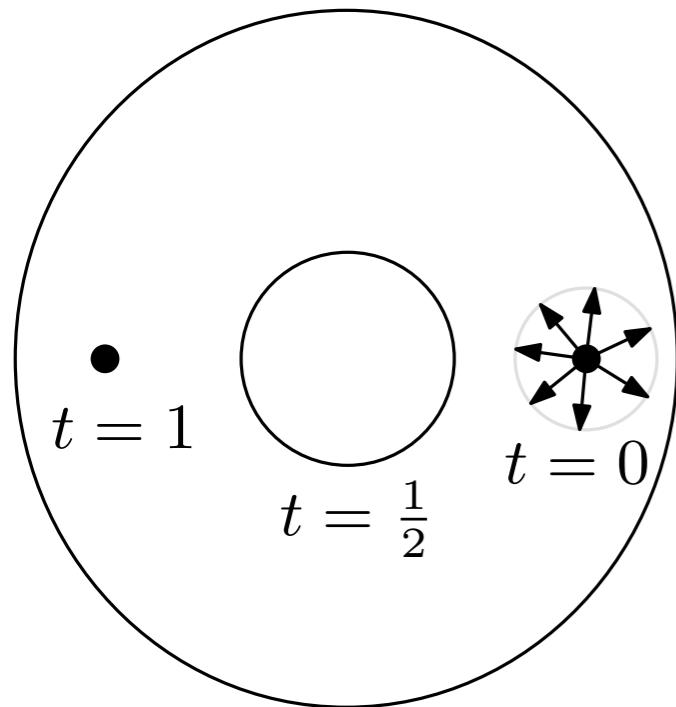
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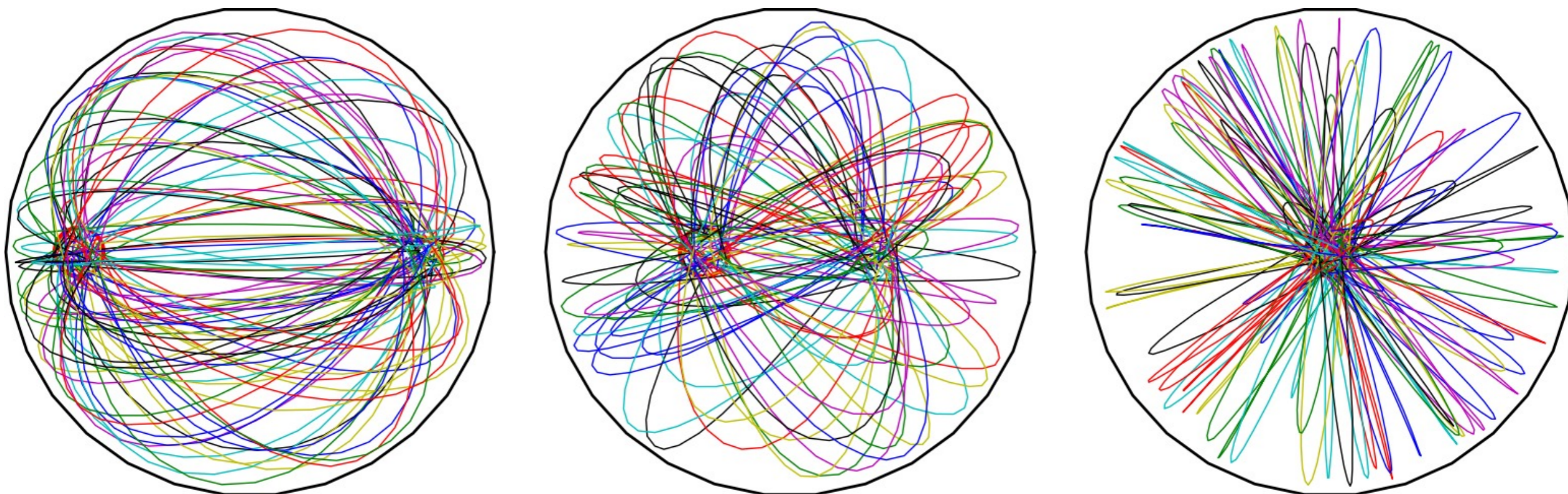
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Computed trajectories for $N = 10^5$, $T = 17$

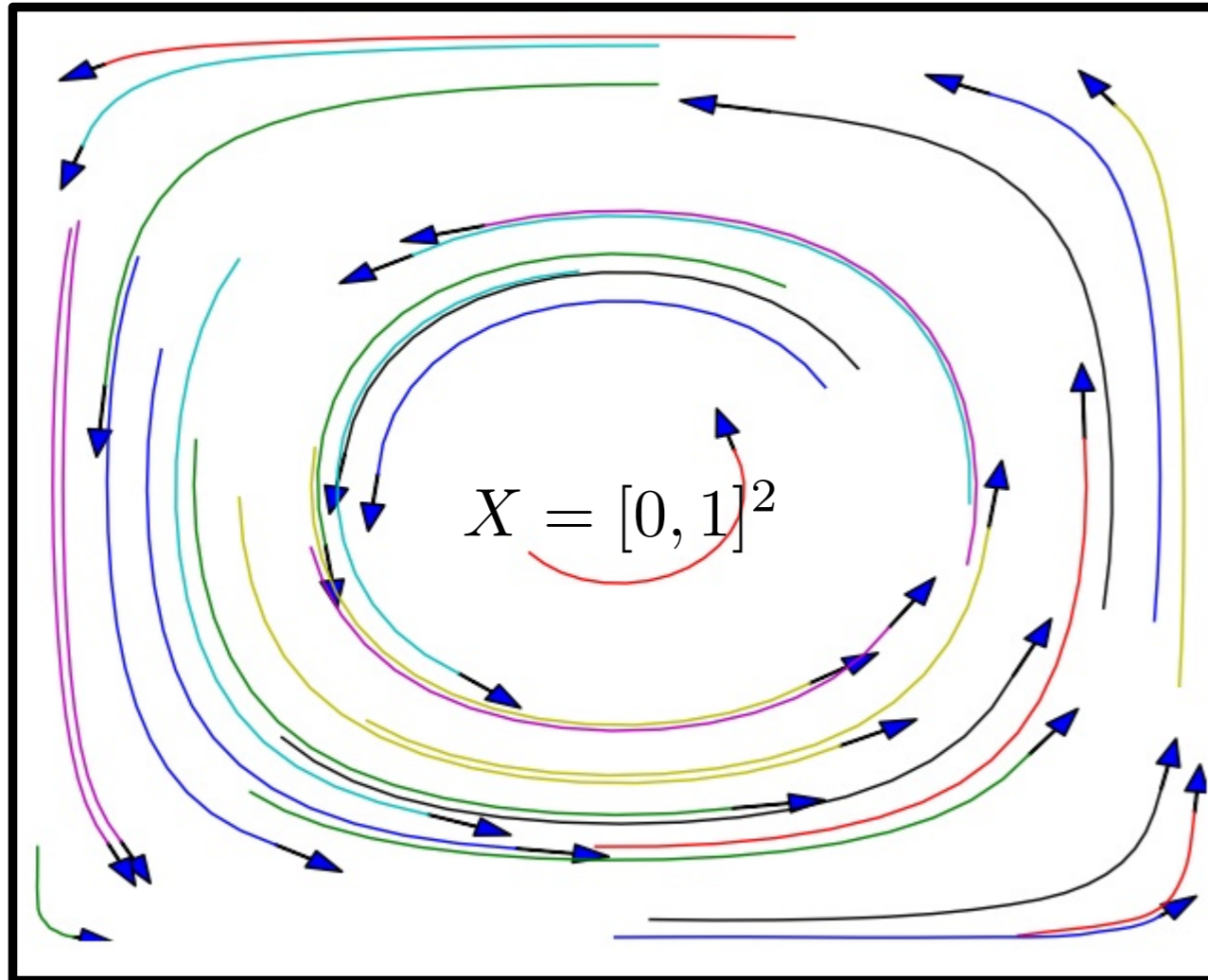


Numerical result: Beltrami Flow in Square

Stationary flow on $[0, 1]^2$: speed: $u(\mathbf{x}) = (\cos(\pi x_1) \sin(\pi x_2), \sin(\pi x_1) \cos(\pi x_2))$

