Heat Kernel Smoothing

Results

References

Heat Kernel Smoothing Using Laplace-Beltrami Eigenfunctions [Seo, Chung, and Vorperian, 2010]

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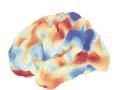
Contributions

Simulations

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Conclusions





Gabriel Riegner, March 2025 DSC 291 Network Science and Graph Theory

Cortical Surface Data

Tissue Segmentation of Anatomical MRI





Pial Surface White Matter Surface

Triangular Surface Mesh



~40 k vertices ~80 k triangles per hemisphere





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FreeSurfer Population Surface



Cortical Surface Features



curvature. surface area. cortical thickness.

[Dale et al., 1999]

Surface Distances

Distances on the Cortical Surface



Regular Domain



Irregular Domain



Geodesic Distance

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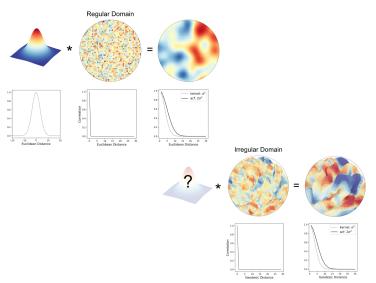
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Surface Data Smoothing



[Chung et al., 2018, Chung and Wang, 2019]

Heat Kernel

Y defined on manifold $\mathcal{M} \subset \mathbb{R}^3$:

 $Y(p) = \theta(p) + \epsilon(p), \quad \epsilon(p) \stackrel{\text{iid}}{\sim} \mathcal{N}(0.1)$

 $\Delta \psi_i = -\lambda \psi_i$

Eigenvalue problem for the Laplace-Beltrami operator Δ on \mathcal{M} :

(2)

(1)

(3)

(5)

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Heat kernel:

$$\mathcal{K}_{\sigma}(p,q) = \sum_{i=0}^{\infty} \mathrm{e}^{-\lambda_{j}\sigma} \psi_{j}(p) \psi_{j}(q)$$

$$=\sum_{i=1}^{\infty}$$

 $Y_{\sigma}(p) \equiv K_{\sigma} * Y(p) = \sum_{i=0}^{\infty} e^{-\lambda_{j}\sigma} \beta_{j} \psi_{j}(p)$

(4)

Simplification:

Heat kernel smoothing:

$$z_i \equiv \beta_i =$$

 $z_i \equiv \beta_i = \langle Y, \psi_i \rangle \stackrel{iid}{\sim} \mathcal{N}(0,1)$

[Seo et al., 2010]

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Mean:

$$\mathbb{E}[Y_{\sigma}(p)] = 0 \tag{6}$$

Variance:

$$\mathbb{V}[Y_{\sigma}(p)] = \sum_{i=0}^{\infty} e^{-2\sigma\lambda_{i}}$$
(7)

Covariance:

$$\mathbb{C}[Y_{\sigma}(p), Y_{\sigma}(q)] = \sum_{j=0}^{\infty} e^{-\lambda_j 2\sigma} \psi_j(p) \psi_j(q)$$
 (8)

Asymptotics:

$$\sigma \to 0: Y_{\sigma} \to Y$$
 (9)

$$\sigma \to \infty : Y_{\sigma} \to 0$$
 (10)

[Chung and Wang, 2019]

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Laplace Matrix: $\Delta = D - A$

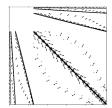


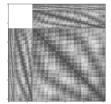
П	√ −5	1	1	1	1	0	0	0	1	0	0	0
	1	-5	0	1	0	1	0	0	1	1	0	0
	1	0	-5	0	1	0	1	0	1	0	1	0
	1	1	0	-5	1	1	0	1	0	0	0	0
	1	0	1	1	-5	0	1	1	0	0	0	0
	0	1	0	1	0	-5	0	1	0	1	0	1
	0	0	1	0	1	0	-5	1	0	0	1	1
	0	0	0	1	1	1	1	-5		0	0	1
	1	1	1	0	0	0	0	0	-5	1	1	0
	0	1	0	0	0	1	0	0	1	-5	1	1
	0	0	1	0	0	0	1	0	1	1	-5	1
	0	0	0	0	0	1	1	1	0	1	1	-5

Triangle Mesh

Laplace Matrix







Surface Mesh

Laplace Matrix

Geodesic Distance Matrix

[Reuter et al., 2006], [https://github.com/Deep-MI/LaPy]

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Eigenvalues:

$$0 = \lambda_1 \le \lambda_2 \le \dots \le \lambda_k \tag{from 2}$$

Eigenfunctions:

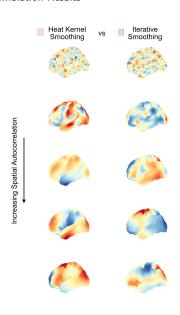
$$\psi_1, \psi_2, \dots, \psi_k \tag{from 2}$$

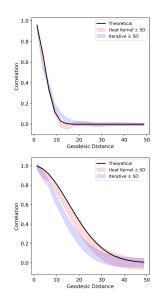
White noise

$$z_1, z_2, \dots, z_k \stackrel{\textit{iid}}{\sim} \mathcal{N}(0, 1)$$
 (from 5)

Heat Kernel Smoothing:

Simulation Results





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- **Geometry-aware smoothing**: Laplace-Beltrami eigenfunctions enable random field modeling on brain surfaces
- Spectral implementation: Heat flow modeled via adjacency and Laplacian graphs
- Validated accuracy: Simulations align with theoretical autocovariance decay with geodesic distance
- Computational efficiency: Sparse Laplacian methods scale to large meshes (40k vertices)

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