

Estimations de surfaces moyennes via les métamorphose de formes fonctionnelles

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avec N. Charon (JHU, Baltimore) et A. Trouvé (ENS Cachan)

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Motivations

Propose a well-grounded mathematical framework to estimate and analyze variability of geometric-functional data coming from medical imaging.

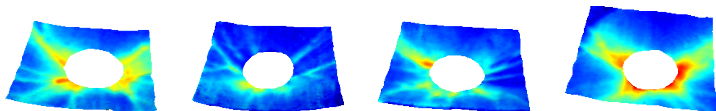
- ▶ Data: for any $i = 1, \dots, N$ the observation Y^i is in a Hilbert space \mathcal{H} ,
- ▶ Deformations: γ^i belongs to a group \mathcal{G} acting on H (i.e. $\gamma^i \cdot Y^i \in \mathcal{H}$).
- ▶ Model:

$$Y^i = \gamma_i^* \cdot X + \varepsilon^i$$

where X is an unknown mean template and $\gamma_i^* \in \mathcal{G}$ are unknown deformations and ε^i is an error term in \mathcal{H} .

1 An overview of functional shapes:

Functional shapes



Definition (functional shape, a.k.a. fshape)

A couple (X, f) where :

- ▶ X is a compact smooth surface of \mathbb{R}^3
- ▶ $f : X \rightarrow \mathbb{R}$ a function on X , with $f \in L^2(X)$, i.e :

$$\int_X f^2(x) d\mathcal{H}^d(x) < \infty$$

where \mathcal{H}^d is the d -dimensional Hausdorff (or volume) measure.

Metamorphosis : Geometrical-functional transformations on fshape

- ▶ **Geometrical deformations** : Action of a diffeomorphism $\phi : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ on a fshape (X, f)

$$\phi \cdot (X, f) = (\phi(X), f \circ \phi^{-1})$$

(i.e transport of the support X preserving the values of the signal.)

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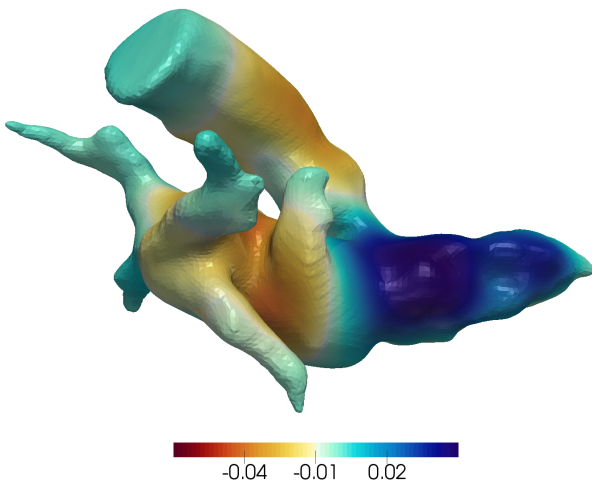
- ▶ **Functional deformations** : For a fshape $(X, f) \in \mathcal{F}$ and $(\phi, \zeta) \in G \times L^2(X)$:

$$(\phi, \zeta) \cdot (X, f) = (\phi(X), (f + \zeta) \circ \phi^{-1})$$

(i.e photometric variations of the signal f itself.)

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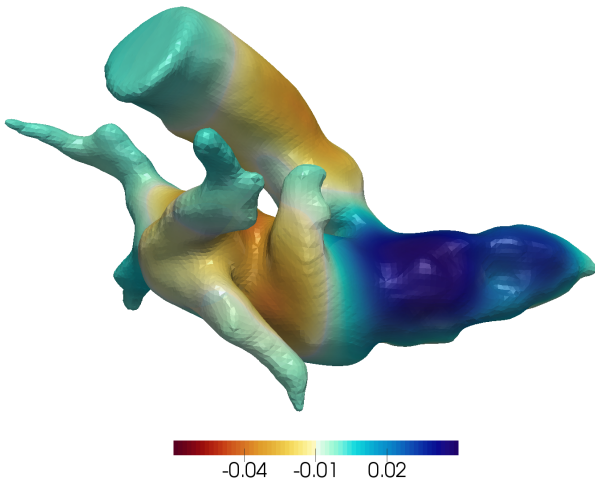
Metamorphosis: an example



Data courtesy of C. Chnafa, S. Mendez, F. Nicoud (Université de Montpellier)