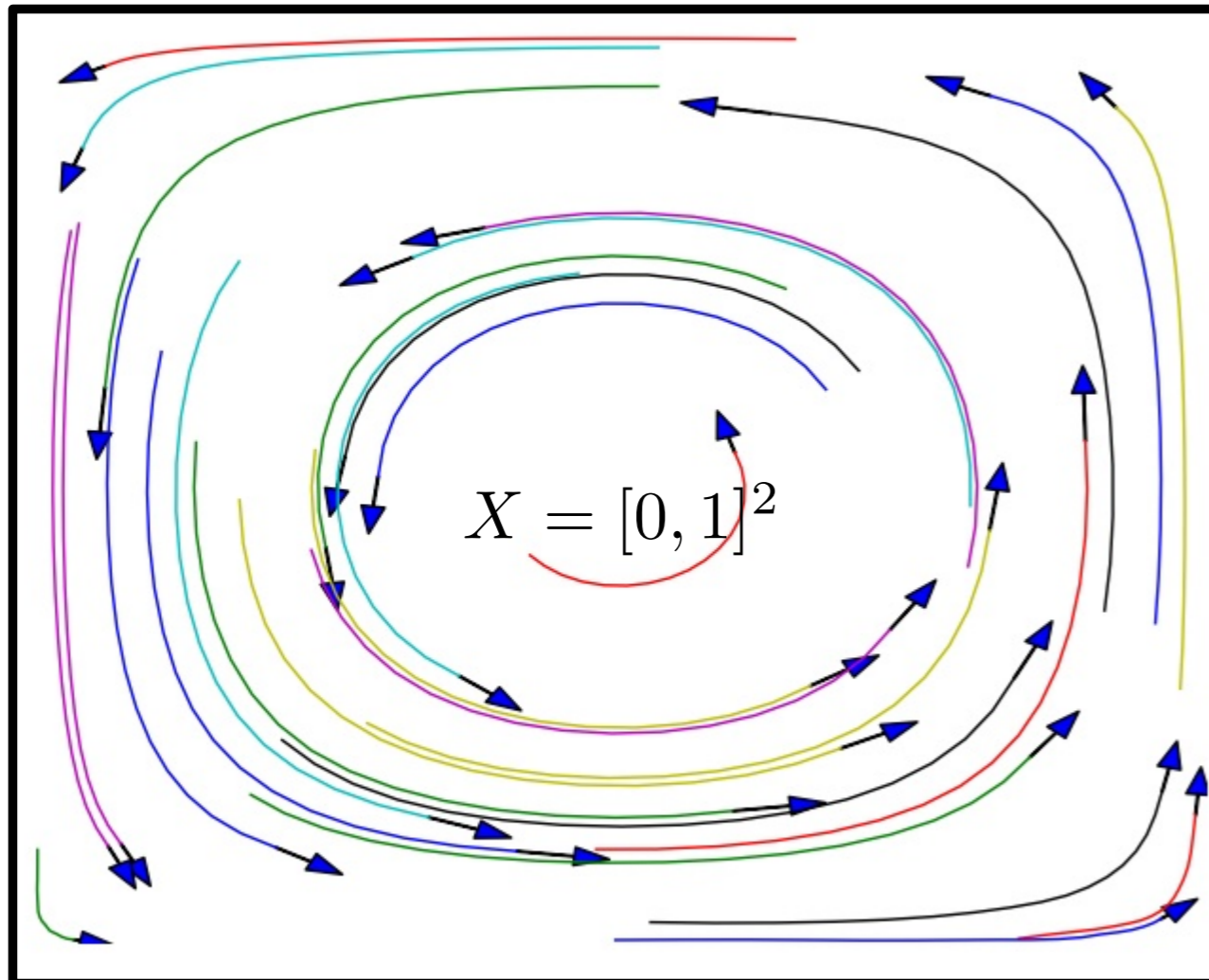
[Brenier–Roesch]

pressure: $p(\mathbf{x}) = \frac{1}{4}(\sin^2(\pi x_1) + \sin^2(\pi x_2))$



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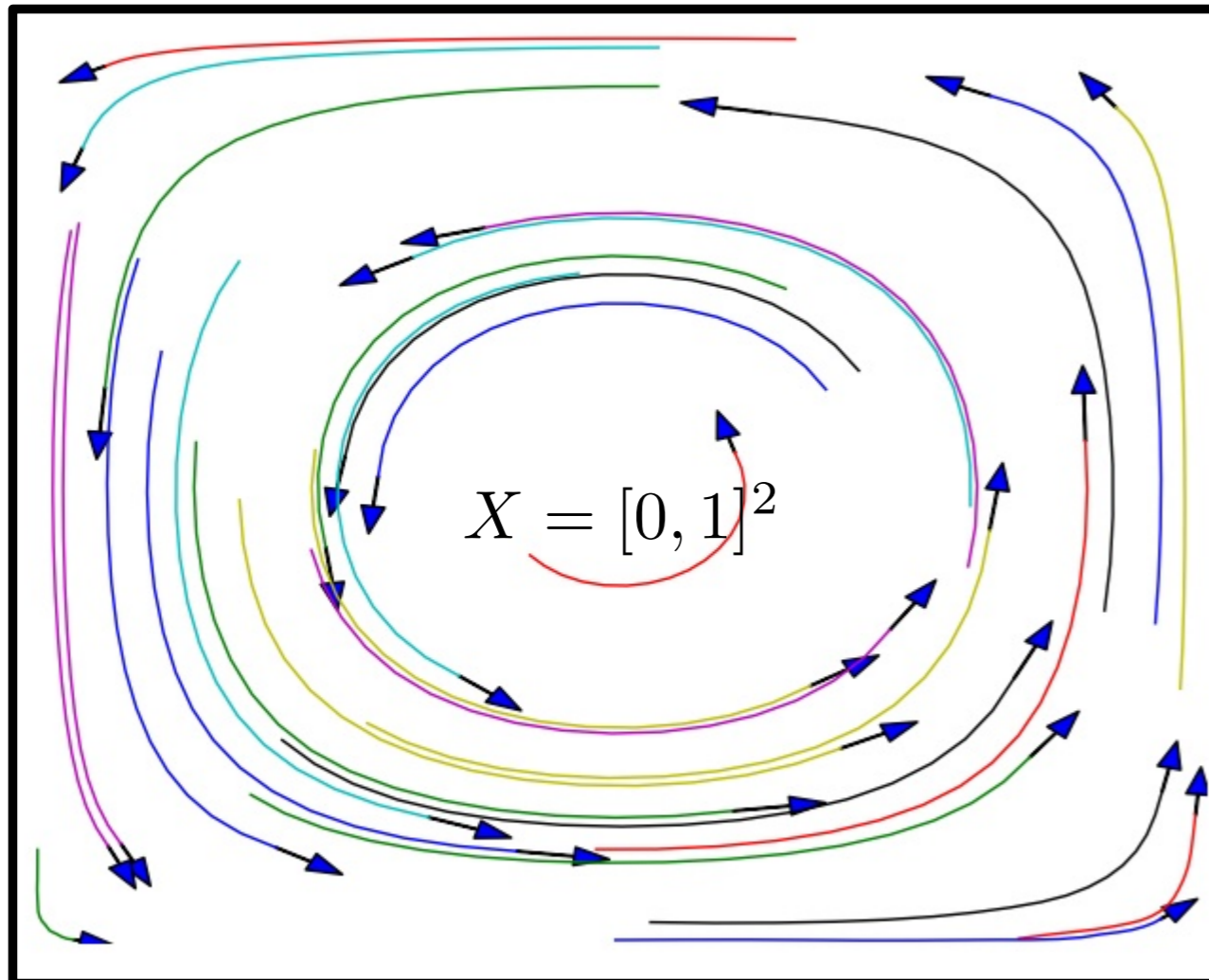
Exact Lagrangian solution:

