

Anisotropic Fast-Marching methods

With applications to curvature penalization

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Jean-Marie Mirebeau

University Paris Sud, CNRS, University Paris-Saclay

February 3, 2017

Mathematical Coffees, Huawei-FSMP

In collaboration Remco Duits (Eindhoven, TU/e University),
Laurent Cohen, Da Chen (Univ. Paris-Dauphine)
Johann Dreo (Thales TRT)

This work was partly funded by ANR JCJC NS-LBR

Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

What exactly can solve the Fast-Marching Algorithm ?

The semi-
Lagrangian
paradigm

The semi-Lagrangian paradigm

The
Hamiltonian
paradigm

The Hamiltonian paradigm

Fast-Marching: the Semi-Lagrangian approach

Let X be a finite set, and $U : X \rightarrow \mathbb{R}$ be the unknown.

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Example : Dijkstra's algorithm, 1959

For each $p \in X$ let $\text{Neigh}(p) \subseteq X$ be a collection of neighbors, and $\delta(p, q)$ the corresponding positive edge lengths.

$$\Lambda U(p) := \min_{q \in \text{Neigh}(p)} U(q) + \delta(q, p).$$

(Except for seed points where $\Lambda U(p) := 0$.)

Fast-Marching: the Semi-Lagrangian approach

Let X be a finite set, and $U : X \rightarrow \mathbb{R}$ be the unknown.

A fixed point problem $\Lambda U \equiv U$ is FM-solvable. . .

provided operator $\Lambda : \mathbb{R}^X \rightarrow \mathbb{R}^X$ obeys, $\forall U, V \in \mathbb{R}^X, \forall \lambda \in \mathbb{R}$

Example : Dijkstra's algorithm, 1959

For each $p \in X$ let $\text{Neigh}(p) \subseteq X$ be a collection of neighbors, and $\delta(p, q)$ the corresponding positive edge lengths.

$$\Lambda U(p) := \min_{q \in \text{Neigh}(p)} U(q) + \delta(q, p).$$

(Except for seed points where $\Lambda U(p) := 0$.)

Fast-Marching: the Semi-Lagrangian approach

Let X be a finite set, and $U : X \rightarrow \mathbb{R}$ be the unknown.

A fixed point problem $\Lambda U \equiv U$ is FM-solvable. . .

provided operator $\Lambda : \mathbb{R}^X \rightarrow \mathbb{R}^X$ obeys, $\forall U, V \in \mathbb{R}^X, \forall \lambda \in \mathbb{R}$

▶ (Monotony) $U \leq V \Rightarrow \Lambda U \leq \Lambda V$.

Example : Dijkstra's algorithm, 1959

For each $p \in X$ let $\text{Neigh}(p) \subseteq X$ be a collection of neighbors, and $\delta(p, q)$ the corresponding positive edge lengths.

$$\Lambda U(p) := \min_{q \in \text{Neigh}(p)} U(q) + \delta(q, p).$$

(Except for seed points where $\Lambda U(p) := 0$.)

Fast-Marching: the Semi-Lagrangian approach

Let X be a finite set, and $U : X \rightarrow \mathbb{R}$ be the unknown.

A fixed point problem $\Lambda U \equiv U$ is FM-solvable. . .

provided operator $\Lambda : \mathbb{R}^X \rightarrow \mathbb{R}^X$ obeys, $\forall U, V \in \mathbb{R}^X, \forall \lambda \in \mathbb{R}$

- ▶ (Monotony) $U \leq V \Rightarrow \Lambda U \leq \Lambda V$.
- ▶ (Causality) $U^{<\lambda} = V^{<\lambda} \Rightarrow (\Lambda U)^{\leq \lambda} = (\Lambda V)^{\leq \lambda}$.

Example : Dijkstra's algorithm, 1959

For each $p \in X$ let $\text{Neigh}(p) \subseteq X$ be a collection of neighbors, and $\delta(p, q)$ the corresponding positive edge lengths.

$$\Lambda U(p) := \min_{q \in \text{Neigh}(p)} U(q) + \delta(q, p).$$

(Except for seed points where $\Lambda U(p) := 0$.)

Fast-Marching: the Hamiltonian approach

Let X be a finite set, and $s : X \rightarrow \mathbb{R}_+$ be a speed function.

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Example : upwind discretization of $\|\nabla u\|^2 = s^2$. (Rouy,92)

Assume that $X \subseteq h\mathbb{Z}^d$ is a cartesian grid, and let (e_i) be the canonical basis. Define for $U \in \mathbb{R}^X$, $p \in X$

$$HU(p) := h^{-2} \sum_{1 \leq i \leq d} \max\{0, U(p) - U(p + he_i), U(p) - U(p - he_i)\}^2.$$

Fast-Marching: the Hamiltonian approach

Let X be a finite set, and $s : X \rightarrow \mathbb{R}_+$ be a speed function.

An inverse problem $HU \equiv s^2$ is FM-solvable. . .

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Example : upwind discretization of $\|\nabla u\|^2 = s^2$. (Rouy,92)

Assume that $X \subseteq h\mathbb{Z}^d$ is a cartesian grid, and let (e_i) be the canonical basis. Define for $U \in \mathbb{R}^X$, $p \in X$

$$HU(p) \approx \sum_{1 \leq i \leq d} \left(\frac{\partial U}{\partial x_i}(p) \right)^2 = \|\nabla U\|^2.$$

Fast-Marching: the Hamiltonian approach

Let X be a finite set, and $s : X \rightarrow \mathbb{R}_+$ be a speed function.

An inverse problem $HU \equiv s^2$ is FM-solvable. . .

provided operator H has the following form

$$HU(p) := \mathcal{H}(p, U(p), (U(p) - U(q))_{q \in X}),$$

and satisfies

Example : upwind discretization of $\|\nabla u\|^2 = s^2$. (Rouy,92)

Assume that $X \subseteq h\mathbb{Z}^d$ is a cartesian grid, and let (e_i) be the canonical basis. Define for $U \in \mathbb{R}^X$, $p \in X$

$$HU(p) := h^{-2} \sum_{1 \leq i \leq d} \max\{0, U(p) - U(p + he_i), U(p) - U(p - he_i)\}^2.$$

Fast-Marching: the Hamiltonian approach

Let X be a finite set, and $s : X \rightarrow \mathbb{R}_+$ be a speed function.

An inverse problem $HU \equiv s^2$ is FM-solvable. . .

provided operator H has the following form

$$HU(p) := \mathcal{H}(p, U(p), (U(p) - U(q))_{q \in X}),$$

and satisfies

- ▶ (Monotony) \mathcal{H} is non-decreasing w.r.t. 2nd and 3rd var.

