Circular Coordinates under Different Cost Functions

Group 8 Project

ICERM August 2019

Applied Mathematical Modeling with Topological Techniques

Problem Setup

- We have a (simplicial) complex K from the dataset X, we can consider the homology and cohomology with a fixed coefficient field K.
- Homology and boundary operator: $\partial: C_k(K) \to C_{k-1}(K)$. e.g. $\partial[a,b,c] = [a,b] + [b,c] + [c,a]$
- ullet Cohomology and coboundary operator: $\delta: C^k(K)
 ightarrow C^{k+1}$.

e.g.
$$\delta \begin{bmatrix} a \mapsto 1 \\ b \mapsto 0 \\ c \mapsto 0 \end{bmatrix} = -[a, b]^* - [a, c]^* \text{ with } [a, b]^* = \begin{bmatrix} ab \mapsto 1 \\ bc \mapsto 0 \\ ca \mapsto 0 \end{bmatrix}.$$

• **Intuition:** If you think of boundary operators as "derivation", then the coboundary operator is like "anti-derivation".

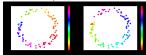
- Theorem (Circular Coordinates): Given a $[f] \in H^1(K)$, f can be made into a function $X \to S^1$. This circular coordinate can be found through the optimization problem $\min_{z \in C^0(X)} \|f \delta z\|_{L^2}$.
- **Problem** (New): How about we change the cost function $||x||_{L^2} := (\sum_i x_i^2)^{1/2}$ into:
 - $||x||_{L^1} := (\sum_i |x_i|)$ L1-norm It may introduce sparsity across coordinates instead of smoothness.
 - $(1-\lambda)\|x\|_{L^1} + \lambda \|x\|_{L^2}$ elastic net It may find a balance between L1 and L2 norms.
 - $\|x\|_{L^1} + \lambda \|x\|_{L^p}$ It may produces some other kind of smoothness.
 - Localized penalty. Only take a penalty norm for some subvector of x.
 - In addition, we can penalized not only $x=f-\delta z$ but also
 - $x = \delta z$ (minimize edits?)
 - x = z or $x = z \mod 1$ (smaller values for functions as $X \to S^1$?)

Optimization Problem: Gradient Descent

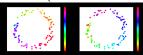
- Generic matrix optimization without Jacobian estimate: Slow and inefficient.
- Matrix optimization using Gradient Descent with Jacobian.
 - Example 1: Annulus.



• L2 norm $(x = f - \delta z \text{ and } x = z \text{ mod } 1)$



• L1 norm $(x = f - \delta z \text{ and } x = z \text{ mod } 1)$



Another Example

- Matrix optimization using Gradient Descent with Jacobian.
- Example 2: Double Annulus.



• Mixed L2 norm $((1-\lambda)\|f-\delta z\|_{L^2}+\lambda\|f-\delta z\|_{L^1})$ with $\lambda\in[0,1].$

