

## The problem

Find a  $C^1$  curve  $\mathbf{B} \in \mathcal{M}$ , fitting data-points  $d_i \in \mathcal{M}$  s.t.

$$\operatorname{argmin}_{\mathbf{B} \in \mathcal{C}^1} E_\lambda(\mathbf{B}) := \int_{t_0}^{t_n} \left\| \frac{D^2 \mathbf{B}(t)}{dt^2} \right\|_{\mathbf{B}(t)}^2 dt + \lambda \sum_{i=0}^n d^2(\mathbf{B}(t_i), d_i),$$

Bézier spline!

regularizer

data attachment

Motivated by:

- Denoising or resampling
- Medical applications
- Computer graphics

A Riemannian manifold  $\mathcal{M}$ .

A set of  $m+1$  manifold-valued

data points  $d_0, \dots, d_m \in \mathcal{M}$ .

A set of  $m+1$  time-parameters  $t_0, \dots, t_m \in [0, n]$ .

The Riemannian exponential "Exp"...

... and its inverse "Log".

And that's all!

## What it needs

## What it returns

A composite  $C^1$  curve made out of  $n$  pieces

$$\mathbf{B} : [0, n] \rightarrow \mathcal{M} : t \mapsto \mathbf{B}(t) = \beta_i(t - i), \text{ with } i = \lfloor t \rfloor \dots$$

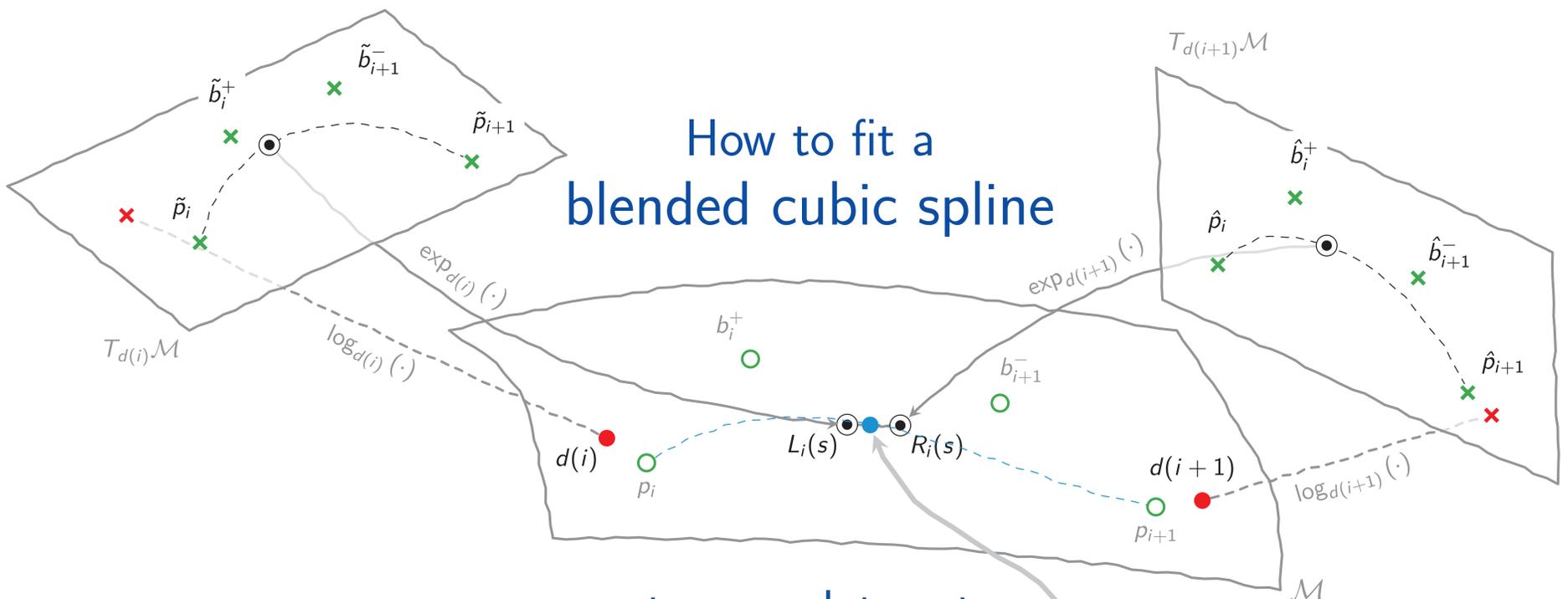
- (i) ... differentiable on  $[0, n]$ ,
- (ii) ... that interpolates the data points if  $m = n$  when  $\lambda \rightarrow \infty$ ,
- (iii) ... that is the natural cubic smoothing spline when  $\mathcal{M} = \mathbb{R}^r$ .

An efficient and easy-to-use method

- (iv) ... that only uses Exp's and Log's on  $\mathcal{M}$ ;
- (v) ... that only stores  $\mathcal{O}(n)$  tangent vectors;
- (vi) ... that needs  $\mathcal{O}(1)$  operations to compute  $\mathbf{B}(t)$  at a given  $t$ .

## How to fit a blended cubic spline

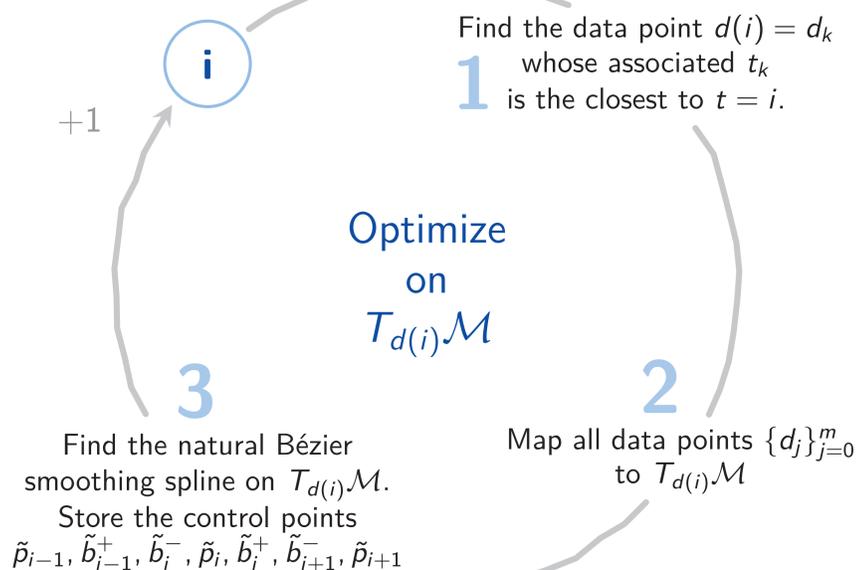
to your data set  
on a manifold?



$\mathbf{B}(t) = \beta_i(s) = \operatorname{av}[L_i(s), R_i(s)]_w$   
 is a weighted geodesic averaging  
 of  $L_i(s)$  and  $R_i(s)$   
 with a weight  $w(s) = 3s^2 - 2s^3$

## How it gets to it

## What it looks like



$\lambda = 100, m = 100, n = 4$

$\lambda = 10^8, m = n = 10$