Example : upwind discretization of $\|\nabla u\|^2 = s^2$. (Rouy,92)

Assume that $X \subseteq h\mathbb{Z}^d$ is a cartesian grid, and let (e_i) be the canonical basis. Define for $U \in \mathbb{R}^X$, $p \in X$

$$HU(p) := h^{-2} \sum_{1 \leq i \leq d} \max\{0, U(p) - U(p + he_i), U(p) - U(p - he_i)\}^2.$$

Fast-Marching: the Hamiltonian approach

Let X be a finite set, and $s : X \rightarrow \mathbb{R}_+$ be a speed function.

An inverse problem $HU \equiv s^2$ is FM-solvable. . .

provided operator H has the following form

$$HU(p) := \mathcal{H}(p, U(p), (U(p) - U(q))_{q \in X}),$$

and satisfies

- ▶ (Monotony) \mathcal{H} is non-decreasing w.r.t. 2nd and 3rd var.
- ▶ (Causality) \mathcal{H} only depends on the positive part of the third variable(s).

Example : upwind discretization of $\|\nabla u\|^2 = s^2$. (Rouy,92)

Assume that $X \subseteq h\mathbb{Z}^d$ is a cartesian grid, and let (e_i) be the canonical basis. Define for $U \in \mathbb{R}^X$, $p \in X$

$$HU(p) := h^{-2} \sum_{1 \leq i \leq d} \max\{0, U(p) - U(p + he_i), U(p) - U(p - he_i)\}^2.$$

Anisotropic
Fast-
Marching

What we want to solve

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

What we want to solve

Setting: Finsler geometry

Consider a domain, a metric, and a speed function

$$\Omega \subseteq \mathbb{R}^d, \quad \mathcal{F} : \bar{\Omega} \times \mathbb{R}^d \rightarrow [0, +\infty], \quad s : \bar{\Omega} \rightarrow]0, \infty[.$$

Define for each smooth path $\gamma : [0, 1] \rightarrow \bar{\Omega}$

$$\text{length}_{\mathcal{F}}(\gamma) := \int_0^1 \mathcal{F}_{\gamma(t)}(\dot{\gamma}(t)) \frac{dt}{s(\gamma(t))}.$$

What we want to solve

Setting: Finsler geometry

Consider a domain, a metric, and a speed function

$$\Omega \subseteq \mathbb{R}^d, \quad \mathcal{F} : \bar{\Omega} \times \mathbb{R}^d \rightarrow [0, +\infty], \quad s : \bar{\Omega} \rightarrow]0, \infty[.$$

Define for each smooth path $\gamma : [0, 1] \rightarrow \bar{\Omega}$

$$\text{length}_{\mathcal{F}}(\gamma) := \int_0^1 \mathcal{F}_{\gamma(t)}(\dot{\gamma}(t)) \frac{dt}{s(\gamma(t))}.$$

Objective: compute a front arrival time

Given a set of seeds $S \subseteq \bar{\Omega}$ compute $u : \bar{\Omega} \rightarrow \mathbb{R}$ defined by

$$u(p) := \inf \{ \text{length}_{\mathcal{F}}(\gamma); \gamma(0) \in S, \gamma(1) = p \},$$

and extract the corresponding minimal paths.

**Anisotropic
Fast-
Marching**

**Jean-Marie
Mirebeau**

What exactly
can solve the
Fast-
Marching
Algorithm ?

**The semi-
Lagrangian
paradigm**

The
Hamiltonian
paradigm

What exactly can solve the Fast-Marching Algorithm ?

The semi-Lagrangian paradigm

The Hamiltonian paradigm

Using notations Ω (domain), S (seeds), u (front arrival time),
 \mathcal{F} (metric), s (speed function).

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Using notations Ω (domain), S (seeds), u (front arrival time), \mathcal{F} (metric), s (speed function).

Bellman's optimality principle

$$q \in V \subseteq \Omega \setminus S \quad \Rightarrow \quad u(q) = \inf_{p \in \partial V} u(p) + d_{\mathcal{F}}(p, q).$$

where $d_{\mathcal{F}}(q, p)$ is the length of the shortest path from p to q .

Using notations Ω (domain), S (seeds), u (front arrival time), \mathcal{F} (metric), s (speed function).

Bellman's optimality principle

$$q \in V \subseteq \Omega \setminus S \quad \Rightarrow \quad u(q) = \inf_{p \in \partial V} u(p) + d_{\mathcal{F}}(p, q).$$

where $d_{\mathcal{F}}(q, p)$ is the length of the shortest path from p to q .

Discretization

Let $X \subseteq \Omega$ and $\partial X \subseteq \mathbb{R}^d \setminus \Omega$ be finite sets. Let $V(q)$ be a polytope enclosing each $q \in X$, with vertices in $X \cup \partial X$. Define

$$\Lambda U(q) = \min_{p \in \partial V(q)} \mathcal{F}_p(q - p) + I_{V(q)} U(p),$$

where I_V denotes piecewise linear interpolation on V .

Using notations Ω (domain), S (seeds), u (front arrival time), \mathcal{F} (metric), s (speed function).

Bellman's optimality principle

$$q \in V \subseteq \Omega \setminus S \quad \Rightarrow \quad u(q) = \inf_{p \in \partial V} u(p) + d_{\mathcal{F}}(p, q).$$

where $d_{\mathcal{F}}(q, p)$ is the length of the shortest path from p to q .

Discretization

Let $X \subseteq \Omega$ and $\partial X \subseteq \mathbb{R}^d \setminus \Omega$ be finite sets. Let $V(q)$ be a polytope enclosing each $q \in X$, with vertices in $X \cup \partial X$. Define

$$\Lambda U(q) = \min_{p \in \partial V(q)} \mathcal{F}_p(q - p) + I_{V(q)} U(p),$$

where I_V denotes piecewise linear interpolation on V .

- Monotony holds by construction.

Using notations Ω (domain), S (seeds), u (front arrival time), \mathcal{F} (metric), s (speed function).

Bellman's optimality principle

$$q \in V \subseteq \Omega \setminus S \quad \Rightarrow \quad u(q) = \inf_{p \in \partial V} u(p) + d_{\mathcal{F}}(p, q).$$

where $d_{\mathcal{F}}(q, p)$ is the length of the shortest path from p to q .

Discretization

Let $X \subseteq \Omega$ and $\partial X \subseteq \mathbb{R}^d \setminus \Omega$ be finite sets. Let $V(q)$ be a polytope enclosing each $q \in X$, with vertices in $X \cup \partial X$. Define

$$\Lambda U(q) = \min_{p \in \partial V(q)} \mathcal{F}_p(q - p) + I_{V(q)} U(p),$$

where I_V denotes piecewise linear interpolation on V .

- ▶ Monotony holds by construction.
- ▶ Causality is equivalent to the acuteness of $V(p)$ w.r.t. \mathcal{F}_p .

Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

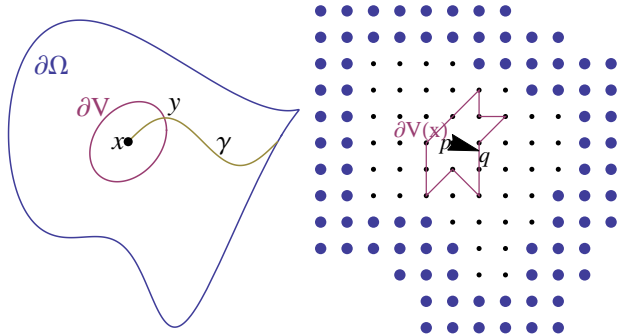


Figure: Illustration of Bellman's optimality principle, and of its discretization.

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy.

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

- ▶ $\langle v, w \rangle \geq 0$, assuming $F(e) := \lambda \|e\|$. (Euclidean)

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy.

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

- ▶ $\langle v, w \rangle \geq 0$, assuming $F(e) := \lambda \|e\|$. (Euclidean)
- ▶ $\langle v, Mw \rangle \geq 0$ assuming $F(e) := \langle e, Me \rangle$. (Riemannian)

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy.

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

- ▶ $\langle v, w \rangle \geq 0$, assuming $F(e) := \lambda \|e\|$. (Euclidean)
- ▶ $\langle v, Mw \rangle \geq 0$ assuming $F(e) := \langle e, Me \rangle$. (Riemannian)
- ▶ $\langle v, \nabla F(w) \rangle \geq 0$ and $\langle w, \nabla F(v) \rangle \geq 0$ in general (Finsler)

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy.

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

- ▶ $\langle v, w \rangle \geq 0$, assuming $F(e) := \lambda \|e\|$. (Euclidean)
- ▶ $\langle v, Mw \rangle \geq 0$ assuming $F(e) := \langle e, Me \rangle$. (Riemannian)
- ▶ $\langle v, \nabla F(w) \rangle \geq 0$ and $\langle w, \nabla F(v) \rangle \geq 0$ in general (Finsler)

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy. In our applications $\mu \gtrsim 10 \Rightarrow$ completely impractical.

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

- ▶ $\langle v, w \rangle \geq 0$, assuming $F(e) := \lambda \|e\|$. (Euclidean)
- ▶ $\langle v, Mw \rangle \geq 0$ assuming $F(e) := \langle e, Me \rangle$. (Riemannian)
- ▶ $\langle v, \nabla F(w) \rangle \geq 0$ and $\langle w, \nabla F(v) \rangle \geq 0$ in general (Finsler)

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy. In our applications $\mu \gtrsim 10 \Rightarrow$ completely impractical.

A polytope design problem

Given an asymmetric norm N on \mathbb{R}^d , find a polytope V which

- ▶ Is acute with respect to N . (\Rightarrow causality)

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

- ▶ $\langle v, w \rangle \geq 0$, assuming $F(e) := \lambda \|e\|$. (Euclidean)
- ▶ $\langle v, Mw \rangle \geq 0$ assuming $F(e) := \langle e, Me \rangle$. (Riemannian)
- ▶ $\langle v, \nabla F(w) \rangle \geq 0$ and $\langle w, \nabla F(v) \rangle \geq 0$ in general (Finsler)

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy. In our applications $\mu \gtrsim 10 \Rightarrow$ completely impractical.

A polytope design problem

Given an asymmetric norm N on \mathbb{R}^d , find a polytope V which

- ▶ Is acute with respect to N . (\Rightarrow causality)
- ▶ Has its vertices in \mathbb{Z}^d . (\Rightarrow cartesian grid discretizations)

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

- ▶ $\langle v, w \rangle \geq 0$, assuming $F(e) := \lambda \|e\|$. (Euclidean)
- ▶ $\langle v, Mw \rangle \geq 0$ assuming $F(e) := \langle e, Me \rangle$. (Riemannian)
- ▶ $\langle v, \nabla F(w) \rangle \geq 0$ and $\langle w, \nabla F(v) \rangle \geq 0$ in general (Finsler)

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy. In our applications $\mu \gtrsim 10 \Rightarrow$ completely impractical.

A polytope design problem

Given an asymmetric norm N on \mathbb{R}^d , find a polytope V which

- ▶ Is acute with respect to N . (\Rightarrow causality)
- ▶ Has its vertices in \mathbb{Z}^d . (\Rightarrow cartesian grid discretizations)
- ▶ Has few vertices. (\Rightarrow complexity)

Definition (Acute polytope V w.r.t. a metric F)

A polytope V centered at 0 is said F -acute iff for any v, w in a common face of ∂V .

- ▶ $\langle v, w \rangle \geq 0$, assuming $F(e) := \lambda \|e\|$. (Euclidean)
- ▶ $\langle v, Mw \rangle \geq 0$ assuming $F(e) := \langle e, Me \rangle$. (Riemannian)
- ▶ $\langle v, \nabla F(w) \rangle \geq 0$ and $\langle w, \nabla F(v) \rangle \geq 0$ in general (Finsler)

Constructions proposed by Sethian & Vladimirsky (03), Alton & Mitchell (2010), with μ^d vertices where μ measures anisotropy. In our applications $\mu \gtrsim 10 \Rightarrow$ completely impractical.

A polytope design problem

Given an asymmetric norm N on \mathbb{R}^d , find a polytope V which

- ▶ Is acute with respect to N . (\Rightarrow causality)
- ▶ Has its vertices in \mathbb{Z}^d . (\Rightarrow cartesian grid discretizations)
- ▶ Has few vertices. (\Rightarrow complexity)
- ▶ Has small vertices. (\Rightarrow accuracy)

Delaunay stars for 3D Riemannian metrics

Needle-like

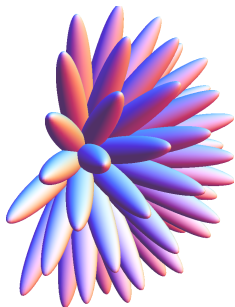
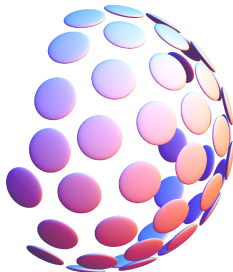


Plate-like



What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

► Metric $\mathcal{F}_p(v) := \sqrt{v^T M(p)v}$.

Delaunay stars for 3D Riemannian metrics

Needle-like

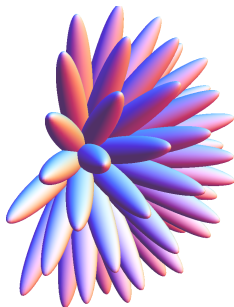
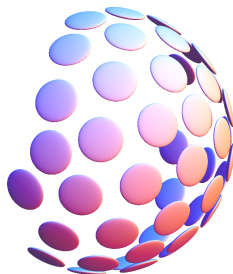


Plate-like



What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

- ▶ Metric $\mathcal{F}_p(v) := \sqrt{v^T M(p)v}$.
- ▶ Define $V(p)$ as the union of all simplices containing p in the Delaunay triangulation of X w.r.t. the distance $d_p(x, y) := \sqrt{(x - y)^T M(p)(x - y)}$.