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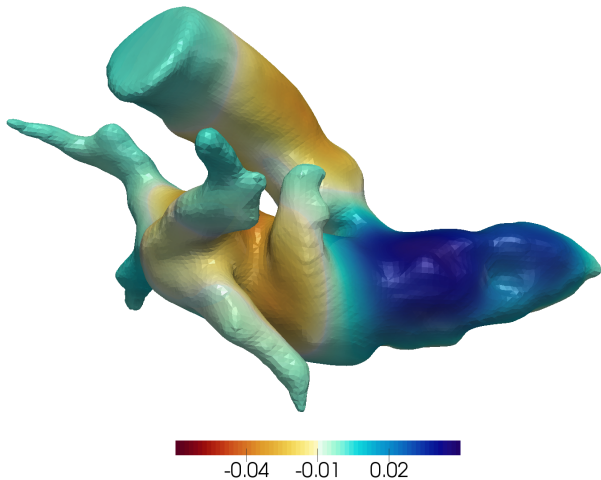
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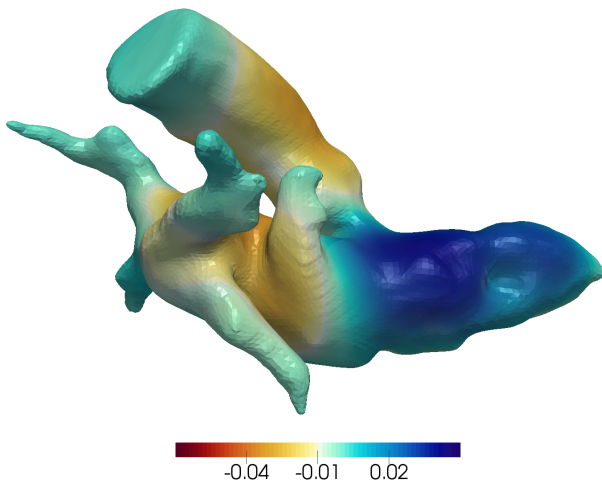
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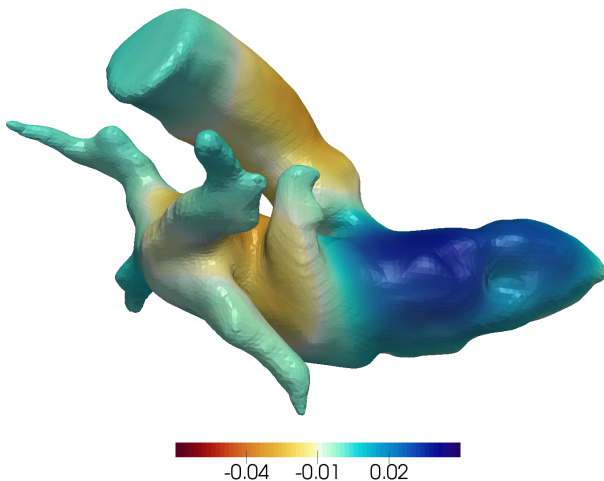
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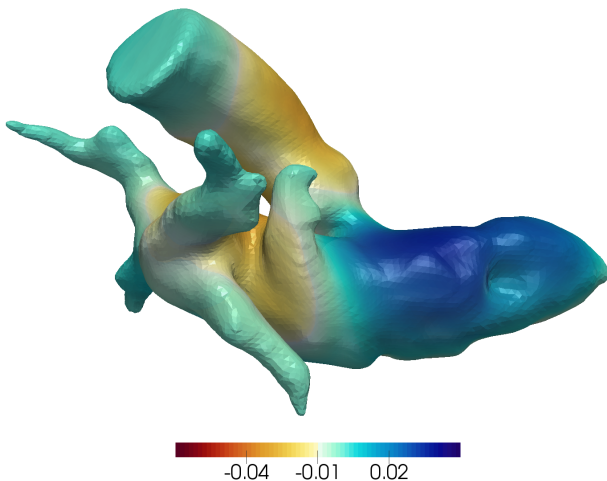
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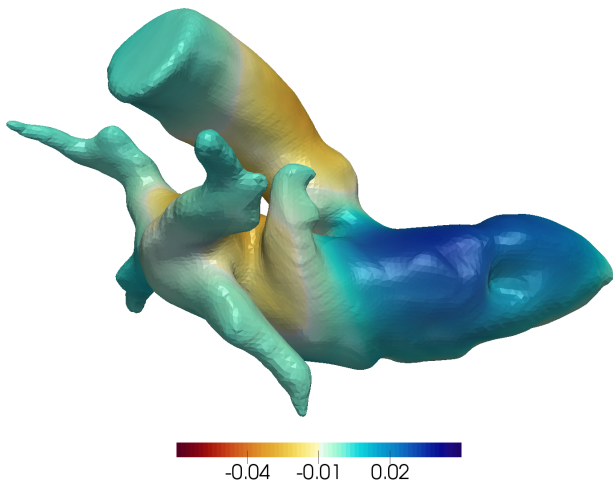
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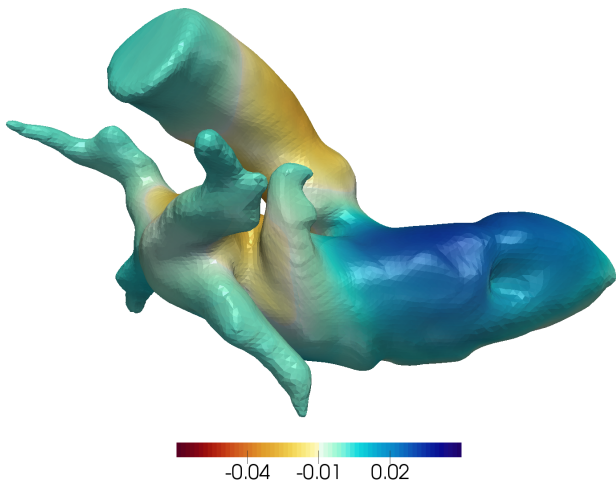
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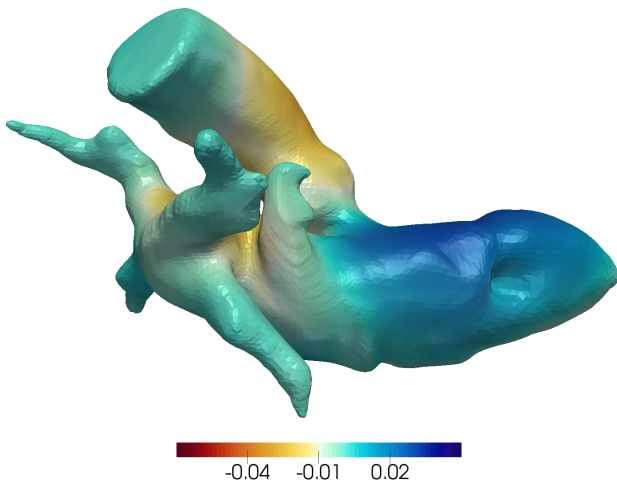
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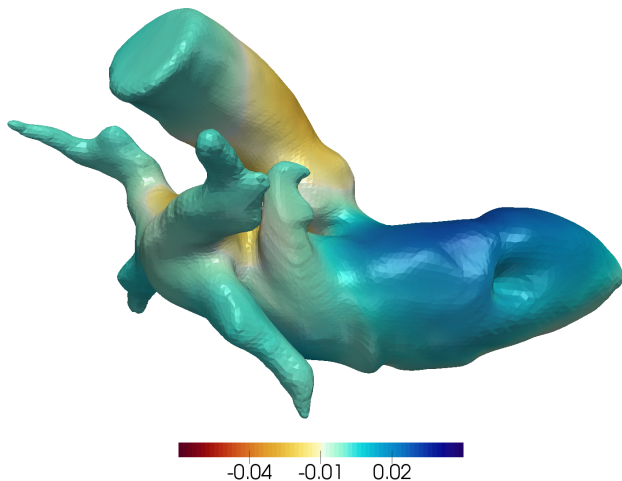
