

# Contractivity and convergence of refinement schemes in Riemannian geometry

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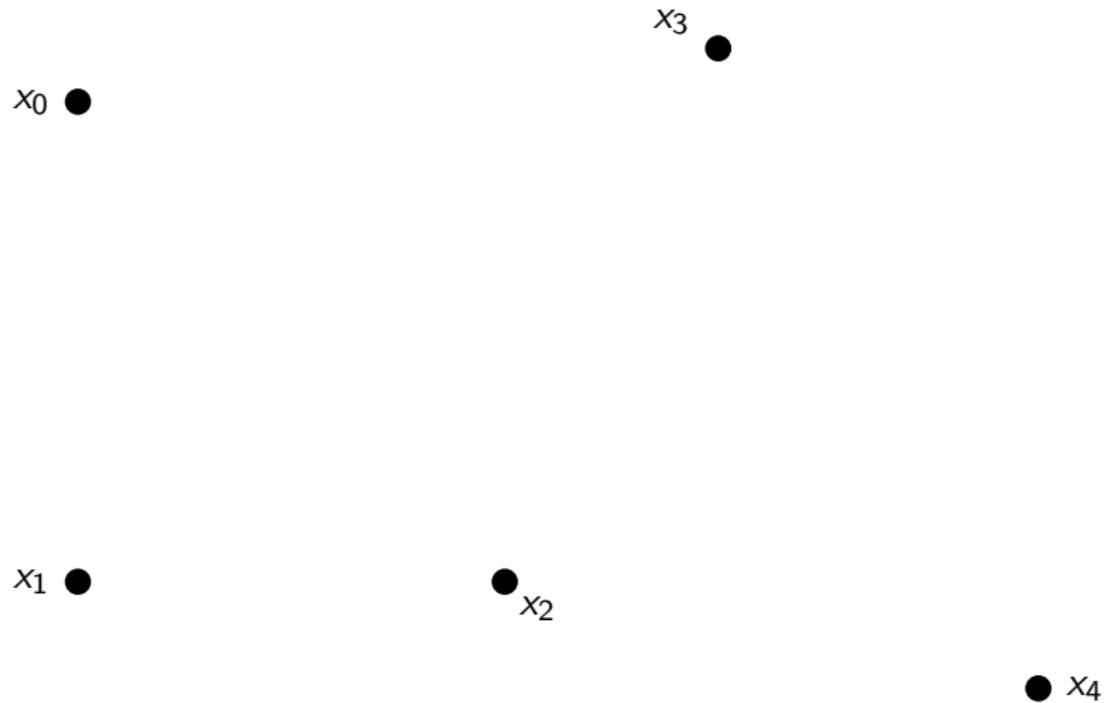


