## Data fitting on manifolds with blended cubic splines

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We address the problem of curve fitting on a Riemannian manifold  $\mathcal{M}$ : given n + 1 data points  $d_0, \ldots, d_n \in \mathcal{M}$ , associated with real (time-)parameters  $t_0, \ldots, t_n$ , we seek a curve  $\gamma : [0, n] \to \mathcal{M}$  being, on the one hand, "sufficiently close" to the data points, while, on the other hand, being "sufficiently straight". A strategy to do so is to encapsulate the two above mentioned goals in an optimization problem

$$\min_{\gamma \in \Gamma} E_{\lambda}(\gamma) \coloneqq \int_{t_0}^{t_n} \left\| \frac{\mathrm{D}^2 \gamma(t)}{\mathrm{d}t^2} \right\|_{\gamma(t)}^2 \mathrm{d}t + \lambda \sum_{i=0}^n \mathrm{d}^2(\gamma(t_i), d_i),$$
(1)

where  $\Gamma$  is an admissible set of curves  $\gamma$  on  $\mathcal{M}$ ,  $\frac{D^2}{dt^2}$  is the (Levi-Civita) second covariant derivative,  $\|\cdot\|_{\gamma(t)}$  is the Riemannian metric at  $\gamma(t)$ , and  $d(\cdot, \cdot)$  is the Riemannian distance. The problem also has a parameter  $\lambda$  that strikes the balance between the two goals of the problem, i.e., the regularizer  $\int_{t_0}^{t_n} \|\frac{D^2\gamma(t)}{dt^2}\|_{\gamma(t)}^2 dt$  and the fitting term  $\sum_{i=0}^n d^2(\gamma(t_i), d_i)$ .

We present here a method that extends the work of (Arnould et al., 2015). In a nutshell, we reduce the search space of (1) to the space of  $C^1$  composite curves

$$\mathbf{B}: [0,n] \to \mathcal{M}: f_i(t-i), \ i = \lfloor t \rfloor,$$

made of so-called blended functions  $f_i$ . These blended functions are given by

$$f_i(t) = \operatorname{av}[(L_i(t), R_i(t)), (1 - w(t), w(t))],$$

where  $\operatorname{av}[(x, y), (1 - a, a)]$  is a weighted mean,  $w(t) = 3t^2 - 2t^3$ , and where  $R_i(t)$  and  $L_i(t)$  are cubic Bézier curves (Farin, 2002) computed respectively on  $T_{d_i}\mathcal{M}$  and  $T_{d_{i+1}}\mathcal{M}$ ,  $i = 0, \ldots, n - 1$ , with the control points optimized with a technique similar to (Arnould et al., 2015). The blending method is represented in Figure 1.

The method guarantees the five following properties: (i) the curve is  $C^1$  on  $[t_0, t_n]$ ; (ii) the curve interpolates the

data points  $d_0, \ldots, d_n$  when  $\lambda \to \infty$ ; (iii) the curve is the natural cubic spline minimizing (1) over a Sobolev space  $H^2(t_0, t_n)$  when  $\mathcal{M}$  is a Euclidean space; (iv) the method is designed for ease to use: it only requires the knowledge of the Riemannian exponential and the Riemannian logarithm on  $\mathcal{M}$ ; (v) the curve can be stored with only  $\mathcal{O}(n)$  tangent vectors; and, finally, (vi) given this representation, computing  $\gamma(t)$  at  $t \in [t_0, t_n]$  only requires  $\mathcal{O}(1)$  exp and log evaluations.

Further details will be available in (Gousenbourger et al., 2018).

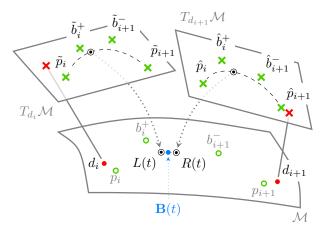


Figure 1. The composite curve  $\mathbf{B}(t)$  is made of cubic Euclidean Bézier curves computed on different tangent spaces, and then blended together with carefully chosen weights.

## References

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Proceedings of the 35<sup>th</sup> International Conference on Machine Learning, Stockholm, Sweden, PMLR 80, 2018. Copyright 2018 by the author(s).