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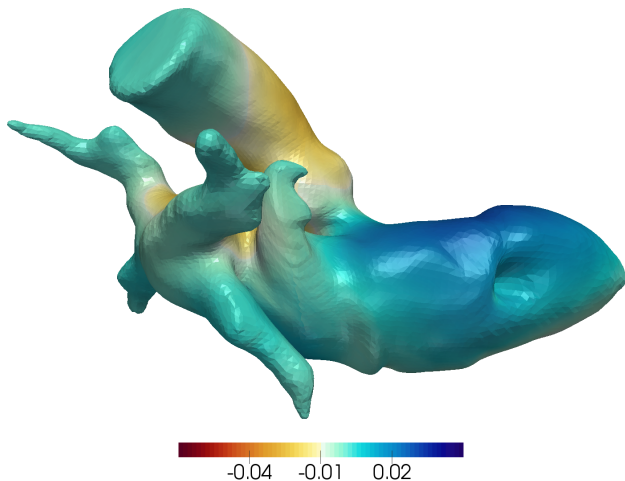
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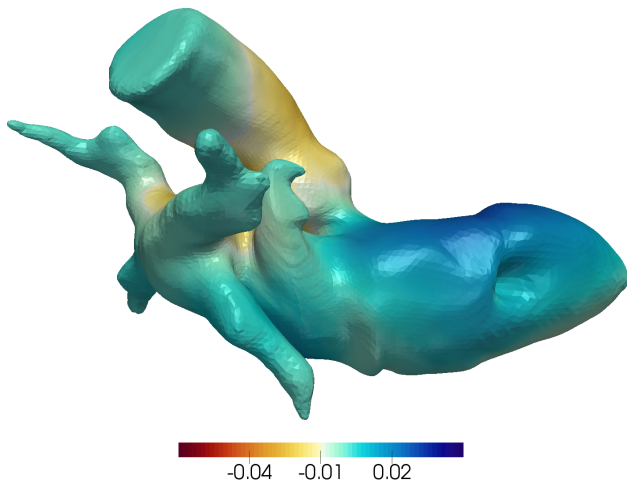
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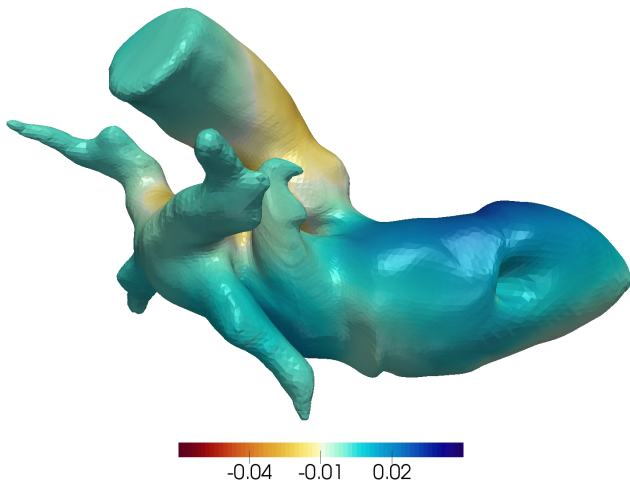
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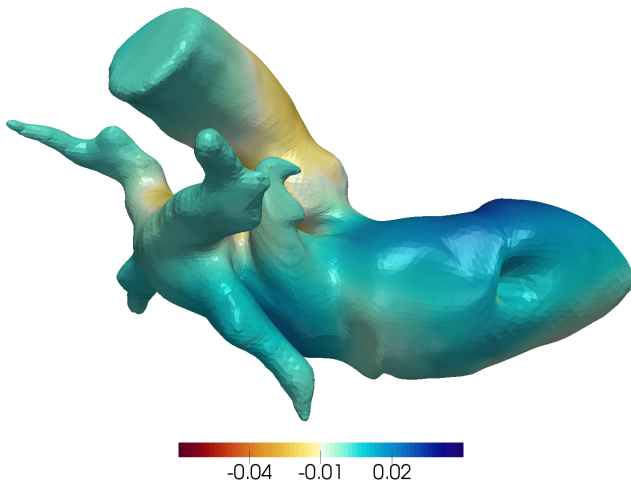
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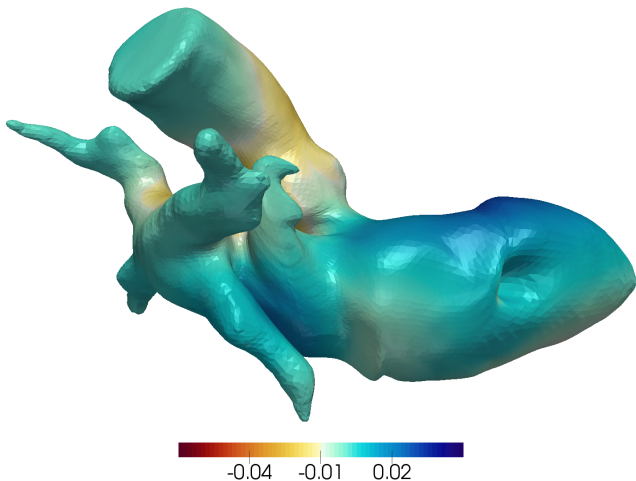
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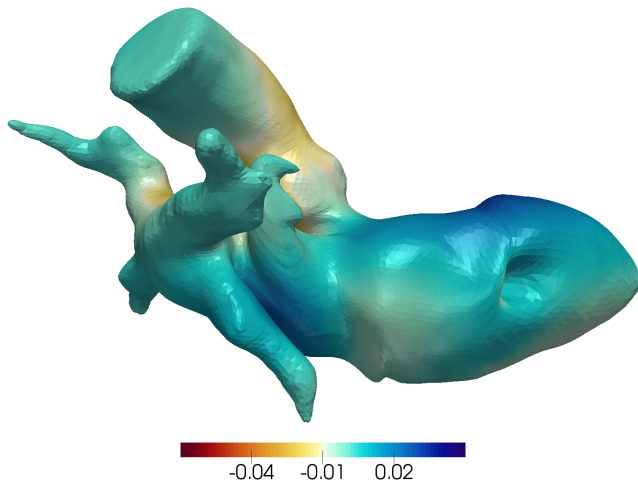
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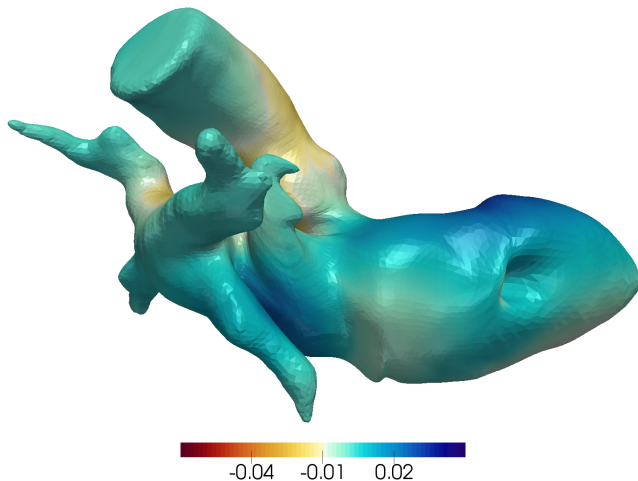
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Atlas estimation for populations of fshapes