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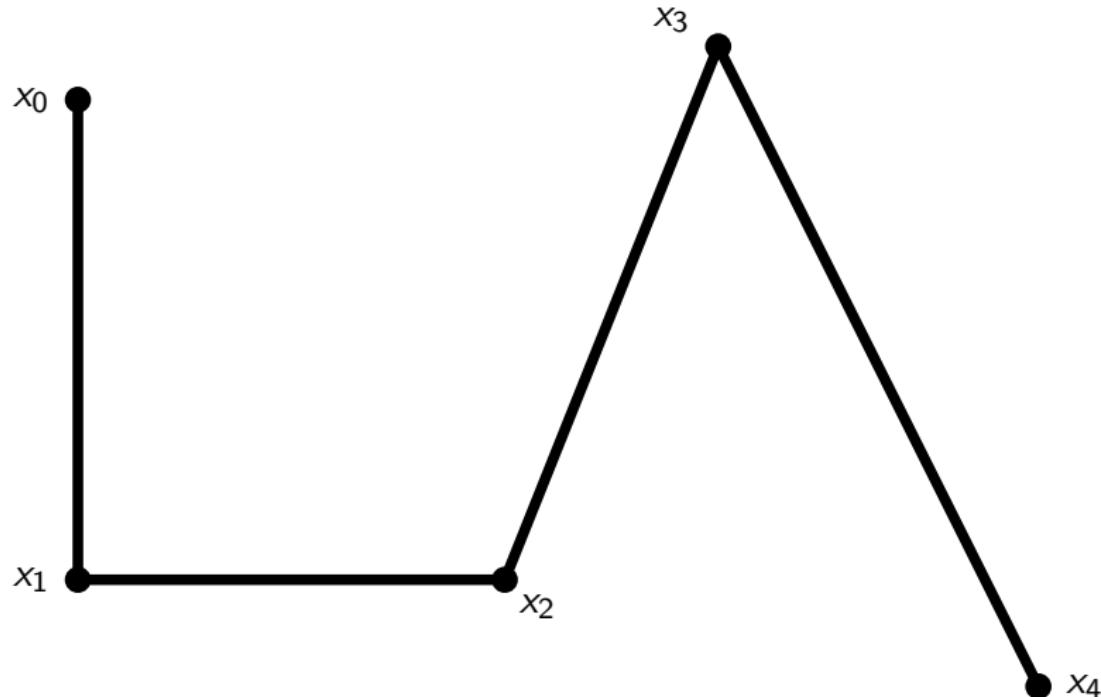
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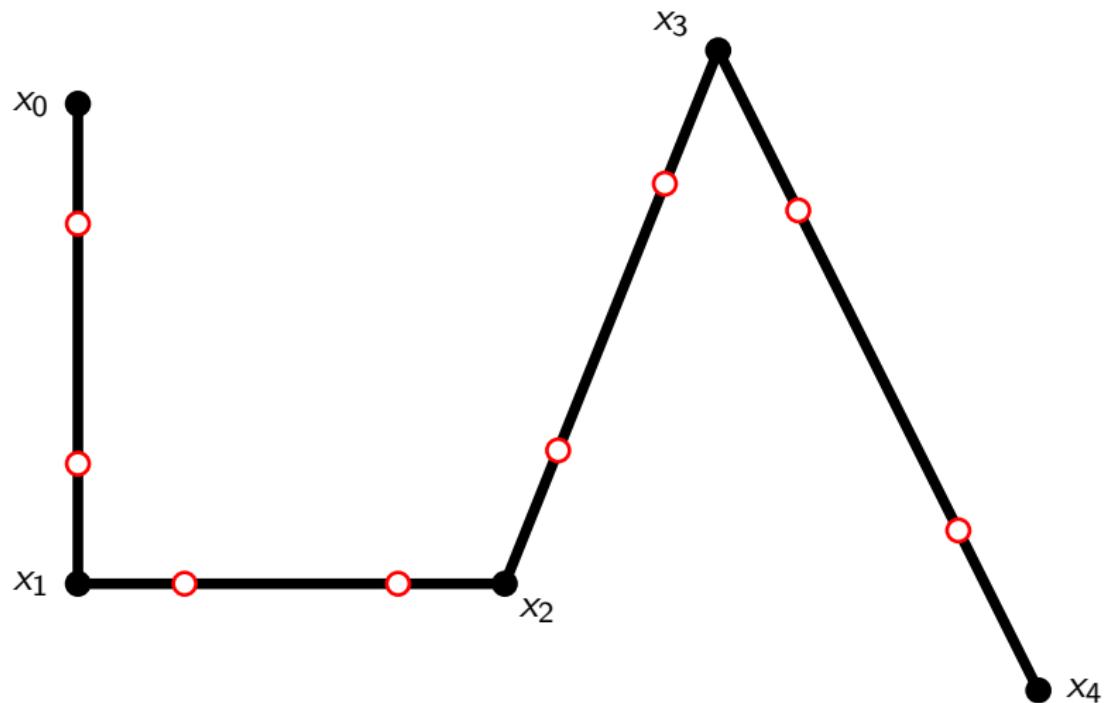