Delaunay stars for 3D Riemannian metrics

Needle-like

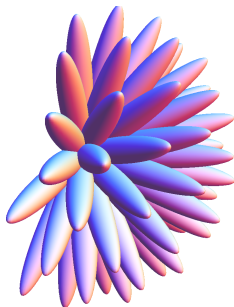
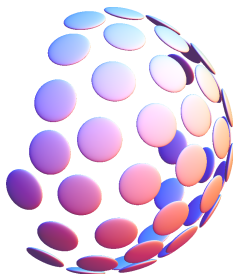


Plate-like



What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

- ▶ Metric $\mathcal{F}_p(v) := \sqrt{v^T M(p)v}$.
- ▶ Define $V(p)$ as the union of all simplices containing p in the Delaunay triangulation of X w.r.t. the distance $d_p(x, y) := \sqrt{(x - y)^T M(p)(x - y)}$.
- ▶ If X is a cartesian grid, then $V(p)$ is \mathcal{F}_p acute.

Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

Delaunay stars for 3D Riemannian metrics

Needle-like

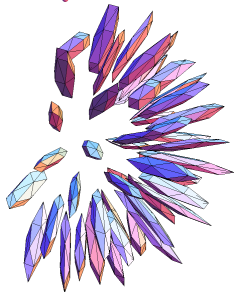
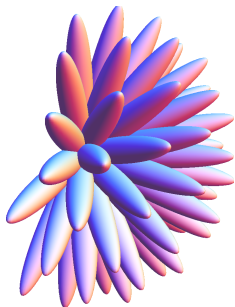
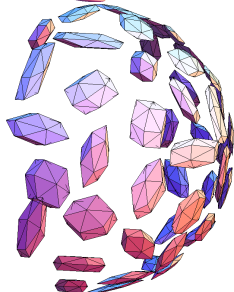
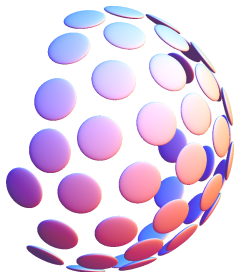


Plate-like



What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

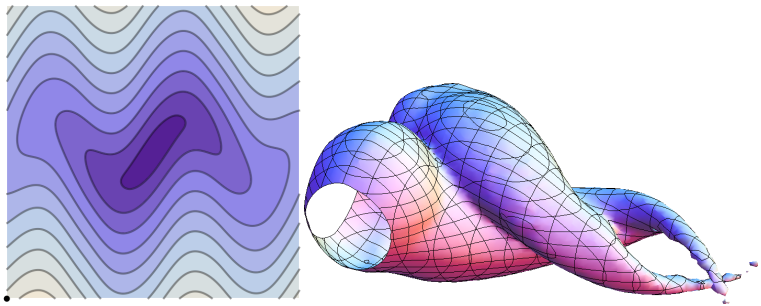


Figure: Some level sets of 2D and 3D riemannian distance maps.

Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

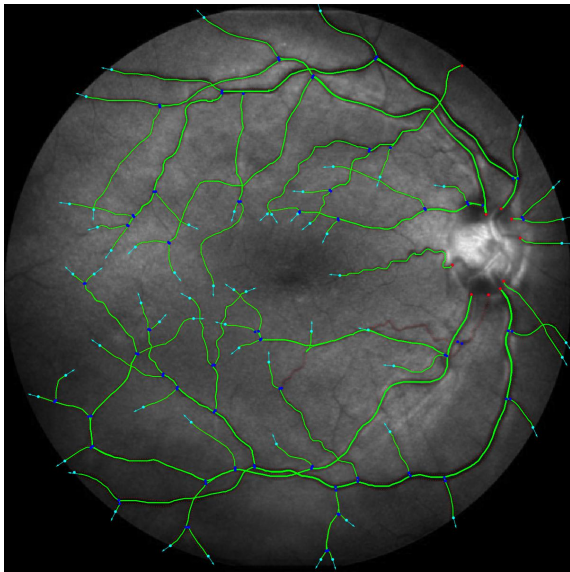


Figure: Segmentation of retina vessels. 📄 G. Sanguinetti, E. Bekkers, R. Duits, M.H.J. Janssen, A. Mashtakov, J.M. Mirebeau, Sub-Riemannian *Fast Marching in SE(2)*, CIARP 2015.

Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

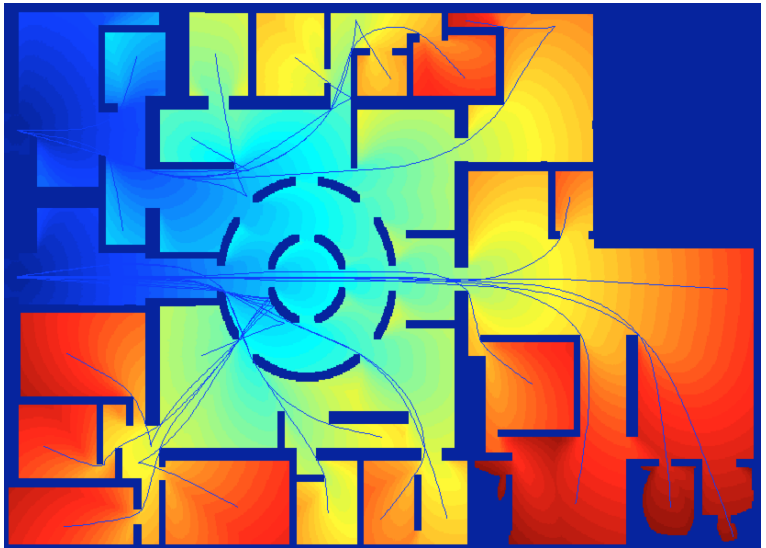


Figure: Shortest way out of centre Pompidou, using a Reeds-Shepp sub-riemannian metric. Note the many cusps.