Consider a population of N fshapes $(X^i, f^i)_{i=1, \dots, N}$.



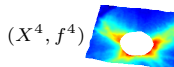
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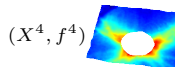


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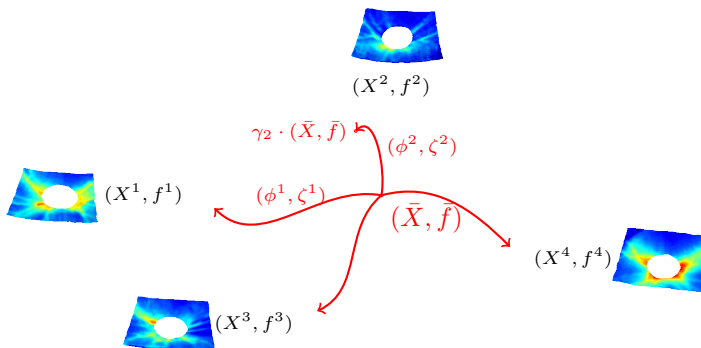
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Atlas:

- ▶ a template fshape (\bar{X}, \bar{f}) that captures the average features of the population,
- ▶ N transformations $(\phi^i, \zeta^i)_{i=1, \dots, N}$ mapping the template to each subject (*i.e.* $(\phi^i, \zeta^i) \cdot (\bar{X}, \bar{f}) \approx (X^i, f^i)$).

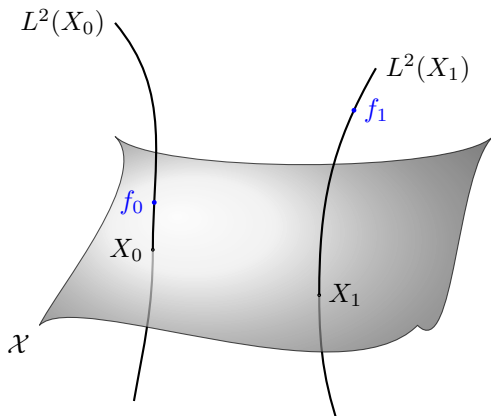
2 Metamorphosis of fshapes:

Fshape bundle

Let $G \subset \text{Diff}(\mathbb{R}^3)$ be a group of deformations. Let X_0 be a fixed submanifold of \mathbb{R}^3 and $\mathcal{X} = G \cdot X_0$ its orbit. We consider the space :

$$\mathcal{F} = \{ (X, f) \mid X \in \mathcal{X} \text{ and } f \in L^2(X) \}$$

which is a vector bundle of fiber $L^2(X_0)$.



2 Metamorphosis of fshapes: Generating Fshapes metamorphoses

Geometrical deformations : Large Deformation Diffeomorphic Metric Mapping

Deformation = flow of smooth vectors field.

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- ▶ **Flow** : let $v = (v_t)_{t \in [0,1]} \in V$ be a time dependant vectors field of \mathbb{R}^3 . Let $\phi : [0, 1] \times \mathbb{R}^3 \rightarrow \mathbb{R}^3$:

$$\begin{cases} \dot{\phi}_t(x) = v_t(\phi_t(x)) \\ \phi_0(x) = x. \end{cases} \quad t \in [0, 1] \text{ and } x \in \mathbb{R}^3$$

The space V should contain smooth vectors fields vanishing at infinity.