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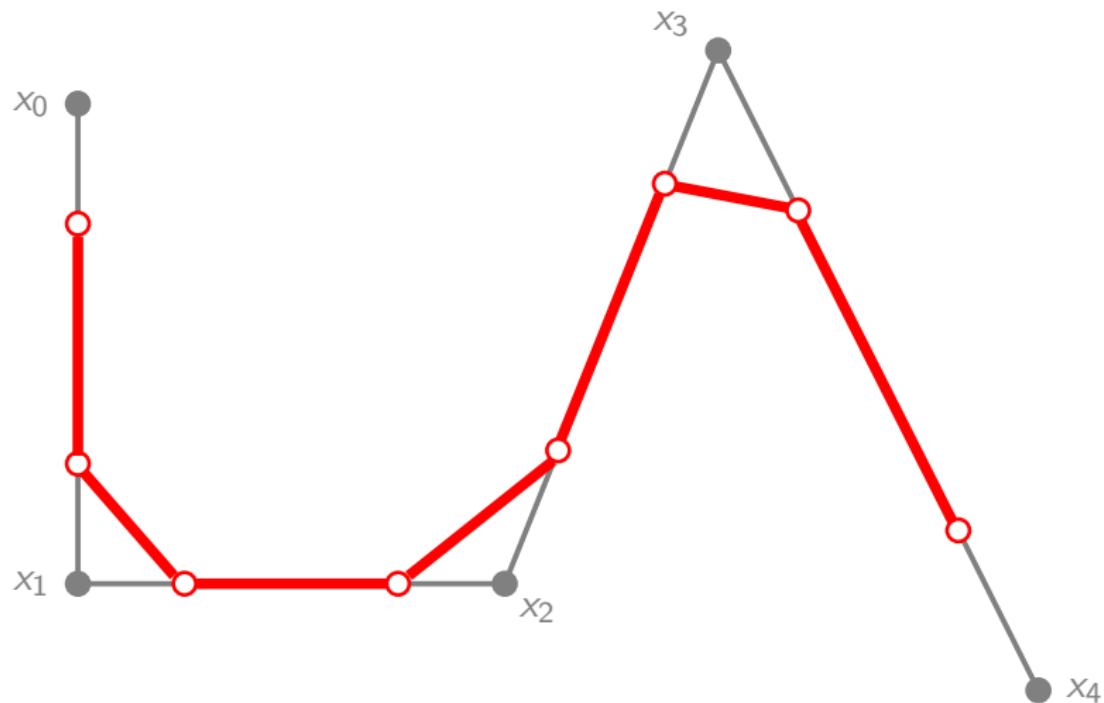
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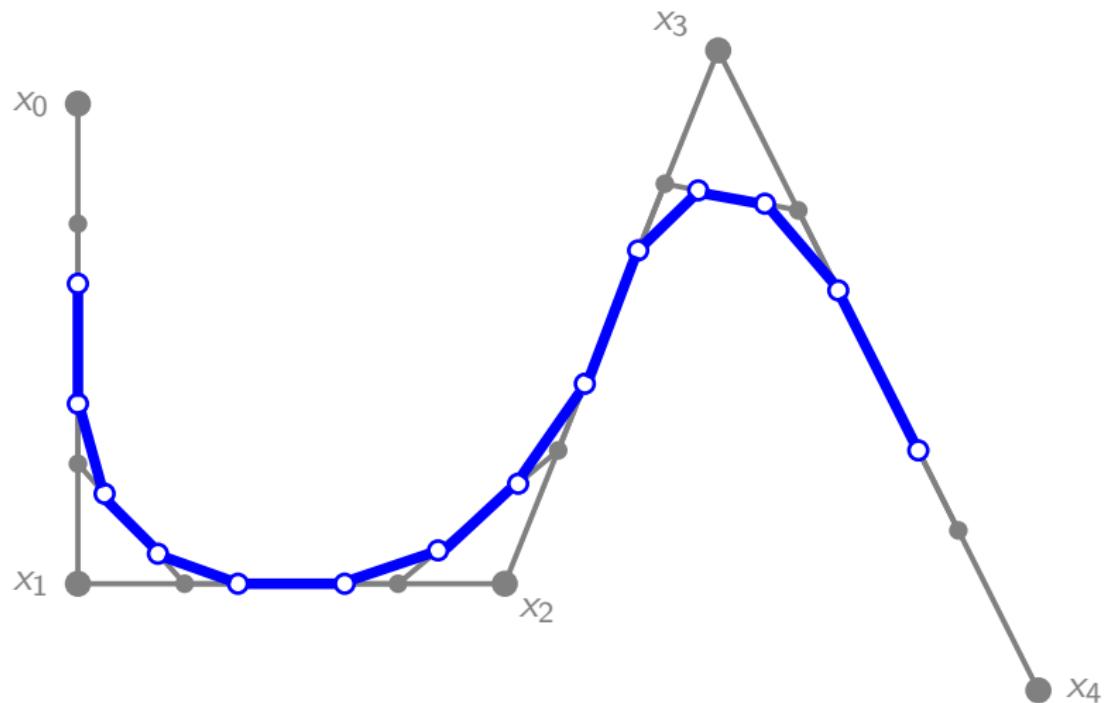