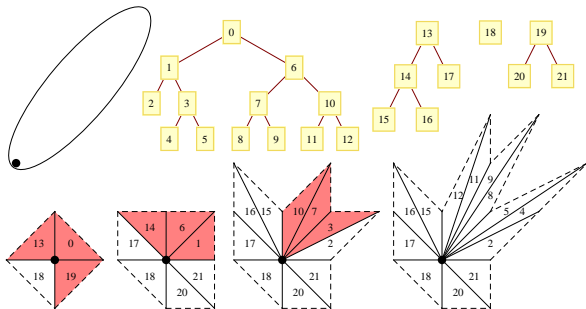
Stencil refinement strategy for 2D Finsler metrics

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm



Anisotropic
Fast-
Marching

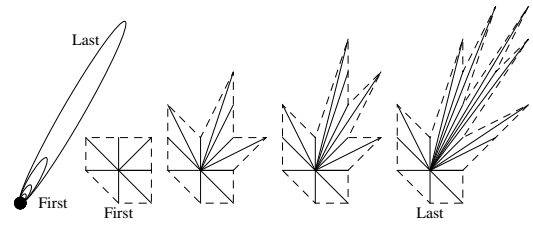
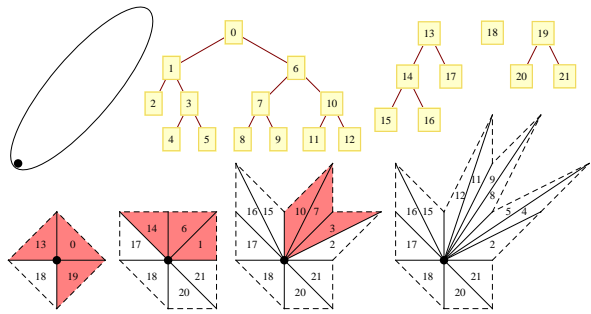
Stencil refinement strategy for 2D Finsler metrics

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm



Anisotropic
Fast-
Marching

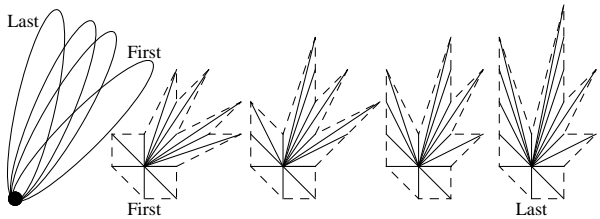
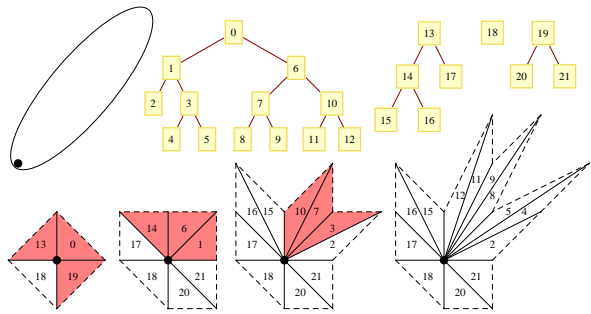
Stencil refinement strategy for 2D Finsler metrics

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm



Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

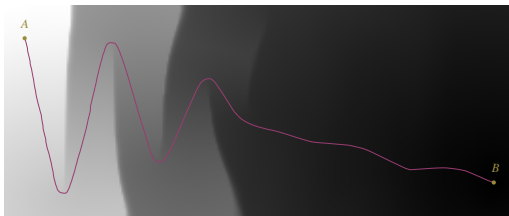
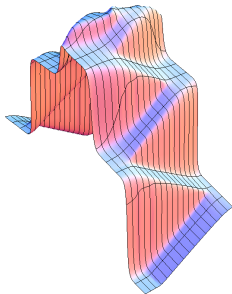


Figure: Finsler metrics can encode asymmetrical situations, e.g. ascent is harder than descent

Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

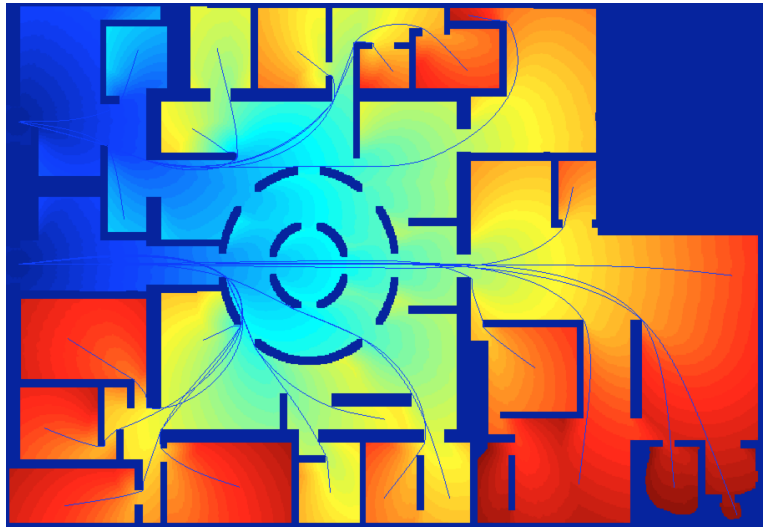


Figure: Shortest way out of centre Pompidou, using a Reeds-Shepp sub-riemannian metric modified to remove the reverse gear.

Conclusion on Semi-Lagrangian

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Pros:

- ▶ Geometrical interpretation.
- ▶ Stencil recipes for 2D Finsler or 3D riemannian metrics on grids.

Cons:

- ▶ No good stencil recipe for 3D Finsler metrics, or for unstructured meshes.
- ▶ A bit costly (iterate over all facets of $V(p)$ of all dims).
- ▶ Rather complex implementation in dimension ≥ 3 .

**Anisotropic
Fast-
Marching**

**Jean-Marie
Mirebeau**

What exactly
can solve the
Fast-
Marching
Algorithm ?

What exactly can solve the Fast-Marching Algorithm ?

The semi-
Lagrangian
paradigm

The semi-Lagrangian paradigm

The
Hamiltonian
paradigm

The Hamiltonian paradigm

Using notations Ω (domain), S (seeds), u (front arrival time), \mathcal{F} (metric), s (speed function).

Generalized eikonal equation

Front arrival times are the unique viscosity solution to

$$\mathcal{H}_p(\nabla u(p)) = s(p)^2$$

for all $p \in \Omega \setminus S$, with $u = 0$ on S , and outflow boundary conditions.

Using notations Ω (domain), S (seeds), u (front arrival time), \mathcal{F} (metric), s (speed function).

Generalized eikonal equation

Front arrival times are the unique viscosity solution to

$$\mathcal{H}_p(\nabla u(p)) = s(p)^2$$

for all $p \in \Omega \setminus S$, with $u = 0$ on S , and outflow boundary conditions. The hamiltonian is defined by

$$\frac{1}{2}\mathcal{H}_p(v) := \sup_{w \in \mathbb{R}^d} \langle v, w \rangle - \frac{1}{2}\mathcal{F}_p(w)^2.$$

Using notations Ω (domain), S (seeds), u (front arrival time), \mathcal{F} (metric), s (speed function).