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- ▶ **Group action** : Let $L_V^2 \doteq L^2([0, 1], V)$. For all $v \in L_V^2$, $\phi_1^v(\cdot)$ is a C^1 -diffeomorphism of \mathbb{R}^3 . The set

$$G_V = \{\phi_1^v : \mathbb{R}^3 \rightarrow \mathbb{R}^3, v \in L_V^2\}$$

is a group endowed with the distance

$$d^2(\text{Id}, \phi) = \inf\{\|v\|_{L_V^2}^2 \doteq \int_0^1 \|v_t\|_V^2 dt, \dot{\phi} = v \circ \phi, \phi_1 = \phi\}$$

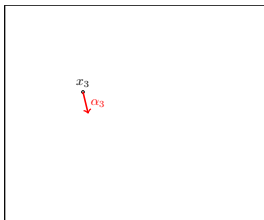
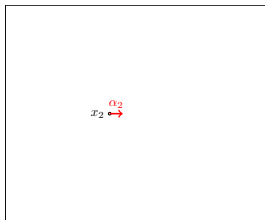
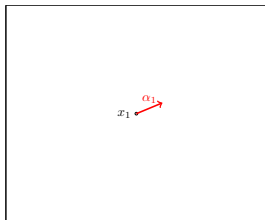
RKHS of vectors fields

- ▶ **Space of vectors fields** V : RKHS of vectors fields (smooth, vanishing at infinity). There exists a kernel $K_V : \mathbb{R}^3 \times \mathbb{R}^3 \rightarrow \mathbb{R}^{3 \times 3}$ such that

$$\text{Span}\{\delta_x^\alpha = K_V(x, \cdot)\alpha, x \in \mathbb{R}^3, \alpha \in \mathbb{R}^3\}$$

is dense in V . In practice :

$$K_V(x, y) = e^{-\frac{\|x-y\|^2}{\sigma_V^2}} Id_n.$$



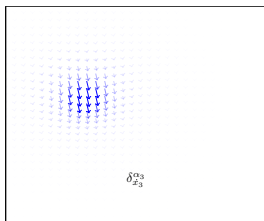
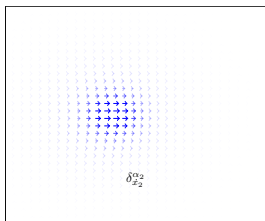
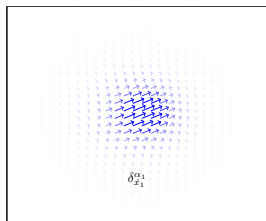
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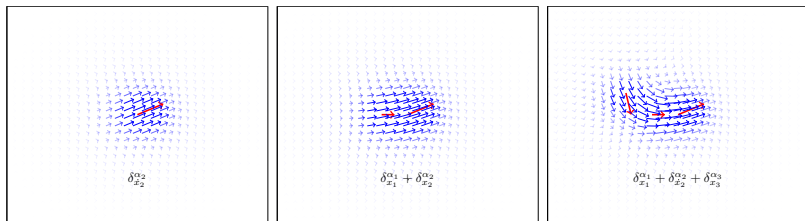
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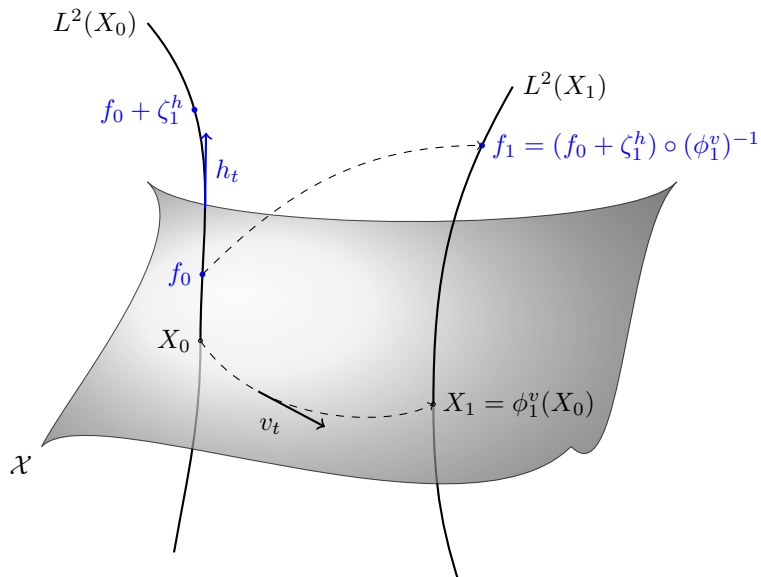
Functional deformations

- ▶ **Functional deformations** : For a given fshape $(X, f) \in \mathcal{F}$, we consider evolutions of signal functions given by

$$f_t = f + \zeta_t^h \in L^2(X)$$

for $h \in L^2([0, 1], L^2(X))$ and $\zeta_t^h(x) \doteq \int_0^t h_s(x) ds$.

Fshape metamorphoses



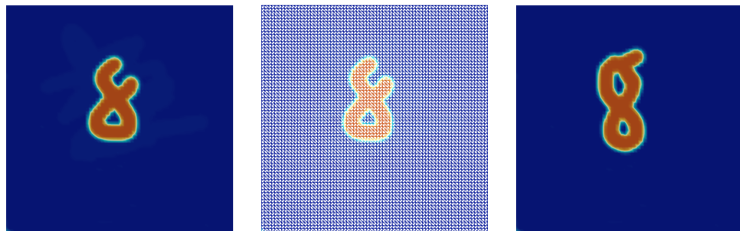
Fshape metamorphoses and metrics

- ▶ We define the energy of a transformation path (v, h) by :

$$E_X(v, h) \doteq \frac{\gamma_V}{2} \int_0^1 |v_t|_V^2 + \frac{\gamma_f}{2} \int_0^1 \int_{X_t} |h_t \circ (\phi_t^v)^{-1}|^2(x) d\mathcal{H}^d(x)$$

where $X_t = \phi_t^v(X)$ and $\gamma_V, \gamma_f > 0$.