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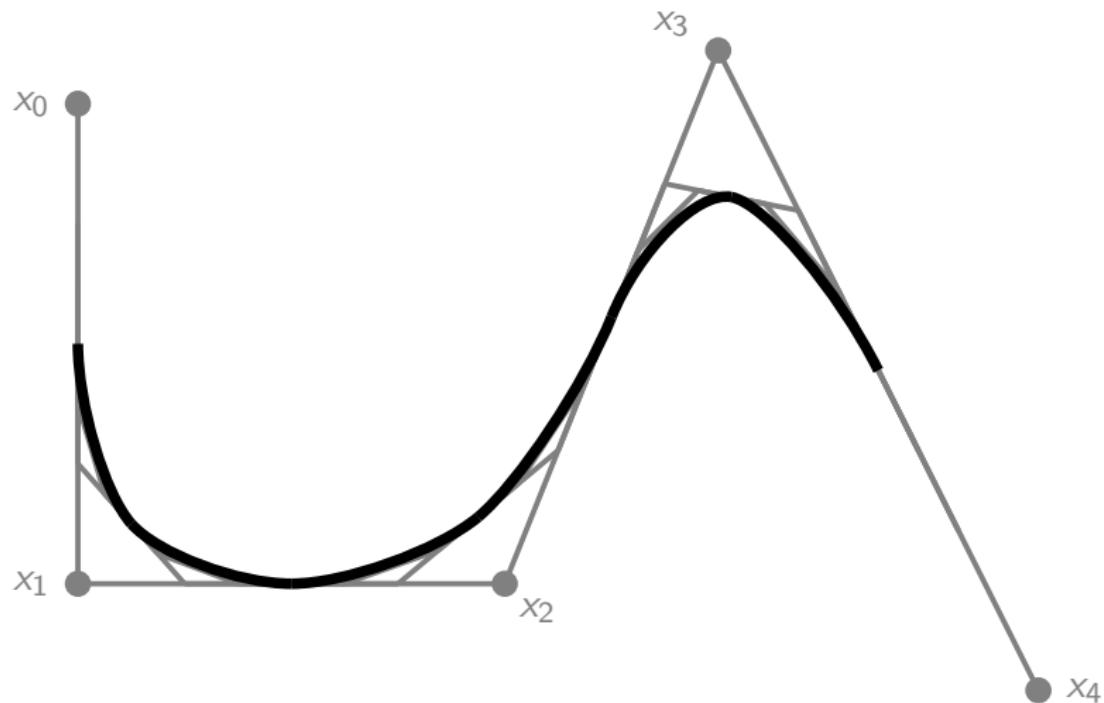
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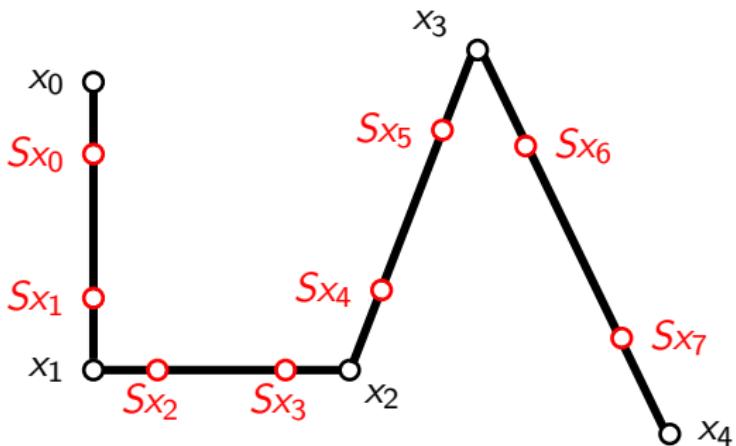