Generalized eikonal equation

Front arrival times are the unique viscosity solution to

$$\mathcal{H}_p(\nabla u(p)) = s(p)^2$$

for all $p \in \Omega \setminus S$, with $u = 0$ on S , and outflow boundary conditions. The hamiltonian is defined by

$$\frac{1}{2}\mathcal{H}_p(v) := \sup_{w \in \mathbb{R}^d} \langle v, w \rangle - \frac{1}{2}\mathcal{F}_p(w)^2.$$

Discrete point set: a grid of scale $h > 0$

$$X := \Omega \cap h\mathbb{Z}^d, \quad \partial X := (\mathbb{R}^d \setminus \Omega) \cap h\mathbb{Z}^d.$$

Sum of squares representation of the Hamiltonian

Express or approximate $v \mapsto \mathcal{H}_p(v)$ in the form

$$H(v) = \sum_{1 \leq i \leq I} \alpha_i \max\{0, \langle v, e_i \rangle\}^2 + \sum_{1 \leq j \leq J} \beta_j \langle v, f_j \rangle^2,$$

where $e_i, f_j \in \mathbb{Z}^d$, $\alpha_i, \beta_j \geq 0$.

Sum of squares representation of the Hamiltonian

Express or approximate $v \mapsto \mathcal{H}_p(v)$ in the form

$$H(v) = \sum_{1 \leq i \leq I} \alpha_i \max\{0, \langle v, e_i \rangle\}^2 + \sum_{1 \leq j \leq J} \beta_j \langle v, f_j \rangle^2,$$

where $e_i, f_j \in \mathbb{Z}^d$, $\alpha_i, \beta_j \geq 0$. Or more generally in the form

$$H(v) = H_0(v) + \max_{1 \leq k \leq K} H_k(v).$$

where H_0, \dots, H_K are as above.

Sum of squares representation of the Hamiltonian

Express or approximate $v \mapsto \mathcal{H}_p(v)$ in the form

$$H(v) = \sum_{1 \leq i \leq I} \alpha_i \max\{0, \langle v, e_i \rangle\}^2 + \sum_{1 \leq j \leq J} \beta_j \langle v, f_j \rangle^2,$$

where $e_i, f_j \in \mathbb{Z}^d$, $\alpha_i, \beta_j \geq 0$. Or more generally in the form

$$H(v) = H_0(v) + \max_{1 \leq k \leq K} H_k(v).$$

where H_0, \dots, H_K are as above.

Upwind differences discretization

Approximate $H(\nabla u(p))$ by inserting

$$\max\{0, \langle \nabla u(p), e_i \rangle\} \approx h^{-1} \max\{0, U(p) - U(p - he_i)\}$$

$$|\langle \nabla u(p), e_i \rangle| \approx h^{-1} \max\{0, U(p) - U(p - he_i), U(p) - U(p + he_i)\}$$

Riemannian hamiltonians and Voronoi's reduction

- ▶ Voronoi introduced the following polytope \mathcal{P} and linear program $\mathcal{L}(D)$

$$\mathcal{P} := \{M \in S_d^{++}; \forall e \in \mathbb{Z}^d, \langle e, Me \rangle \geq 1\},$$

$$\mathcal{L}(D) := \min_{M \in \mathcal{P}} \text{Tr}(DM).$$

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Riemannian hamiltonians and Voronoi's reduction

- ▶ Voronoi introduced the following polytope \mathcal{P} and linear program $\mathcal{L}(D)$

$$\mathcal{P} := \{M \in S_d^{++}; \forall e \in \mathbb{Z}^d, \langle e, Me \rangle \geq 1\},$$
$$\mathcal{L}(D) := \min_{M \in \mathcal{P}} \text{Tr}(DM).$$

- ▶ Voronoi proved feasibility of $\mathcal{L}(D)$, for all $D \in S_d^{++}$.

Riemannian hamiltonians and Voronoi's reduction

- ▶ Voronoi introduced the following polytope \mathcal{P} and linear program $\mathcal{L}(D)$

$$\mathcal{P} := \{M \in S_d^{++}; \forall e \in \mathbb{Z}^d, \langle e, Me \rangle \geq 1\},$$
$$\mathcal{L}(D) := \min_{M \in \mathcal{P}} \text{Tr}(DM).$$

- ▶ Voronoi proved feasibility of $\mathcal{L}(D)$, for all $D \in S_d^{++}$.
- ▶ Vertices of \mathcal{P} are called perfect forms, known in $\text{dim} \leq 7$.

Riemannian hamiltonians and Voronoi's reduction

- ▶ Voronoi introduced the following polytope \mathcal{P} and linear program $\mathcal{L}(D)$

$$\mathcal{P} := \{M \in S_d^{++}; \forall e \in \mathbb{Z}^d, \langle e, Me \rangle \geq 1\},$$

$$\mathcal{L}(D) := \min_{M \in \mathcal{P}} \text{Tr}(DM).$$

- ▶ Voronoi proved feasibility of $\mathcal{L}(D)$, for all $D \in S_d^{++}$.
- ▶ Vertices of \mathcal{P} are called perfect forms, known in $\dim \leq 7$.
- ▶ Kuhn-Tucker optimality conditions: there exists $(\lambda_i, e_i) \in (\mathbb{R}_+ \times \mathbb{Z}^d)^{d'}$, where $d' = d(d+1)/2$, such that

$$D = \sum_{1 \leq i \leq d'} \lambda_i e_i \otimes e_i.$$

Riemannian hamiltonians and Voronoi's reduction

- ▶ Voronoi introduced the following polytope \mathcal{P} and linear program $\mathcal{L}(D)$

$$\mathcal{P} := \{M \in S_d^{++}; \forall e \in \mathbb{Z}^d, \text{Tr}(Me \otimes e) \geq 1\},$$
$$\mathcal{L}(D) := \min_{M \in \mathcal{P}} \text{Tr}(DM).$$

- ▶ Voronoi proved feasibility of $\mathcal{L}(D)$, for all $D \in S_d^{++}$.
- ▶ Vertices of \mathcal{P} are called perfect forms, known in $\dim \leq 7$.
- ▶ Kuhn-Tucker optimality conditions: there exists $(\lambda_i, e_i) \in (\mathbb{R}_+ \times \mathbb{Z}^d)^{d'}$, where $d' = d(d+1)/2$, such that

$$D = \sum_{1 \leq i \leq d'} \lambda_i e_i \otimes e_i.$$

Riemannian hamiltonians and Voronoi's reduction

- ▶ Voronoi introduced the following polytope \mathcal{P} and linear program $\mathcal{L}(D)$

$$\mathcal{P} := \{M \in S_d^{++}; \forall e \in \mathbb{Z}^d, \text{Tr}(Me \otimes e) \geq 1\},$$

$$\mathcal{L}(D) := \min_{M \in \mathcal{P}} \text{Tr}(DM).$$

- ▶ Voronoi proved feasibility of $\mathcal{L}(D)$, for all $D \in S_d^{++}$.
- ▶ Vertices of \mathcal{P} are called perfect forms, known in $\dim \leq 7$.
- ▶ Kuhn-Tucker optimality conditions: there exists $(\lambda_i, e_i) \in (\mathbb{R}_+ \times \mathbb{Z}^d)^{d'}$, where $d' = d(d+1)/2$, such that

$$D = \sum_{1 \leq i \leq d'} \lambda_i e_i \otimes e_i.$$

- ▶ Represents the Riemannian hamiltonian

$$H(v) := \langle v, Dv \rangle = \sum_{1 \leq i \leq d'} \lambda_i \langle v, e_i \rangle^2$$