- ▶ Find minimum energy transformation path:



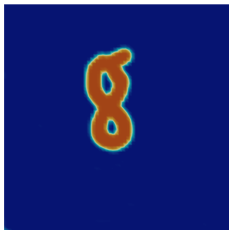
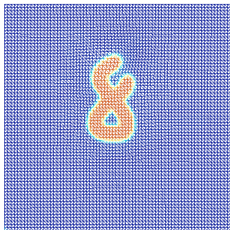
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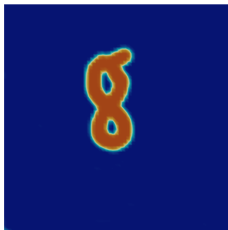
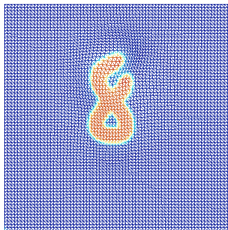
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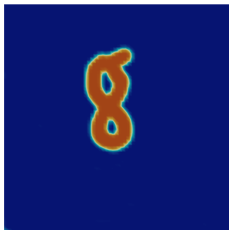
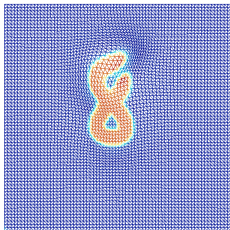
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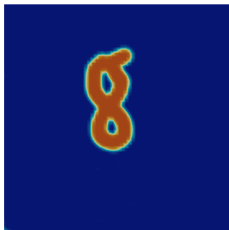
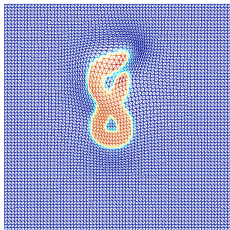
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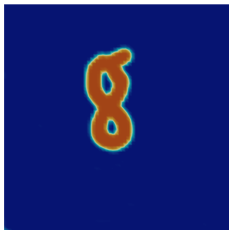
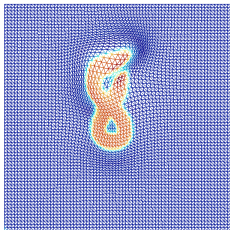
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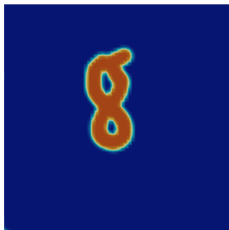
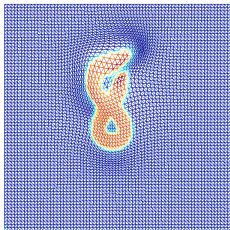
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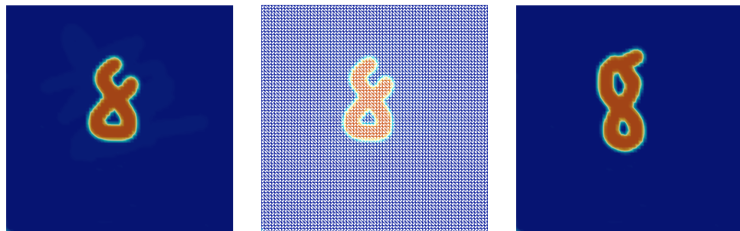
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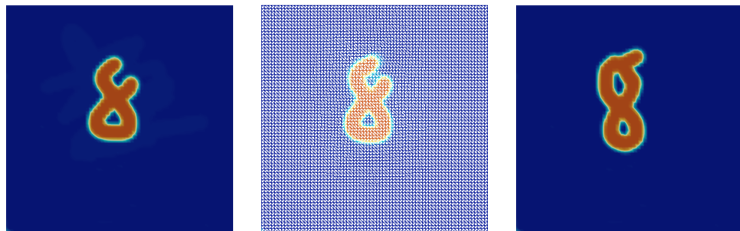
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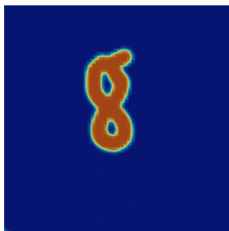
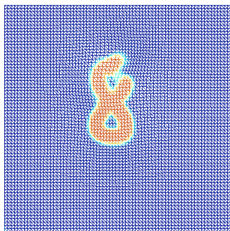
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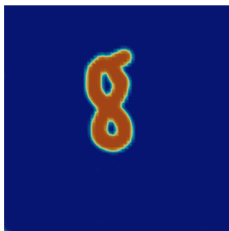
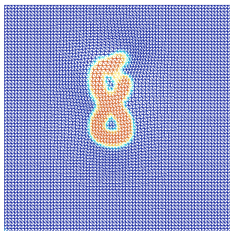
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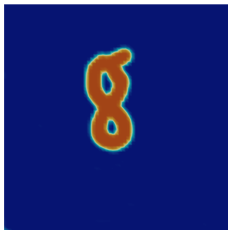
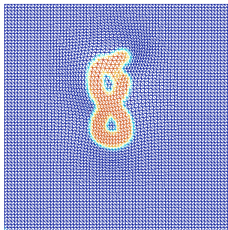
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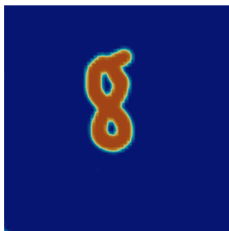
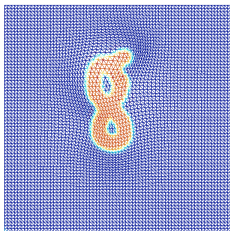
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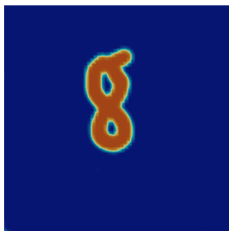
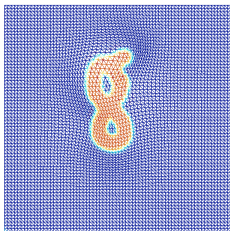
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- ▶ Find minimum energy transformation path:



Some theoretical results

- ▶ $E_X(v, h)$ induces the following Riemannian distance on \mathcal{F} :

$$\left\{ \begin{array}{l} d_{\mathcal{F}}((X, f), (X', f')) \doteq \inf \left\{ E_X(v, h)^{\frac{1}{2}} \right\} \\ \text{with } X' = \phi_1^v(X), f' = (f + \zeta_1^h) \circ (\phi_1^v)^{-1} \end{array} \right.$$

- ▶ Existence of geodesics between (X, f) and (X', f') (both in \mathcal{F}).
- ▶ Existence of Kärcher means of $(X^i, f^i)_{i=1, \dots, n}$ (all in \mathcal{F}).