# What is a subdivision scheme?



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## Chaikin's algorithm



Subdivision scheme  $S$ :

$$(Sx)_i = \sum_{j \in \mathbb{Z}} a_{i-2j} x_j$$

with  $\sum_{j \in \mathbb{Z}} a_{i-2j} = 1$  for all  $i$

Here:  $a_{-2} = \frac{1}{4}$     $a_{-1} = \frac{3}{4}$     $a_0 = \frac{3}{4}$     $a_1 = \frac{1}{4}$

# Subdivision scheme $S$

Linear case:

$$x = (x_i)_{i \in \mathbb{Z}} \xrightarrow[\text{step}]{\text{refinement}} Sx = (Sx_i)_{i \in \mathbb{Z}}$$

$$\text{where } Sx_i = \sum_{j \in \mathbb{Z}} a_{i-2j} x_j$$

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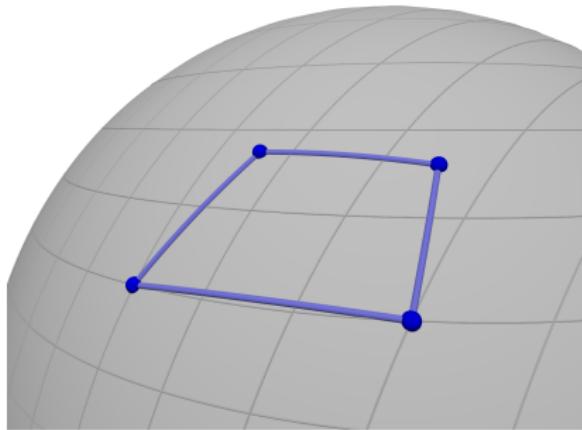
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Nonlinear case:  $\rightarrow$  geodesics



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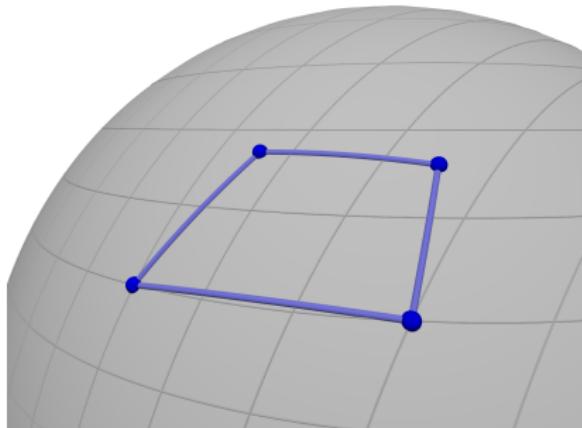
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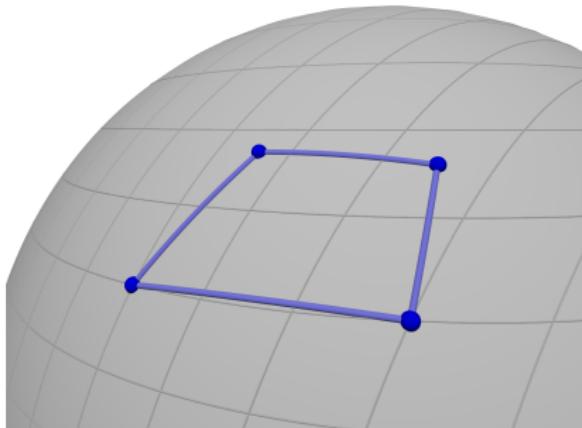
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replacing

$(\tilde{S}x)_i$  minimizer of

$$f(x) = \sum_{j \in \mathbb{Z}} a_{i-2j} \text{dist}(x_j, x)^2$$

where  $\text{dist}(\cdot, \cdot)$  denotes the

Riemannian distance

$\rightarrow$  called Riemannian center of mass

## Binary case

What happens if only two coefficients are nonzero?