$$s_0^e = \text{id} \quad \dot{s}_t^e = u \circ s_t$$

NB: s^e is minimizing on $[0, 1]$

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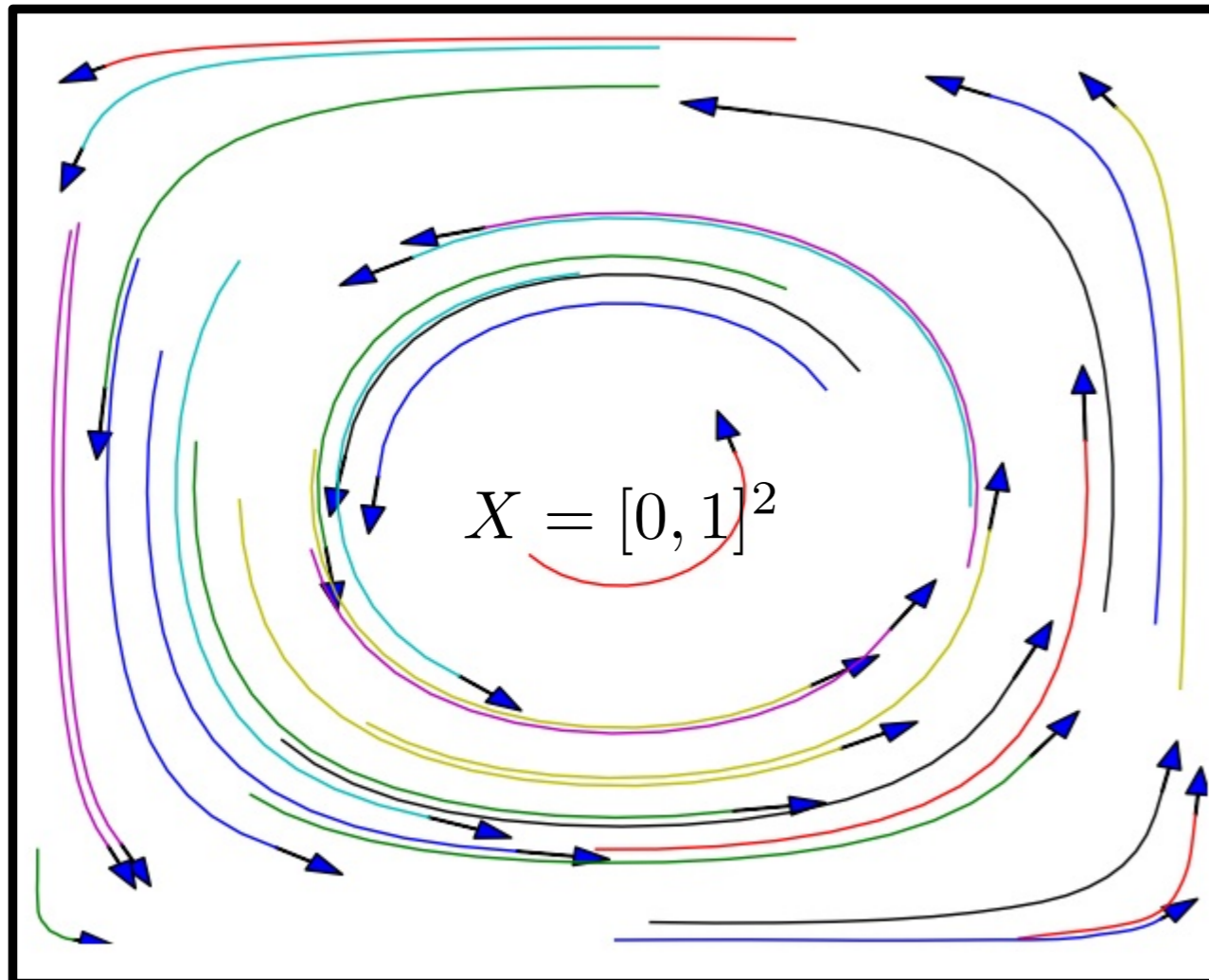
Reconstruction problem:

$$\min \mathcal{E}_{N,T,\lambda}$$

$$s_* = s_0^e, \quad s^* = s_{t_{\max}}^e$$

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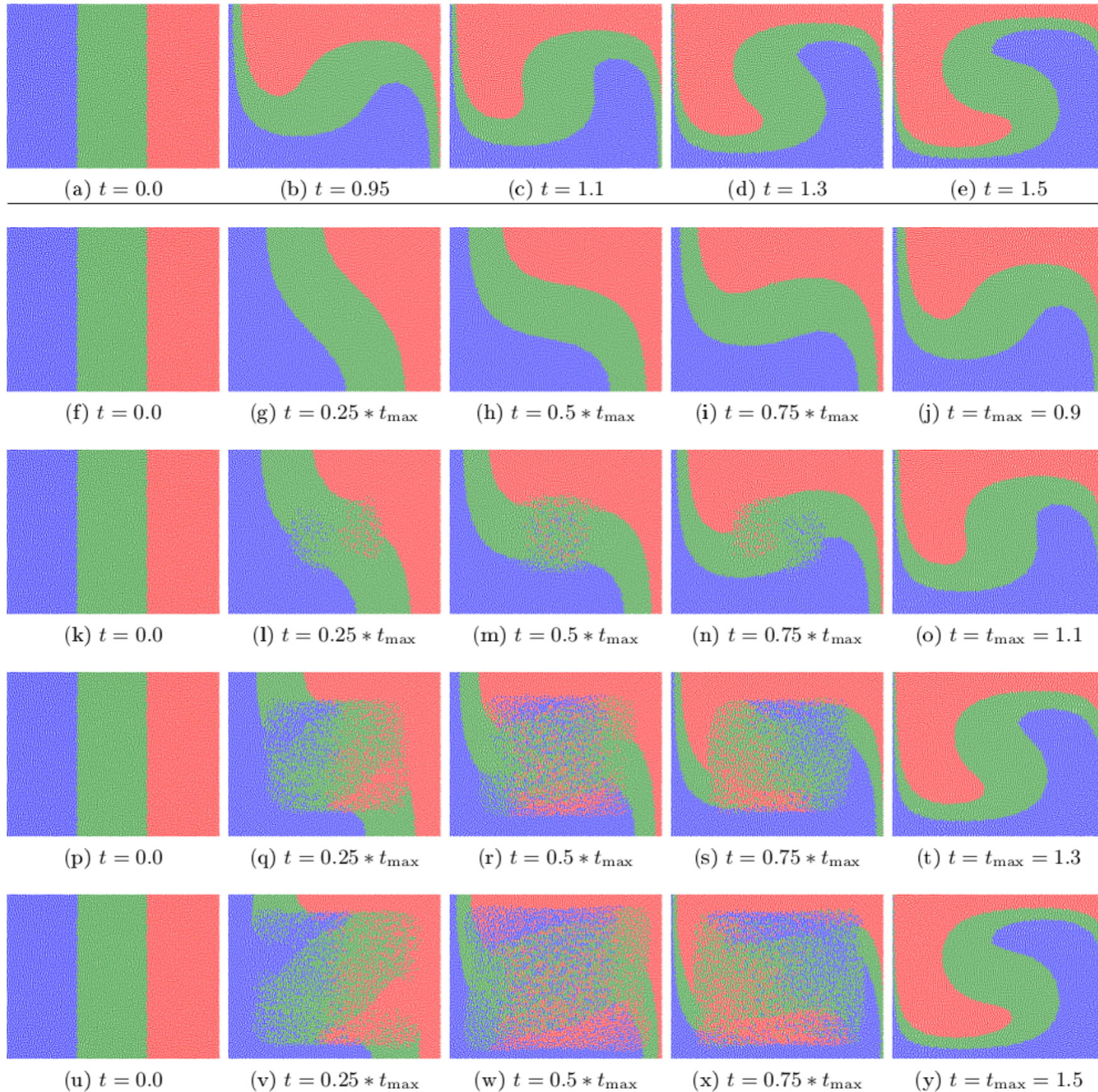
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Parameters:

