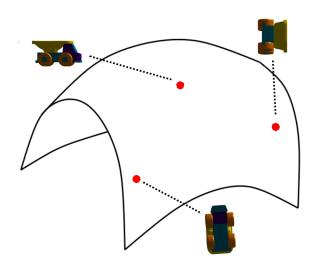
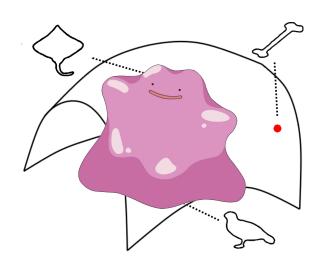
Wind field estimation via C^1 Bézier smoothing on manifolds

Pierre-Yves Gousenbourger, Estelle Massart pierre-yves.gousenbourger@uclouvain.be

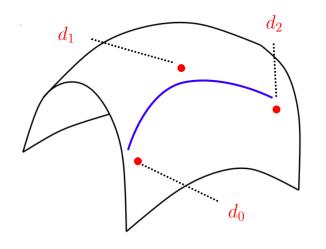
Université catholique de Louvain, ICTEAM

13 décembre 2018

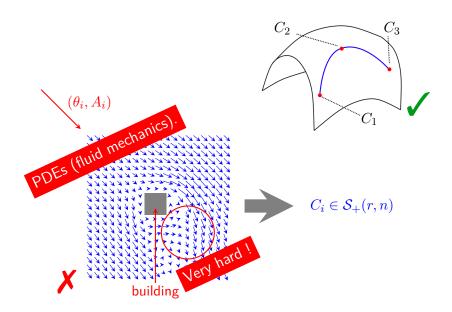


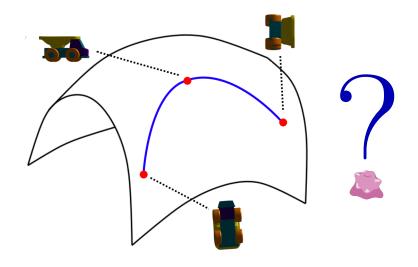






The wind field estimation

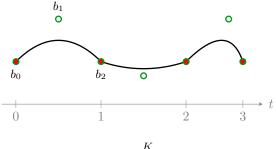




How to fit a curve to points on \mathcal{M} ?

1D : Interpolative Bézier curves

Each segment between two consecutive points is a **Bézier curve** of degree K.

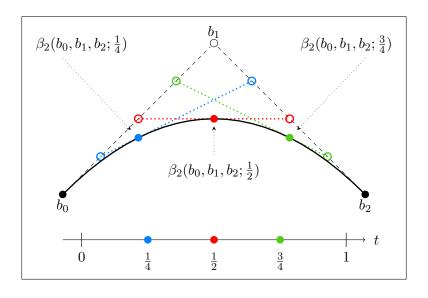


$$\beta_K(t, \mathbf{b}) = \sum_{i=0}^K b_i B_{iK}(t)$$

[G. et al. 2014, Arnould et al. 2015]

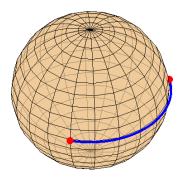
8

Reconstruction : the De Casteljau algorithm



9

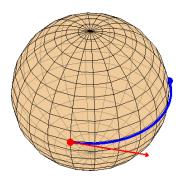
The straight line is a geodesic





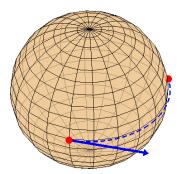
The exponential map to construct the geodesic

$$\gamma(t) = \operatorname{Exp}_{x}(t\xi_{x})$$



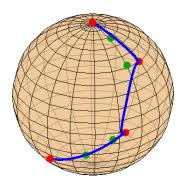
The logarithmic map to determine the starting velocity

$$\operatorname{Log}_{\boldsymbol{x}}(\boldsymbol{y}) = \boldsymbol{\xi}_{\boldsymbol{x}}$$





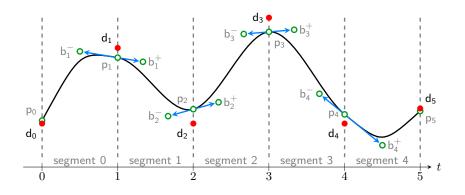
Example on the sphere



It's ugly. Make it **smooth**!

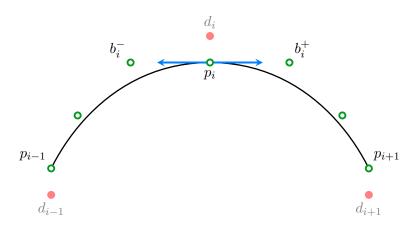
Well?! What about fitting, now?

Smooth fitting with Bézier (in \mathbb{R}^n)



Each segment is a Bézier curve smoothly connected! Unknowns : b_i^- , b_i^+ , p_i .