Curvature penalized shortest paths

Define the cost of a unit speed curve $\gamma : [0, T] \rightarrow U$, with curvature κ , as

$$\int_0^T C(\kappa(t)) \frac{dt}{s(\gamma(t))}$$

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Curvature penalized shortest paths

Define the cost of a unit speed curve $\gamma : [0, T] \rightarrow U$, with curvature κ , as

$$\int_0^T C(\kappa(t)) \frac{dt}{s(\gamma(t))}$$

PDE $\mathcal{H}(\nabla u) = s$, posed on the lifted domain $\Omega = U \times \mathbb{S}^1$, with points $p = (x, \theta)$. Metric, with $n(\theta) := (\cos \theta, \sin \theta)$

$$\mathcal{F}_{(x, \theta)}(\dot{x}, \dot{\theta}) = \begin{cases} \|\dot{x}\| C(\dot{\theta} / \|\dot{x}\|) & \text{if } \dot{x} = \|\dot{x}\| n(\theta) \\ +\infty & \text{otherwise.} \end{cases}$$

Curvature penalized shortest paths

Define the cost of a unit speed curve $\gamma : [0, T] \rightarrow U$, with curvature κ , as

$$\int_0^T \mathcal{C}(\kappa(t)) \frac{dt}{s(\gamma(t))}$$

PDE $\mathcal{H}(\nabla u) = s$, posed on the lifted domain $\Omega = U \times \mathbb{S}^1$, with points $p = (x, \theta)$. Metric, with $n(\theta) := (\cos \theta, \sin \theta)$

$$\mathcal{F}_{(x,\theta)}(\dot{x}, \dot{\theta}) = \begin{cases} \|\dot{x}\| \mathcal{C}(\dot{\theta}/\|\dot{x}\|) & \text{if } \dot{x} = \|\dot{x}\| n(\theta) \\ +\infty & \text{otherwise.} \end{cases}$$

We consider three curvature costs.

- ▶ Reeds-Shepp model $\mathcal{C}(\kappa) := \sqrt{1 + \kappa^2}$
- ▶ Euler elastica model $\mathcal{C}(\kappa) := 1 + \kappa^2$
- ▶ Dubins model $\mathcal{C}(\kappa) := 1$ if $\kappa \leq 1$, and $+\infty$ otherwise.

Curvature penalized shortest paths

Define the cost of a unit speed curve $\gamma : [0, T] \rightarrow U$, with curvature κ , as

$$\int_0^T \mathcal{C}(\kappa(t)) \frac{dt}{s(\gamma(t))}$$

PDE $\mathcal{H}(\nabla u) = s$, posed on the lifted domain $\Omega = U \times \mathbb{S}^1$, with points $p = (x, \theta)$. Metric, with $n(\theta) := (\cos \theta, \sin \theta)$

$$\mathcal{F}_{(x,\theta)}(\dot{x}, \dot{\theta}) = \begin{cases} \|\dot{x}\| \mathcal{C}(\dot{\theta}/\|\dot{x}\|) & \text{if } \dot{x} = \|\dot{x}\| n(\theta) \\ +\infty & \text{otherwise.} \end{cases}$$

We consider three curvature costs.

- ▶ Reeds-Shepp model $\mathcal{C}(\kappa) := \sqrt{1 + \kappa^2}$

$$\mathcal{H}_{(x,\theta)}(\hat{x}, \hat{\theta}) = \langle \hat{x}, n(\theta) \rangle_+^2 + \hat{\theta}^2$$

- ▶ Euler elastica model $\mathcal{C}(\kappa) := 1 + \kappa^2$
- ▶ Dubins model $\mathcal{C}(\kappa) := 1$ if $\kappa \leq 1$, and $+\infty$ otherwise.

Curvature penalized shortest paths

Define the cost of a unit speed curve $\gamma : [0, T] \rightarrow U$, with curvature κ , as

$$\int_0^T \mathcal{C}(\kappa(t)) \frac{dt}{s(\gamma(t))}$$

PDE $\mathcal{H}(\nabla u) = s$, posed on the lifted domain $\Omega = U \times \mathbb{S}^1$, with points $p = (x, \theta)$. Metric, with $n(\theta) := (\cos \theta, \sin \theta)$

$$\mathcal{F}_{(x,\theta)}(\dot{x}, \dot{\theta}) = \begin{cases} \|\dot{x}\| \mathcal{C}(\dot{\theta}/\|\dot{x}\|) & \text{if } \dot{x} = \pm \|\dot{x}\| n(\theta) \\ +\infty & \text{otherwise.} \end{cases}$$

We consider three curvature costs.

- ▶ Reeds-Shepp model $\mathcal{C}(\kappa) := \sqrt{1 + \kappa^2}$ (with rev. gear)

$$\mathcal{H}_{(x,\theta)}(\hat{x}, \hat{\theta}) = \langle \hat{x}, n(\theta) \rangle^2 + \hat{\theta}^2$$

- ▶ Euler elastica model $\mathcal{C}(\kappa) := 1 + \kappa^2$
- ▶ Dubins model $\mathcal{C}(\kappa) := 1$ if $\kappa \leq 1$, and $+\infty$ otherwise.

Curvature penalized shortest paths

Define the cost of a unit speed curve $\gamma : [0, T] \rightarrow U$, with curvature κ , as

$$\int_0^T \mathcal{C}(\kappa(t)) \frac{dt}{s(\gamma(t))}$$

PDE $\mathcal{H}(\nabla u) = s$, posed on the lifted domain $\Omega = U \times \mathbb{S}^1$, with points $p = (x, \theta)$. Metric, with $n(\theta) := (\cos \theta, \sin \theta)$

$$\mathcal{F}_{(x,\theta)}(\dot{x}, \dot{\theta}) = \begin{cases} \|\dot{x}\| \mathcal{C}(\dot{\theta}/\|\dot{x}\|) & \text{if } \dot{x} = \|\dot{x}\| n(\theta) \\ +\infty & \text{otherwise.} \end{cases}$$

We consider three curvature costs.