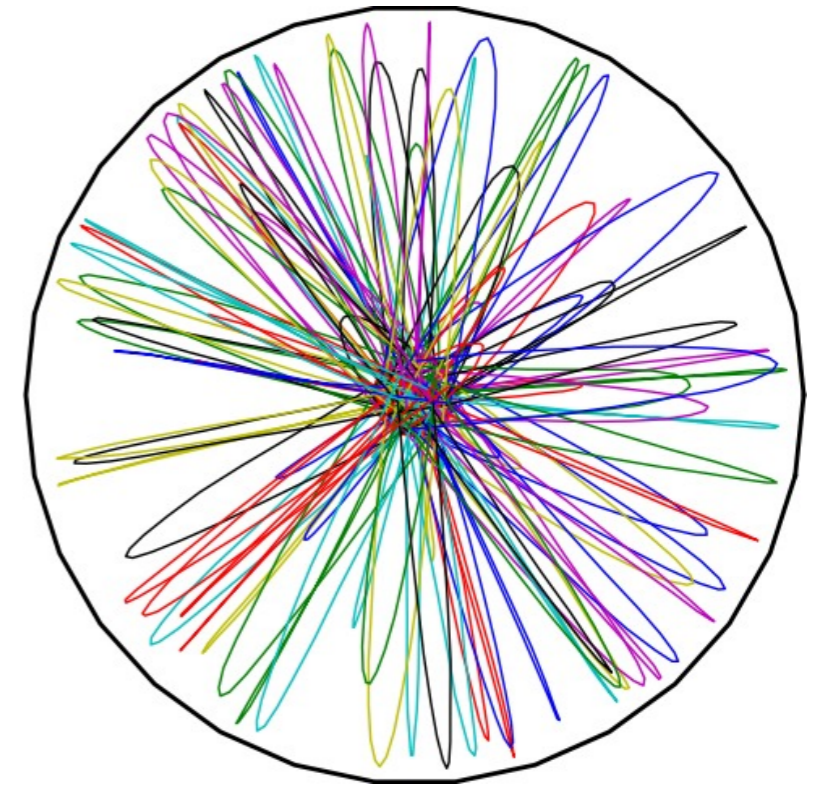
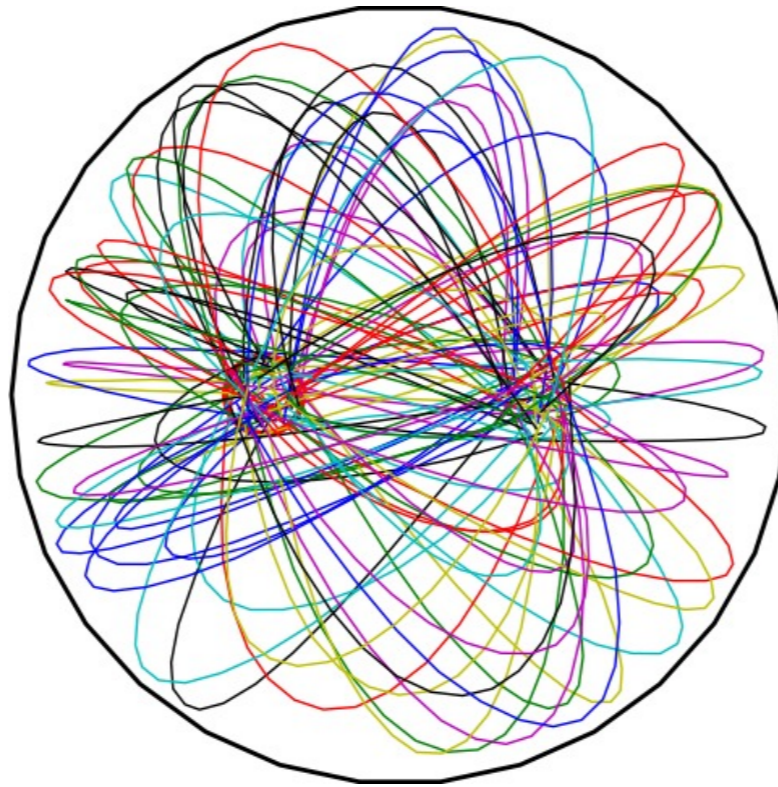
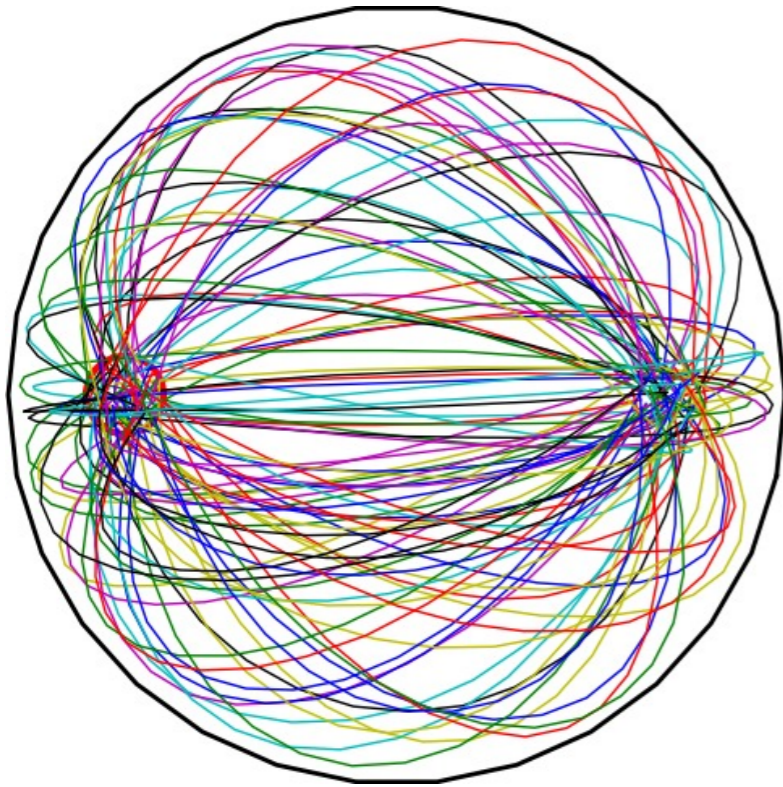
$$t_{\max} \in \{0.9, 1.1, 1.3, 1.5\}$$