Open question : uniqueness geodesics

Definition of fvarifold

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A *fvarifold* on \mathbb{R}^3 is a distribution (or measure) on the space

$$\mathbb{R}^3 \times G_2(\mathbb{R}^3) \times \mathbb{R},$$

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- ▶ A Dirac $\delta_{(x, \vec{n}, f)}$ corresponds to a singular mass located at position $x \in \mathbb{R}^3$ in the direction of the subspace $\text{Vect}(\vec{n})$ carrying signal value f .
- ▶ To any non-oriented fshape (X, f) corresponds the fvarifold $\mu_{(X, f)}$ defined for all $\omega \in C_0^1(\mathbb{R}^3 \times G_2(\mathbb{R}^3) \times \mathbb{R})$:

$$\mu_{(X, f)}(\omega) = \int_X \omega(x, T_x X, f(x)) d\mathcal{H}^2(x) \approx \left(\sum_i r_i \delta_{(x_i, \vec{n}_i, f_i)} \right) (\omega)$$

Kernel based metrics

RKHS: Let W be the RKHS dense in $C_0^1(\mathbb{R}^3 \times G_2(\mathbb{R}^3) \times \mathbb{R})$ generated by a product kernel $k_e \otimes k_t \otimes k_s : (\mathbb{R}^3 \times G_2(\mathbb{R}^3) \times \mathbb{R})^2 \rightarrow \mathbb{R}$ induces a Hilbert space structure on the set of fshapes that writes:

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$$\begin{aligned} & \langle \mu_{(X,f)}, \mu_{(Y,g)} \rangle_W \\ &= \int_{X \times Y} k_e(x, y) k_t(T_x X, T_y Y) k_s(f(x), g(y)) d\mathcal{H}^2(x) d\mathcal{H}^2(y) \end{aligned}$$

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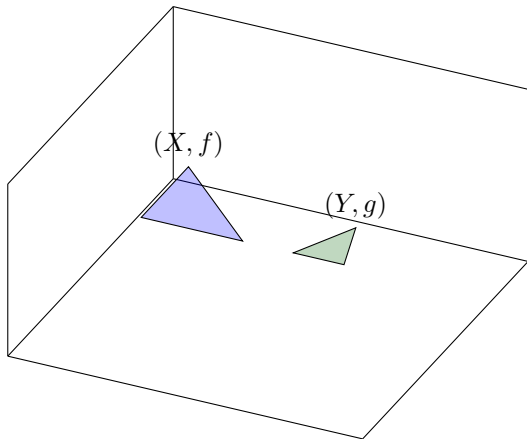
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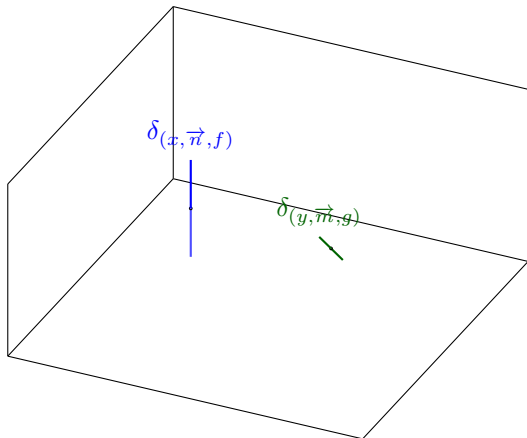
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Distance: $\|\mu_{(X,f)} - \mu_{(Y,g)}\|_{W'}^2 = \langle \mu_{(X,f)}, \mu_{(X,f)} \rangle_{W'} + \langle \mu_{(Y,g)}, \mu_{(Y,g)} \rangle_{W'} - 2 \langle \mu_{(X,f)}, \mu_{(Y,g)} \rangle_{W'}$.

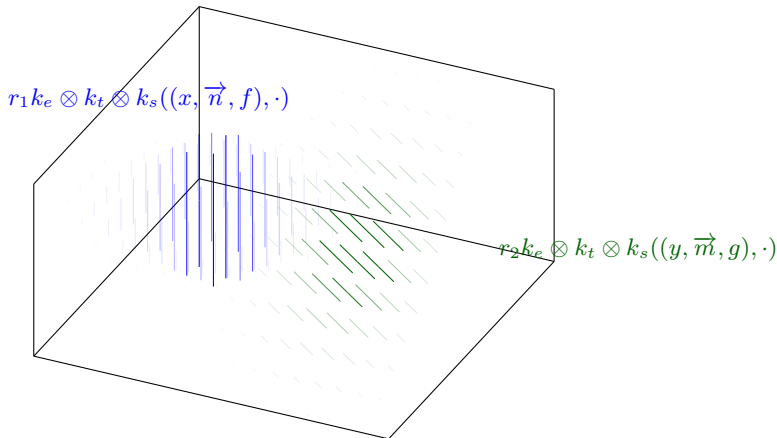
Discrete approximation



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$$\begin{aligned}
 & r_1 r_2 \langle k_e \otimes k_t \otimes k_s((x, \vec{n}, f), \cdot), k_e \otimes k_t \otimes k_s((y, \vec{m}, g), \cdot) \rangle \\
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lsc of varifold norm