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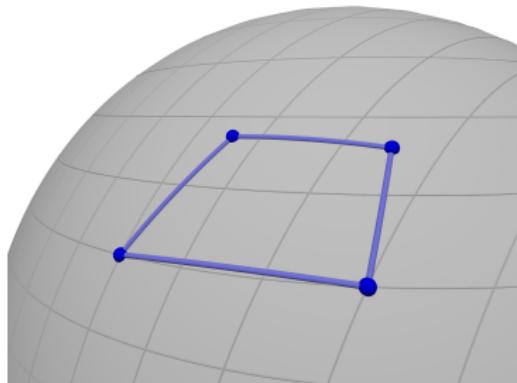
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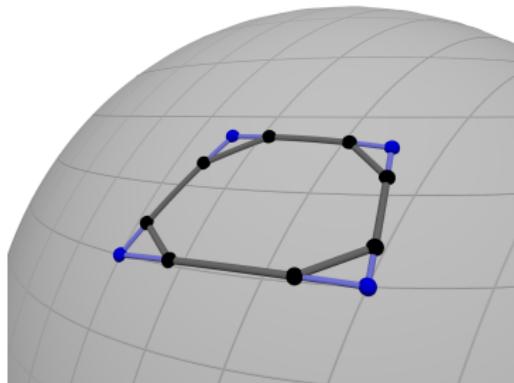
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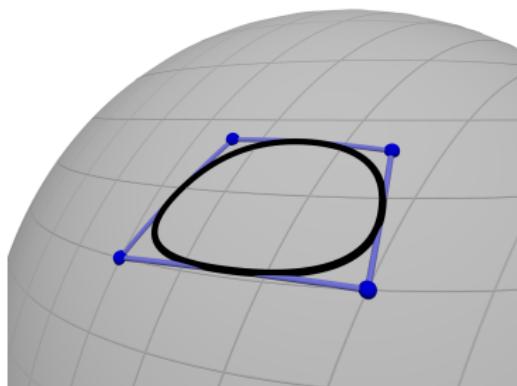
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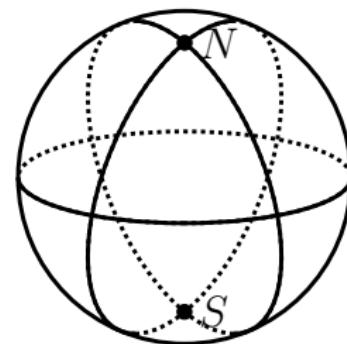
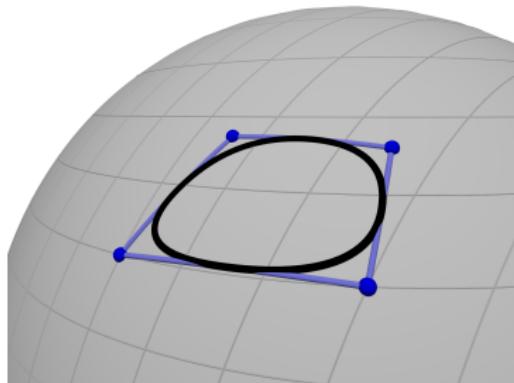
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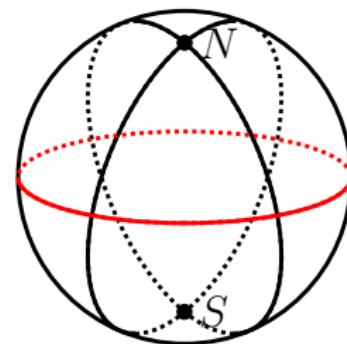
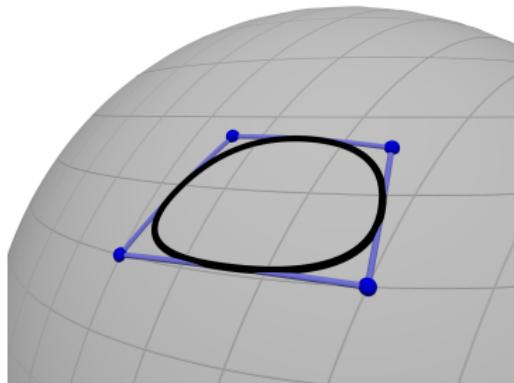
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# Convergence results on manifolds

Different ways of transferring linear schemes to manifold-valued data.