Numerical result: Beltrami Flow in Square

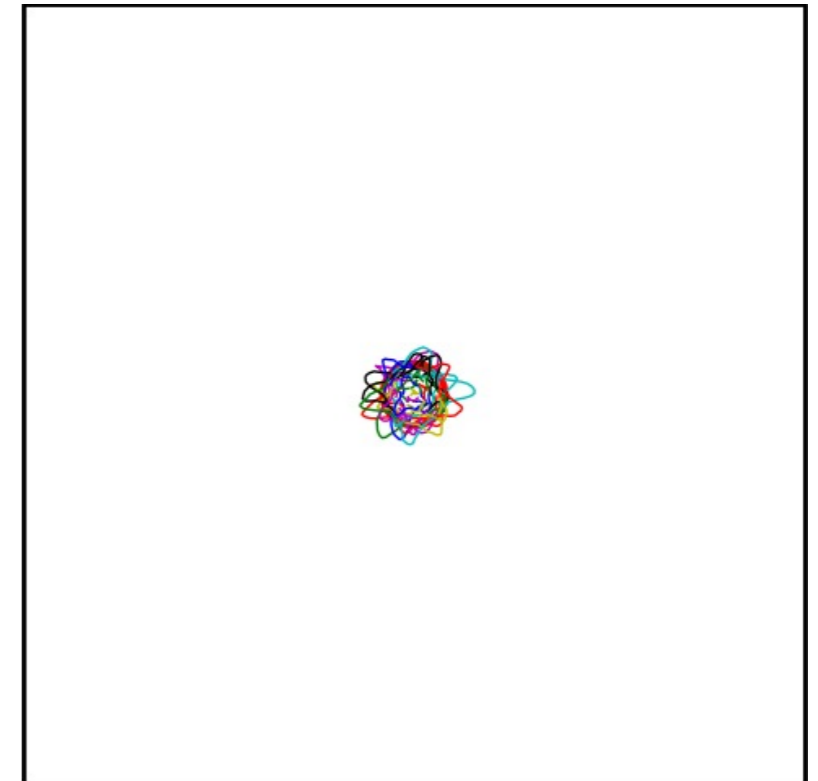
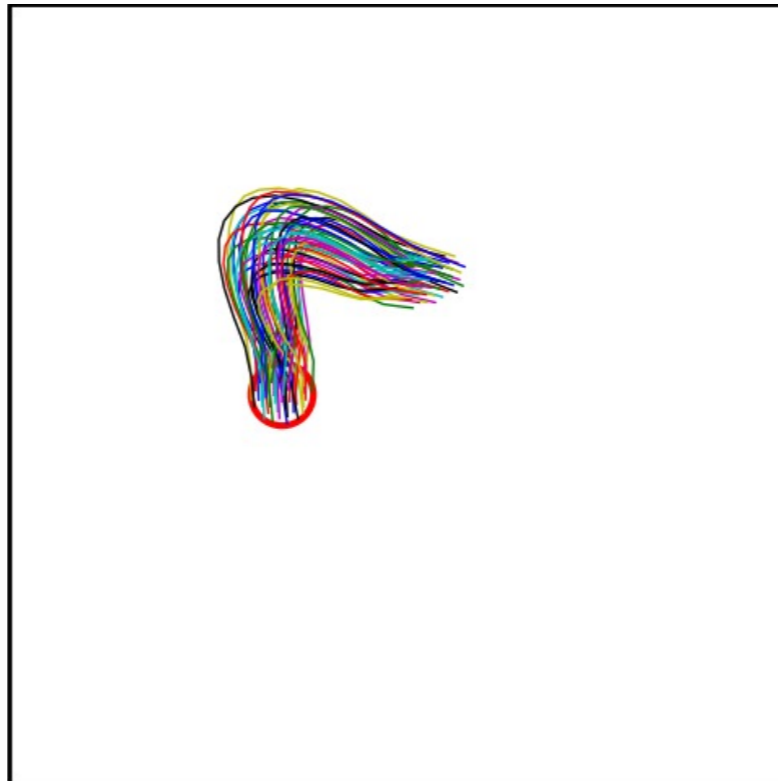
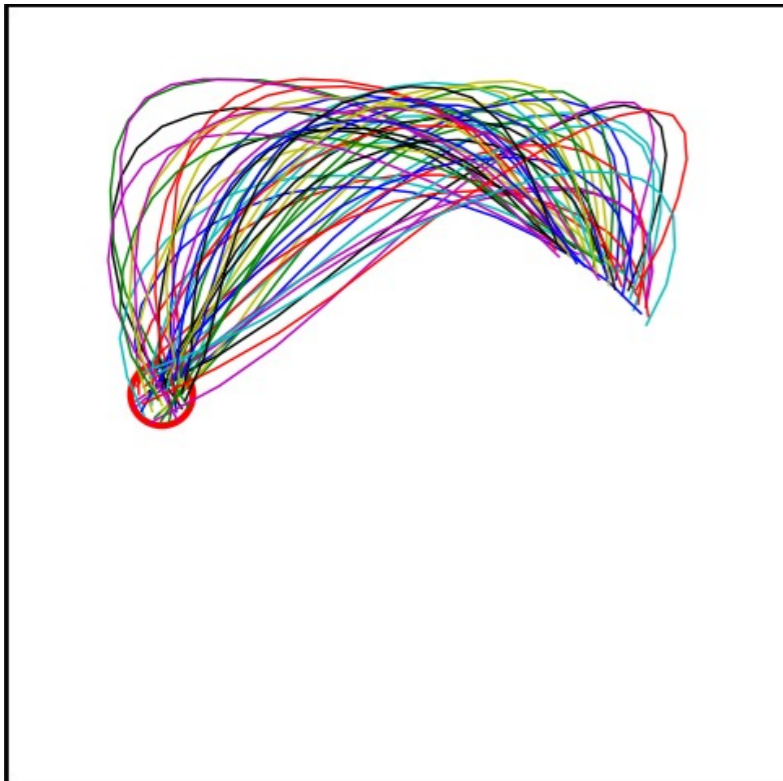