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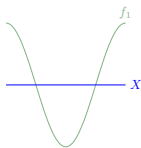
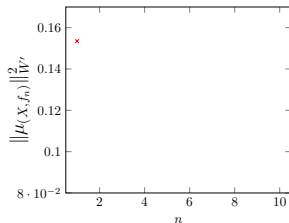
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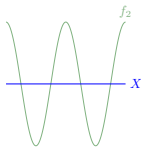
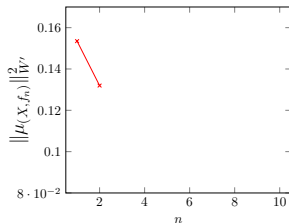
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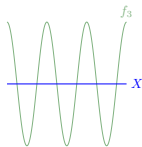
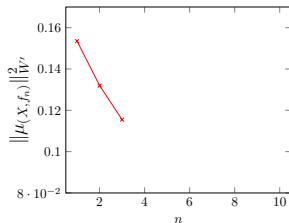
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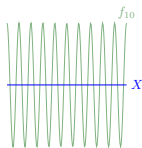
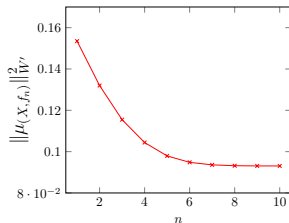
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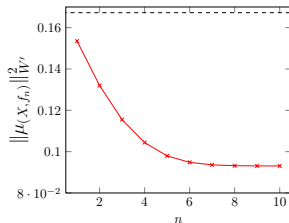
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_____ f_X

(X, f)



$\|\mu_{(X, f_n)}\|_{W'}^2$

Hyper Template

Mean = argmin ...



(X^1, f^1)



(X^2, f^2)



(X^3, f^3)



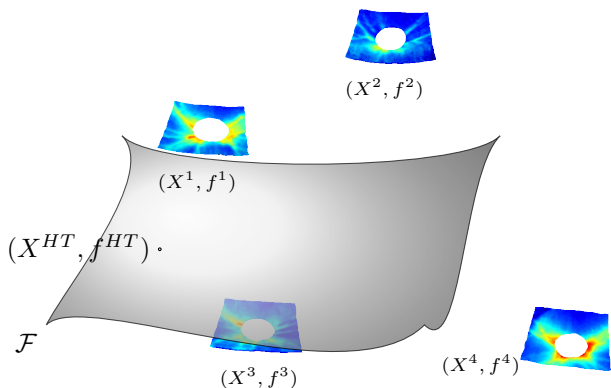
(X^4, f^4)

Space of solutions : Restrict the optimization to an 'orbit' of a certain **hypertemplate** fshape (X^{HT}, f^{HT}) .

\Rightarrow Optimization wrt $(v^0, h^0) \in L^2([0, 1], V_0) \times L^2([0, 1], X^{HT})$.

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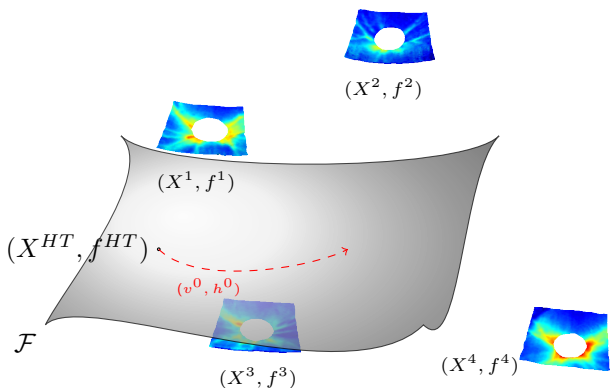


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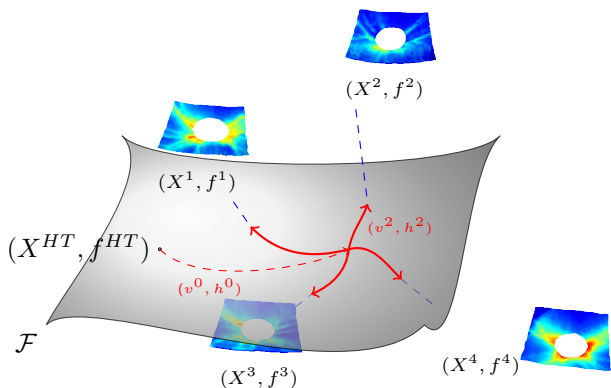


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Atlas Estimation

Variational formulation:

$$(v_*^0, h_*^0, (v_*^i, h_*^i)_{i=1, \dots, N})$$

$$= \underset{(v^0, h^0), (v^i, h^i)_{i=1, \dots, N}}{\operatorname{argmin}} E_{X^{\text{HT}}}(v^0, h^0) + \sum_{i=1}^N \left(E_X(v^i, h^i) + \gamma_W g((\tilde{X}^i, \tilde{f}^i), (X^i, f^i)) \right)$$

where g is the fvarifold squared distance between subject i and $(\tilde{X}^i, \tilde{f}^i) \doteq (\phi^{v^i}, \zeta^{h^i}) \cdot (X, f)$ and $(X, f) \doteq (\phi^{v^0}, \zeta^{h^0}) \cdot (X^{\text{HT}}, f^{\text{HT}})$.

Theorem (Existence of solutions)

Assume that W is continuously embedded in $C_0^2(E \times G_d(E) \times \mathbb{R})$. If $\gamma_{V_0}, \gamma_V > 0$ and γ_f/γ_W and γ_ζ/γ_W are large enough, then :

- ▶ the previous functional achieves its minimum on $v^0, \zeta^0, (v^i, \zeta^i)_{i=1, \dots, N}$

Merci de votre attention.