- ▶ Reeds-Shepp model $\mathcal{C}(\kappa) := \sqrt{1 + \kappa^2}$
- ▶ Euler elastica model $\mathcal{C}(\kappa) := 1 + \kappa^2$

$$\mathcal{H}_{(x,\theta)}(\hat{x}, \hat{\theta}) = \frac{1}{4} \left(\langle \hat{x}, n(\theta) \rangle + \sqrt{\langle \hat{x}, n(\theta) \rangle^2 + \hat{\theta}^2} \right)^2$$

- ▶ Dubins model $\mathcal{C}(\kappa) := 1$ if $\kappa \leq 1$, and $+\infty$ otherwise.

Curvature penalized shortest paths

Define the cost of a unit speed curve $\gamma : [0, T] \rightarrow U$, with curvature κ , as

$$\int_0^T \mathcal{C}(\kappa(t)) \frac{dt}{s(\gamma(t))}$$

PDE $\mathcal{H}(\nabla u) = s$, posed on the lifted domain $\Omega = U \times \mathbb{S}^1$, with points $p = (x, \theta)$. Metric, with $n(\theta) := (\cos \theta, \sin \theta)$

$$\mathcal{F}_{(x,\theta)}(\dot{x}, \dot{\theta}) = \begin{cases} \|\dot{x}\| \mathcal{C}(\dot{\theta}/\|\dot{x}\|) & \text{if } \dot{x} = \|\dot{x}\| n(\theta) \\ +\infty & \text{otherwise.} \end{cases}$$

We consider three curvature costs.

- ▶ Reeds-Shepp model $\mathcal{C}(\kappa) := \sqrt{1 + \kappa^2}$
- ▶ Euler elastica model $\mathcal{C}(\kappa) := 1 + \kappa^2$
- ▶ Dubins model $\mathcal{C}(\kappa) := 1$ if $\kappa \leq 1$, and $+\infty$ otherwise.

$$\mathcal{H}_{(x,\theta)}(\hat{x}, \hat{\theta}) = \max\{0, \langle \hat{x}, n(\theta) \rangle + |\hat{\theta}|\}^2$$

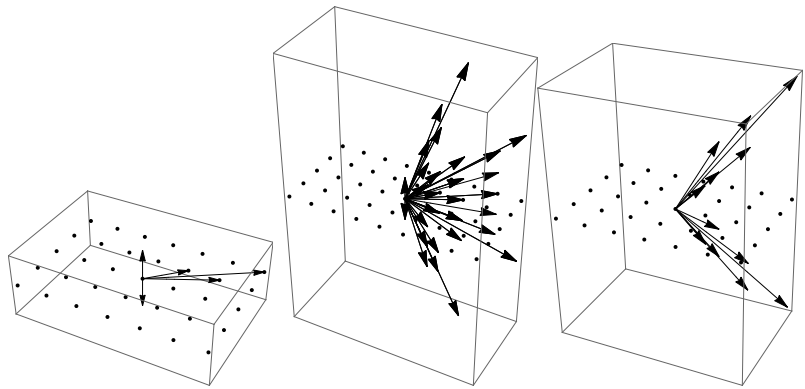
Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm



Reeds-Shepp

Elastica

Dubins

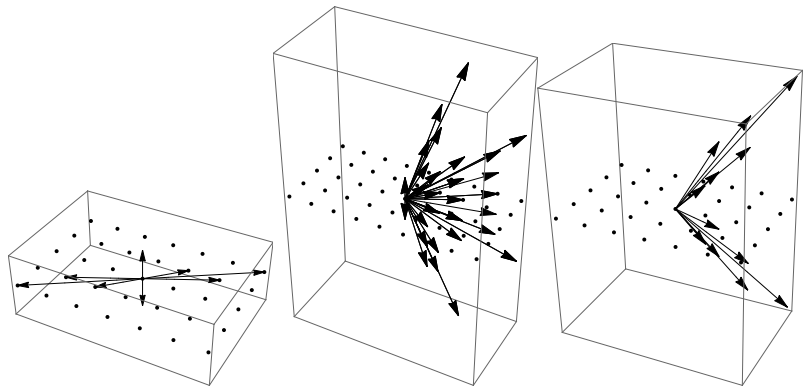
Anisotropic
Fast-
Marching

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm



Reeds-Shepp (rev. gear)

Elastica

Dubins

Anisotropic
Fast-
Marching

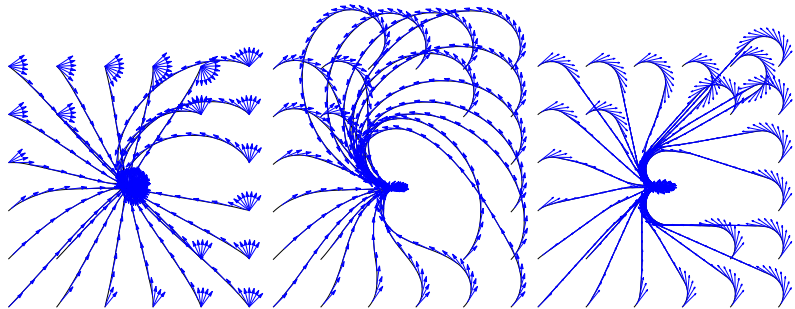
Qualitative features of the models

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm



Reeds-Shepp

Elastica

Dubins

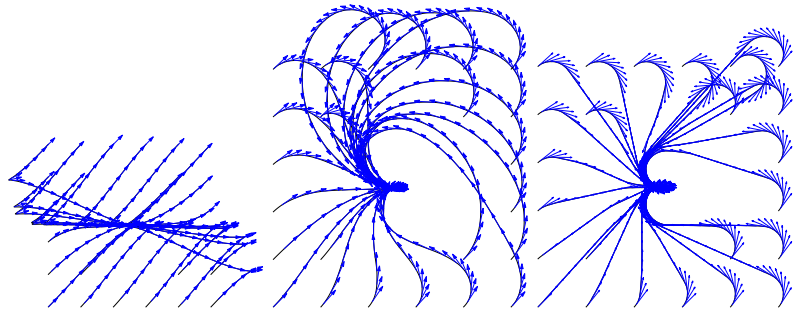
- ▶ Reeds-Shepp's car can rotate in place (w.o. rev gear)
- ▶ Euler's car optimal paths are smooth.
- ▶ Dubin's car has a turning radius of 1.

Qualitative features of the models

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm



Reeds-Shepp (rev. gear)

Elastica

Dubins

- ▶ Reeds-Shepp's car can rotate in place (w.o. rev gear), or do cusps (with rev gear).
- ▶ Euler's car optimal paths are smooth.
- ▶ Dubin's car has a turning radius of 1.

Conclusion: Hamiltonian approach

Jean-Marie
Mirebeau

What exactly
can solve the
Fast-
Marching
Algorithm ?

The semi-
Lagrangian
paradigm

The
Hamiltonian
paradigm

Pros:

- ▶ Applies to a variety of metrics.
- ▶ Easy to implement.
- ▶ Cheap numerically
(Main cost is maintaining the priority queue of FM)

Cons:

- ▶ Hard to adapt to unstructured grids.