NB: qualitatively similar results by Luca Nenna and J.D. Benamou

Numerical result: Comparison of Trajectories



Disk inversion



Square, $t_{\max} = 1.5$

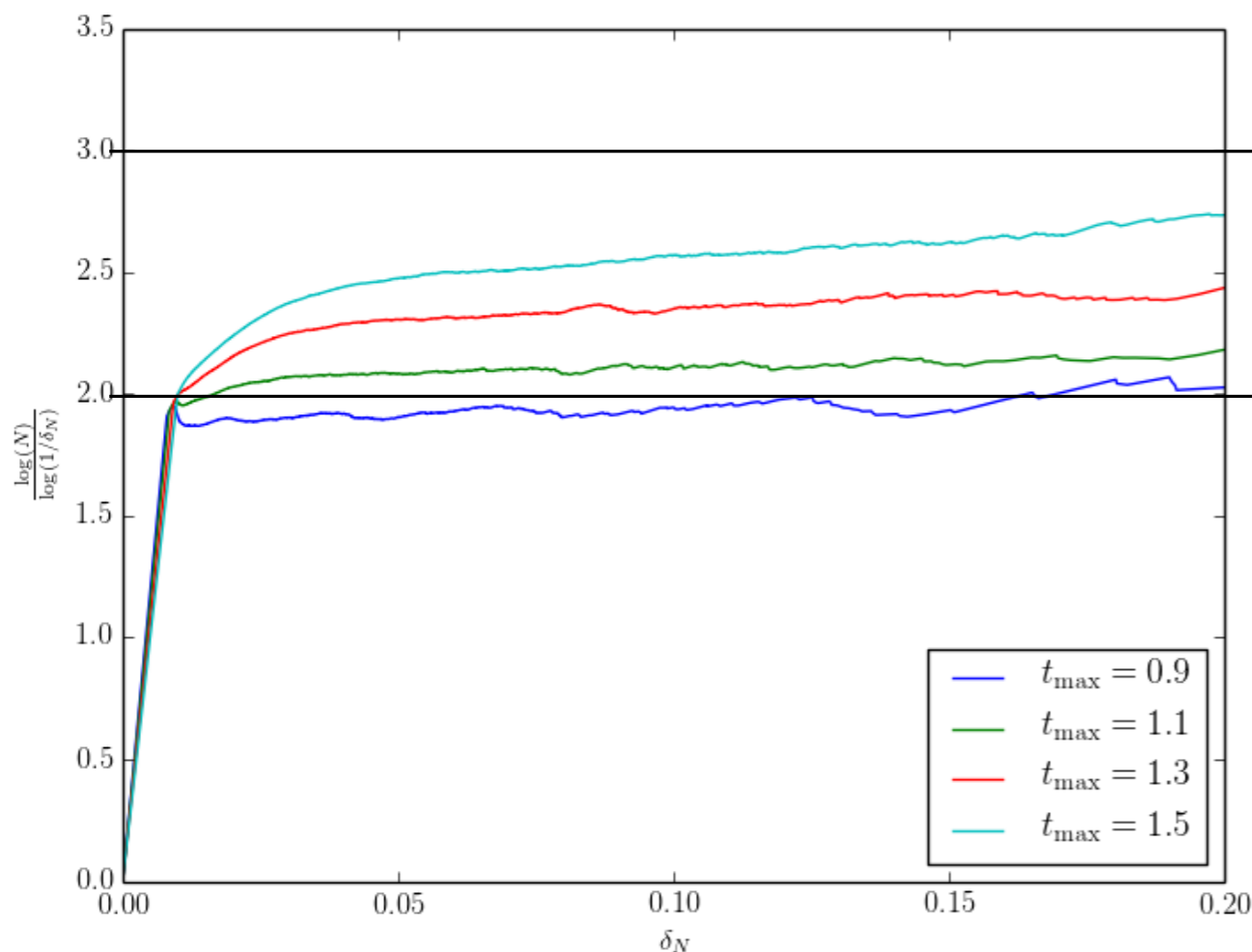
Comparison of Minkowski dimensions

Minkowski dimension Let $S \subseteq \Gamma$ be a compact subset of a metric space.

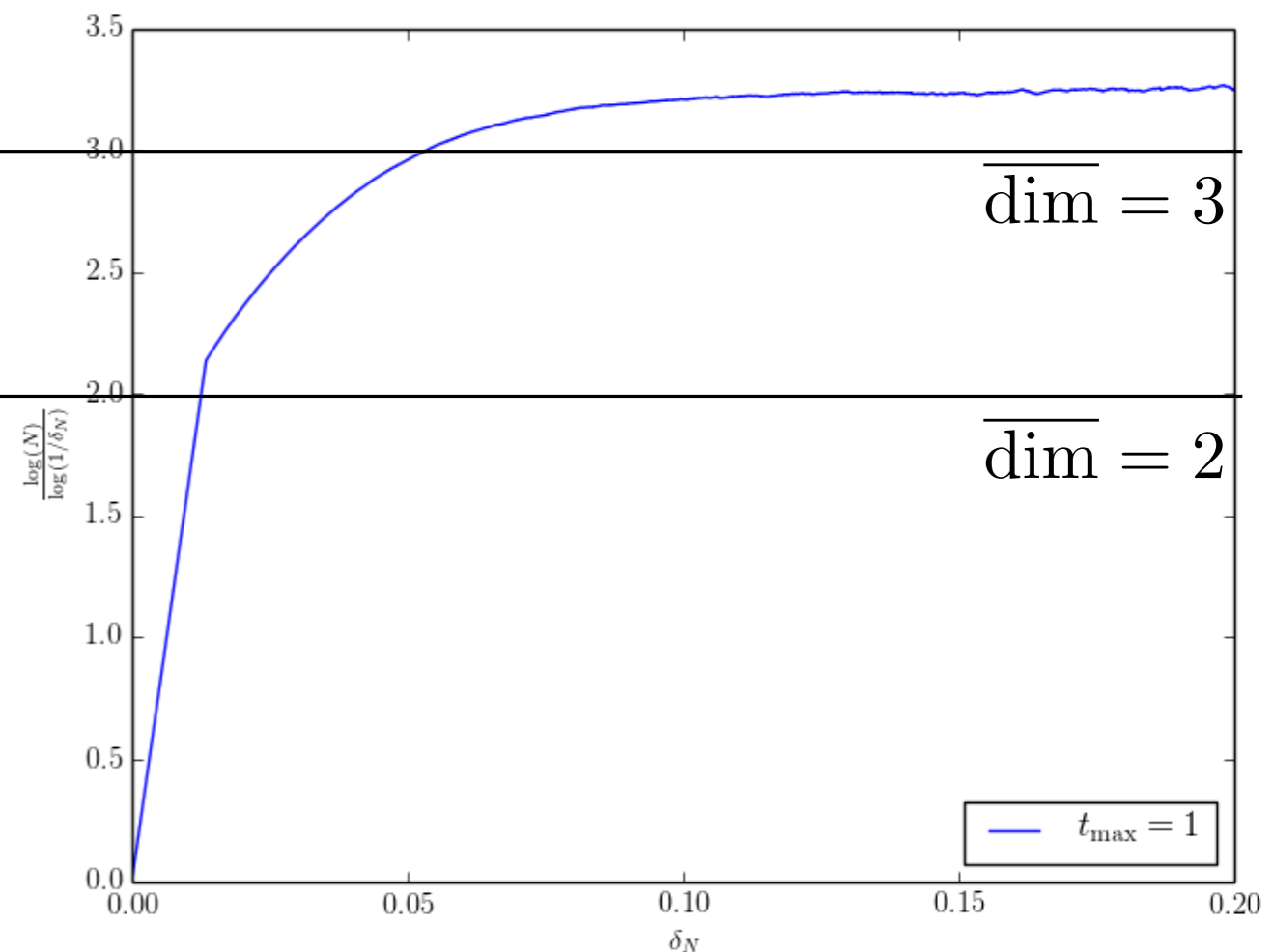
$$\overline{\dim}(S) = \limsup_{N \rightarrow \infty} \log(N) / \log(1/\delta_N)$$

where $\delta_N =$ minimum radius required to cover S with N balls.

Estimation of $\dim(\text{spt}(\mu))$ via $\log(N) / \log(1/\delta_N)$



Square rotation, $t_{\max} \in \{0.9, 1.1, 1.3, 1.5\}$



Disk inversion

Perspectives

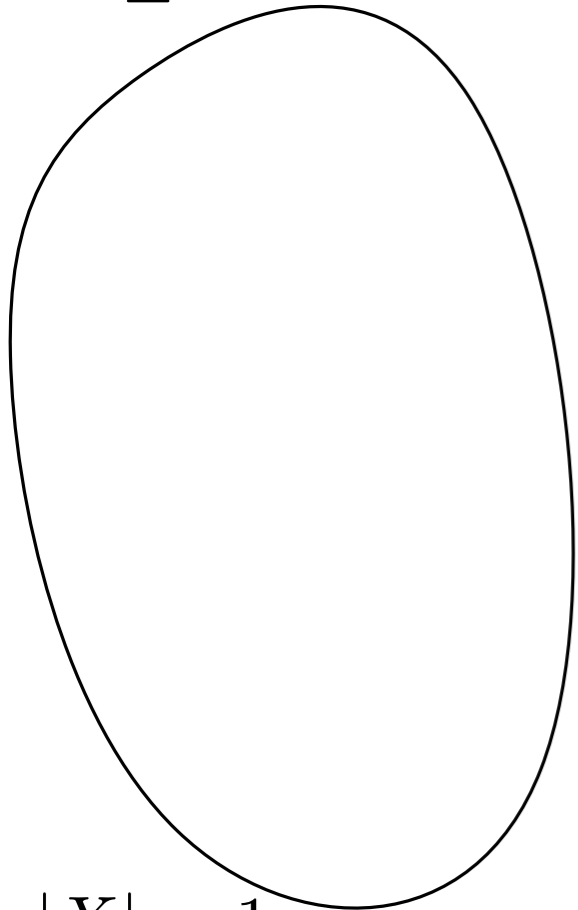
- A) More realistic numerical schemes for the Cauchy problem (e.g. without ε) ?
- B) Changing the polar factorization theorem \longrightarrow other fluid models,
e.g. fluid-structure interactions / Camassa-Holm equation [Gallouet-Vialard 16],
pressureless Euler equation with congestion [Maury-Preux '15]
- C) Viscosity?