Differentiability



$$p_i = \frac{b_i^- + b_i^+}{2}$$

Optimal C^1 -piecewise Bézier fitting (in \mathbb{R}^n)

Minimization of the mean squared acceleration of the path

$$\min_{\substack{p_0, b_i^-, b_i^+, p_n \\ p_0, b_i^-, b_i^+, p_n \\ }} \int_0^1 \|\ddot{\beta}_2^0\|^2 \mathrm{d}t + \sum_{i=1}^{n-1} \int_0^1 \|\ddot{\beta}_3^i\|^2 \mathrm{d}t + \int_0^1 \|\ddot{\beta}_2^n\|^2 \mathrm{d}t + \lambda \sum_{i=0}^n \|\mathbf{d}_i - p_i\|_2^2$$

$$\min_{\substack{p_0, b_i^-, b_i^+, p_n \\ p_0, b_i^-, b_i^+, p_n \\ }} \int_0^1 \|\ddot{\beta}_2^0\|^2 \mathrm{d}t + \sum_{i=1}^{n-1} \int_0^1 \|\ddot{\beta}_3^i\|^2 \mathrm{d}t + \int_0^1 \|\ddot{\beta}_2^n\|^2 \mathrm{d}t + \lambda \sum_{i=0}^n \|\mathbf{d}_i - p_i\|_2^2$$
Second order polynomial $P(p_0, b_i^-, b_i^+, p_n, \lambda)$

$$\min_{\substack{p_0, b_i^-, b_i^+, p_n \\ }} \int_0^1 \|\ddot{\beta}_2^0\|^2 \mathrm{d}t + \sum_{i=1}^{n-1} \int_0^1 \|\ddot{\beta}_3^i\|^2 \mathrm{d}t + \int_0^1 \|\ddot{\beta}_2^n\|^2 \mathrm{d}t + \lambda \sum_{i=0}^n \|\mathbf{d}_i - p_i\|_2^2$$

Second order polynomial $P(p_0, b_i^-, b_i^+, p_n, \lambda)$

$$\nabla P(n_0, h_i^-, h_i^+, n_m)$$

Optimal C^1 -piecewise Bézier fitting (on \mathcal{M})

■ The control points are given by :

$$x_i = \sum_{j=0}^{n} q_{i,j}(\lambda) d_j$$

 \blacksquare These points are invariant under translation, *i.e.*:

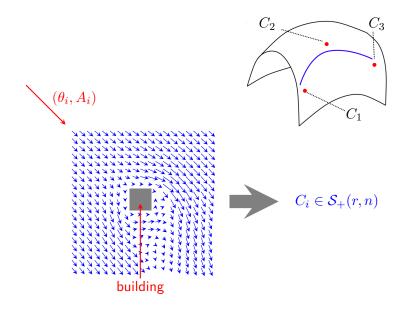
$$x_i - d^{ref} = \sum_{j=0}^n q_{i,j}(\lambda)(d_j - d^{ref})$$

■ On manifolds: projection to the **tangent space** of d^{ref} with the **Log**, as $a - b \Leftrightarrow \text{Log}_b(a)$

$$v_i = \operatorname{Log}_{d^{ref}}(x_i) = \sum_{j=0}^n q_{i,j}(\lambda) \operatorname{Log}_{d^{ref}}(d_j)$$

■ Back to the manifold with the **Exp**: $x_i = \text{Exp}_{d^{ref}}(v_i)$, where $d^{ref} = d_i$ if x_i is b_i^- , p_i , b_i^+ .

Application : Wind field estimation on $S_+(r, p)$.



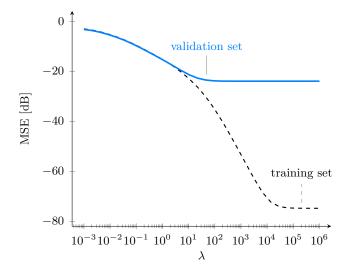
Application: Wind field estimation on $S_+(r,p)$.

How to estimate the error?

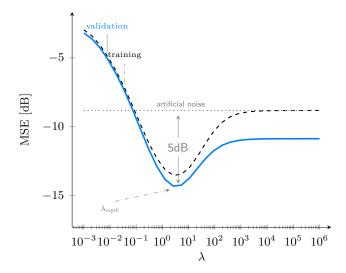
- Test set : $\{C(\theta_i)\}_{i \in I_T}$ $I_T = \{1, 3, ..., 33\};$ Validation set : $\{C(\theta_i)\}_{i \in I_V}$ $I_V = \{2, 4, ..., 32\};$
- Bézier spline $\mathbf{B}(\theta)$ with input data points from I_T
- Mean Squared Error:

$$MSE(\mathbf{B}(\theta)) = 10 \log \left(\frac{\sum_{i \in I_{\Omega}} ||C(\theta_i) - \mathbf{B}(\theta_i)||_F^2}{\sum_{i \in I_{\Omega}} ||C(\theta_i)||_F^2} \right), \quad \Omega = \{I, V\}.$$

Application: Wind field estimation on $S_+(r, p)$. No noise on data



Application: Wind field estimation on $S_+(r, p)$. With artificial noise (8dB) on data



Fitting with Bézier: pros and cons

- ✓ Optimality conditions are a closed form linear system.
 - \checkmark Method only needs exp and log maps.
 - ✓ The curve is C^1 .
 - \nearrow No guarantee on the optimality when \mathcal{M} is not flat.
 - ✓ We can do denoising.

Paper submitted at the ESANN conference, 2017. Joint work with MIT.

Conclusions

General C^1 -interpolative/fitting methods on manifolds... with applications in medical imaging, wind estimation, model reduction,...

light • closed form • uses few elements in \mathcal{M}

Summary on interpolation:

"Differentiable Piecewise-Bézier Surfaces on Riemannian Manifolds" [Absil, Gousenbourger, Striewski, Wirth, SIAM Journal on Imaging Sciences, to appear].

Any questions?



Wind field estimation via C^1 Bézier smoothing on manifolds

Pierre-Yves Gousenbourger, Estelle Massart pierre-yves.gousenbourger@uclouvain.be

Université catholique de Louvain, ICTEAM

13 décembre 2018