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Convergence for special subdivision schemes and all input data:

- Interpolatory subdivision schemes [Wallner, 2014]
- Schemes with nonnegative mask coefficients on Cartan-Hadamard spaces [Wallner, Nava Yazdani, Weinmann, Ebner 2011ff]

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# Convergence analysis on CH manifolds

1) Extension of linear scheme  
to CH manifold



Riemannian center of mass:

$(\tilde{S}x)_i$  minimizer of  
 $\sum_{j \in \mathbb{Z}} a_{i-2j} \text{dist}(x_j, x)^2$



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1) Extension of linear scheme  
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2) convergence analysis



contractivity  
condition

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## Riemannian center of mass

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Question: Existence and uniqueness of minimum of

$$f_\alpha(x) := \sum_{j=0}^m \alpha_j \text{dist}(x_j, x)^2$$

with  $\sum_{j=0}^m \alpha_j = 1$  on CH manifolds?

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  - Yes, for nonnegative mask coefficients on CH manifolds
- Uniqueness corresponds to convexity of  $f_\alpha$ , i.e.

$$\frac{d^2}{ds^2} f_\alpha(\gamma(s)) \geq 0$$

for any geodesic  $\gamma$ .

## Riemannian center of mass

Question: What happens if  $\sum_{j=0}^m \alpha_j = 1$  and  $\alpha_j \in \mathbb{R}$ ?

$$\begin{aligned}f_\alpha(x) &= \sum_{j=0}^m \alpha_j \operatorname{dist}(x_j, x)^2 \\&= - \int_0^{\alpha_-} \operatorname{dist}(x_{\tau(t)}, x)^2 dt + \int_{\alpha_-}^{\alpha_+ + \alpha_-} \operatorname{dist}(x_{\tau(t)}, x)^2 dt\end{aligned}$$

with  $\alpha_+ := \sum_j \alpha_j$  for all  $\alpha_j \geq 0$  and  $\alpha_- := \sum_j |\alpha_j|$  for all  $\alpha_j < 0$ .

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- Existence of minimum of  $f_\alpha$  shown by H. Hardering, 2015.
- Uniqueness follows by convexity using  $\alpha_+ - \alpha_- = 1$ :

$$\frac{1}{2} \frac{d^2}{ds^2} f_\alpha(\gamma(s)) = - \int_0^{\alpha_-} \langle J(1), J'(1) \rangle dt + \int_{\alpha_-}^{\alpha_+ + \alpha_-} \langle J(1), J'(1) \rangle dt \geq 0.$$

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well-defined

2) convergence analysis



contractivity  
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# Contractivity and Convergence result

Contractivity condition:

$$\text{dist}((\tilde{S}x)_{2i}, (\tilde{S}x)_{2i+1}) \leq \max_k (\text{dist}(x_k, x_{k+1})) \cdot \left[ \int_I |\nu_2(t) - \nu_1(t)| dt \right]$$

where  $\nu_1$  and  $\nu_2$  only depend on the mask coefficients.

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Convergence result:

$\int_I |\nu_2(t) - \nu_1(t)| dt < 1 \implies \tilde{S}$  converges to continuous limit function  
for all input data.

## Example

Four-point scheme given as

$$(Sx)_{2i} = x_i$$

$$(Sx)_{2i+1} = -\omega x_{i-1} + \left(\frac{1}{2} + \omega\right)x_i + \left(\frac{1}{2} + \omega\right)x_{i+1} - \omega x_{i+2}$$

with a parameter  $\omega$ .

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On CH manifold:

- $\omega \in (-\frac{1}{2}, 0] \Rightarrow \tilde{S}$  converges [Ebner, 2014]

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- We obtain  $\int_I |\nu_2(t) - \nu_1(t)| dt = \frac{1}{2} + 2|\omega|$  and follow that

$$-\frac{1}{4} < \omega < \frac{1}{4} \implies \tilde{S} \text{ converges.}$$

Thank you for your attention!

## References

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