Solutions to Euler's equations as geodesics in \mathcal{SDiff}

$\mathcal{SDiff} =$ measure-preserving diffeomorphisms from X to itself $\subseteq L^2(X, \mathbb{R}^d)$

[Arnold 1966]

$X \subseteq \mathbb{R}^d$ bounded



$|X| = 1$

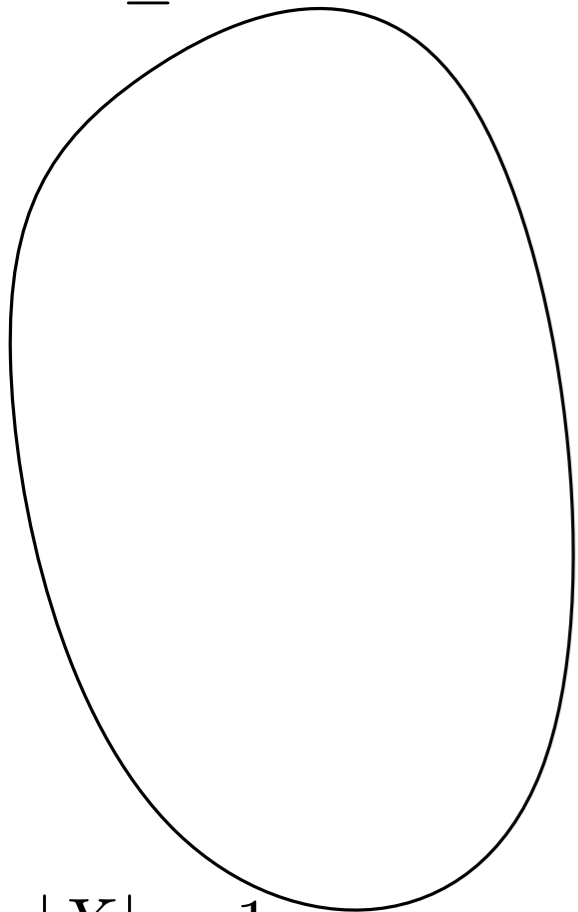
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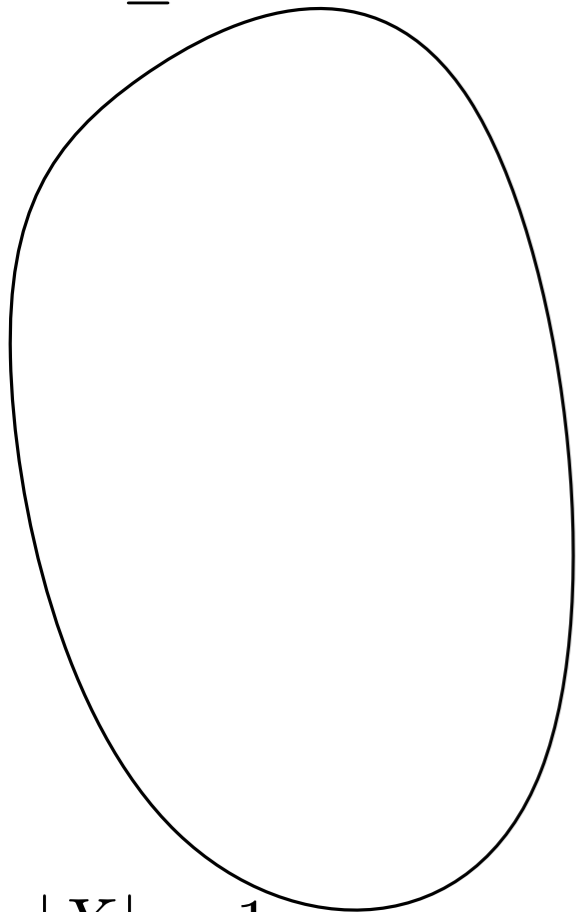
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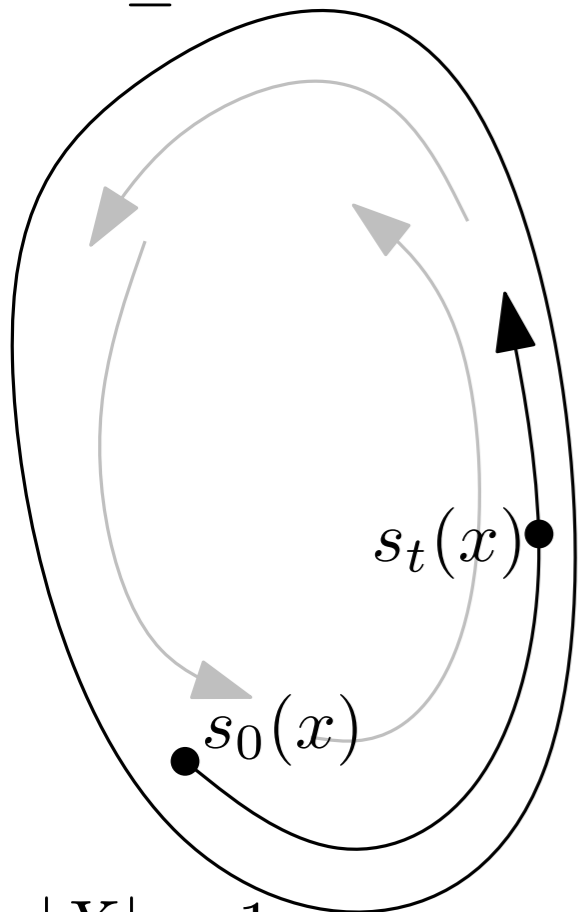
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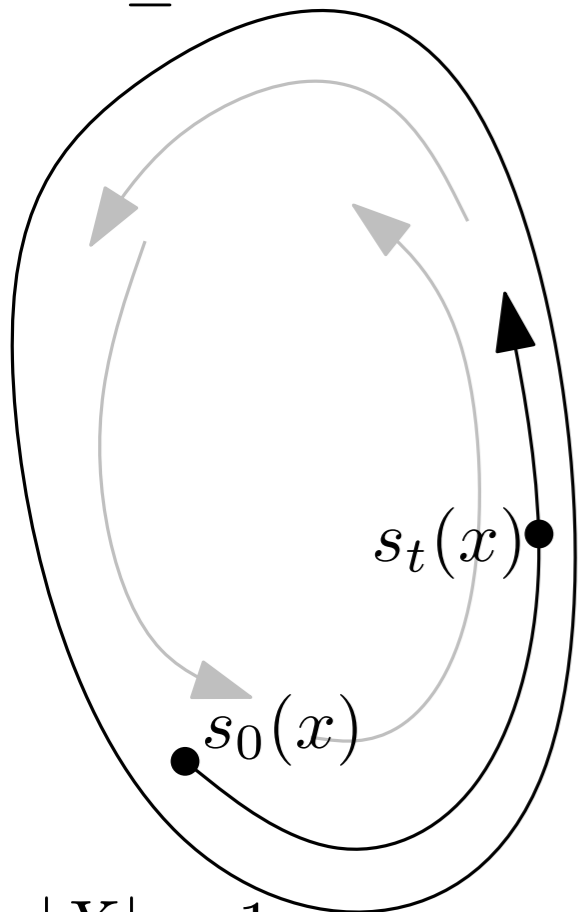
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Solutions to Euler's equations as geodesics in \mathcal{SDiff}

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[Arnold 1966]

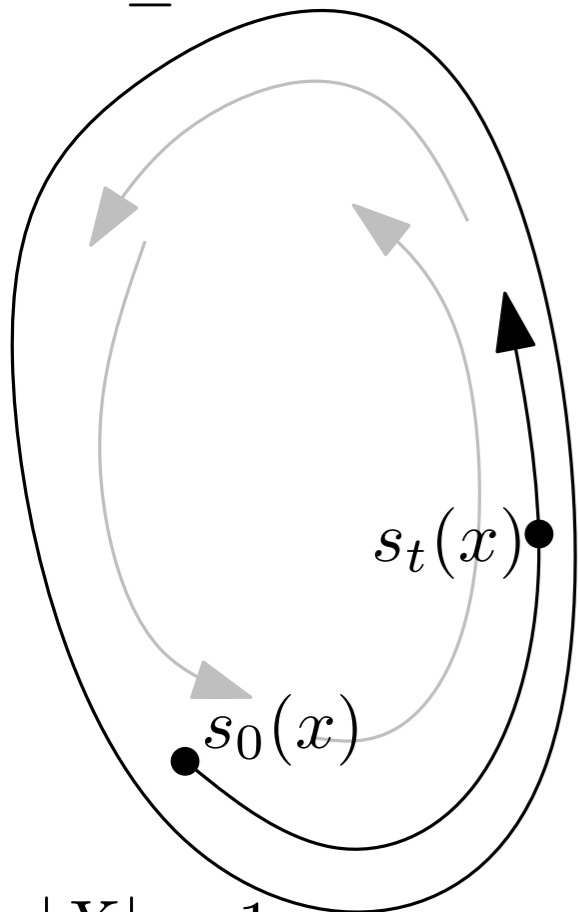
→ $T_{\text{id}}\mathcal{SDiff} = \text{divergence-free vector fields}$
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→ Formally, a path $s : [0, 1] \rightarrow \mathcal{SDiff}$ is a **geodesic** iff

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$$\iff \exists p : [0, 1] \times X \rightarrow \mathbb{R}, \ddot{s}_t = -\nabla p_t \circ s_t$$

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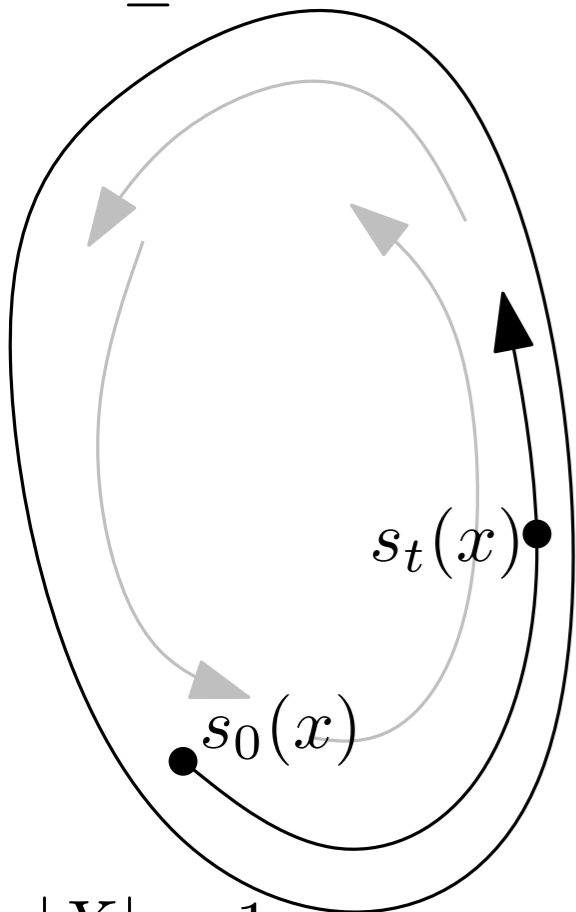
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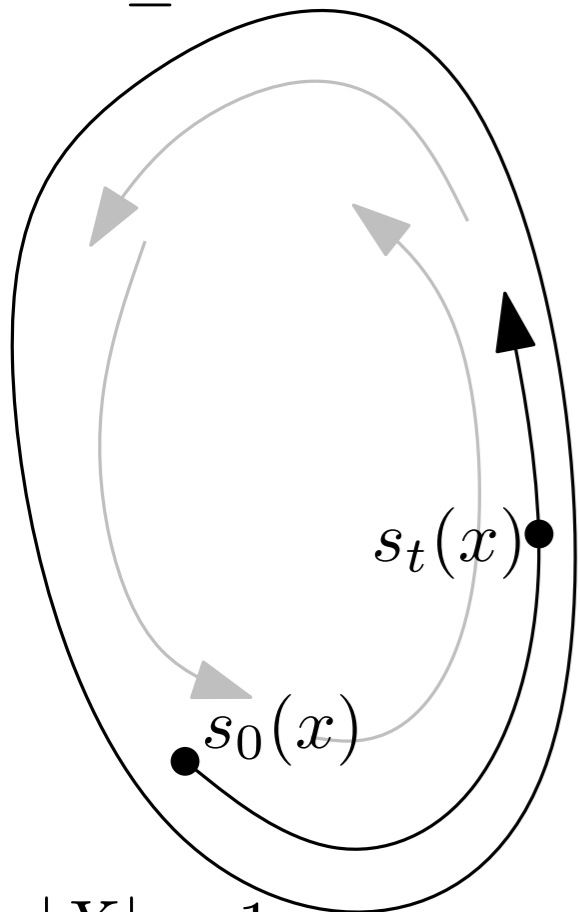
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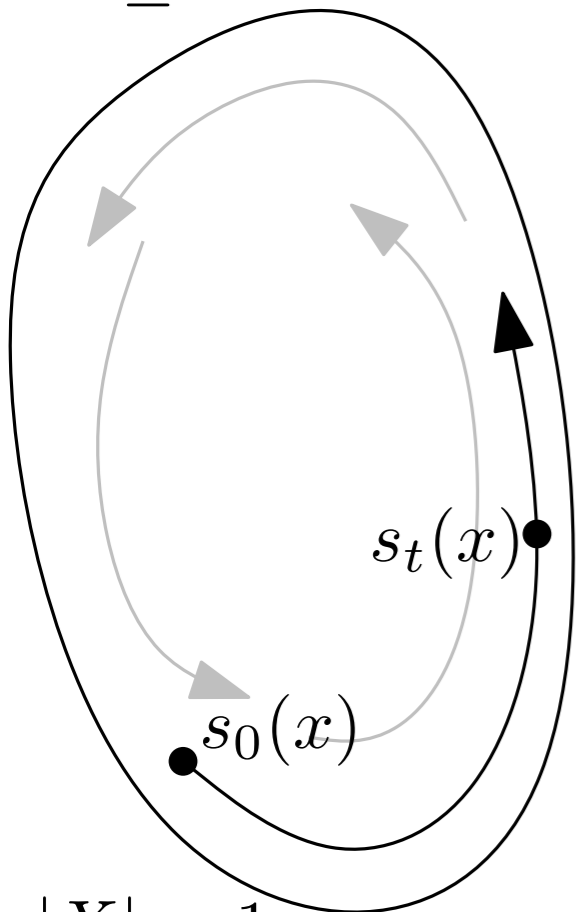
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Use this formulation for numerical computations (following Brenier):

- Minimizing geodesics (with Jean-Marie Mirebeau, 2015)
- Cauchy problem (with Thomas Gallouet, 2016).