

# A Dirac Operator for Extrinsic Shape Analysis

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# Outline

**Goal:** extend spectral geometry processing

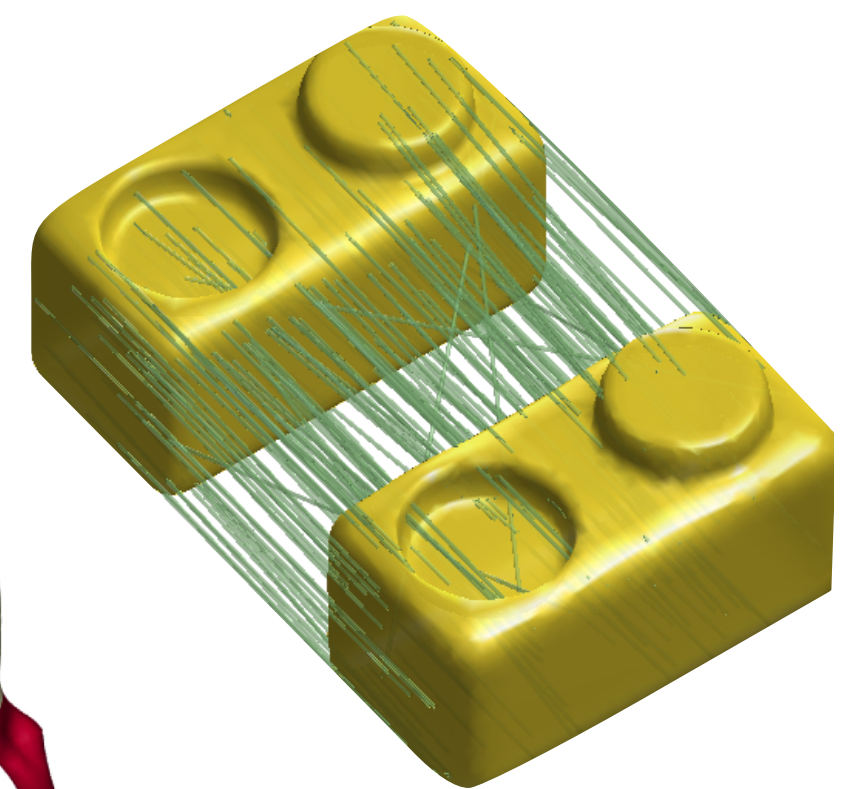
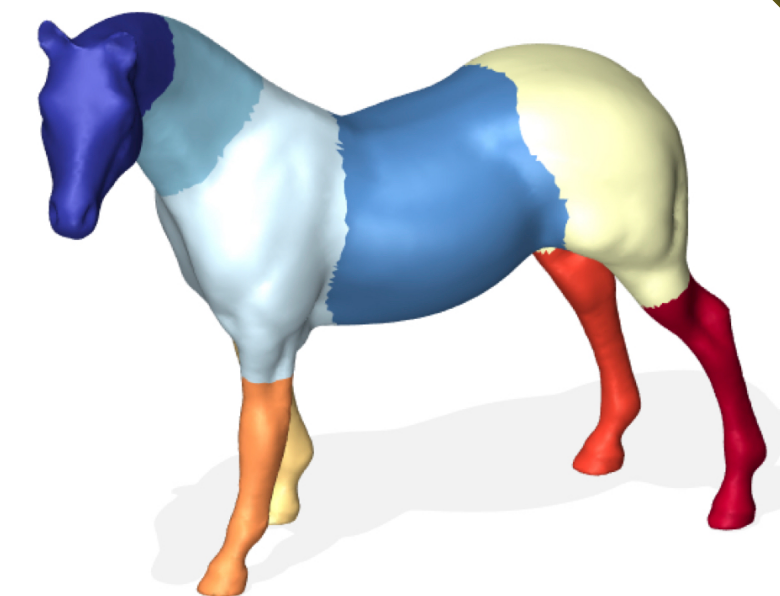
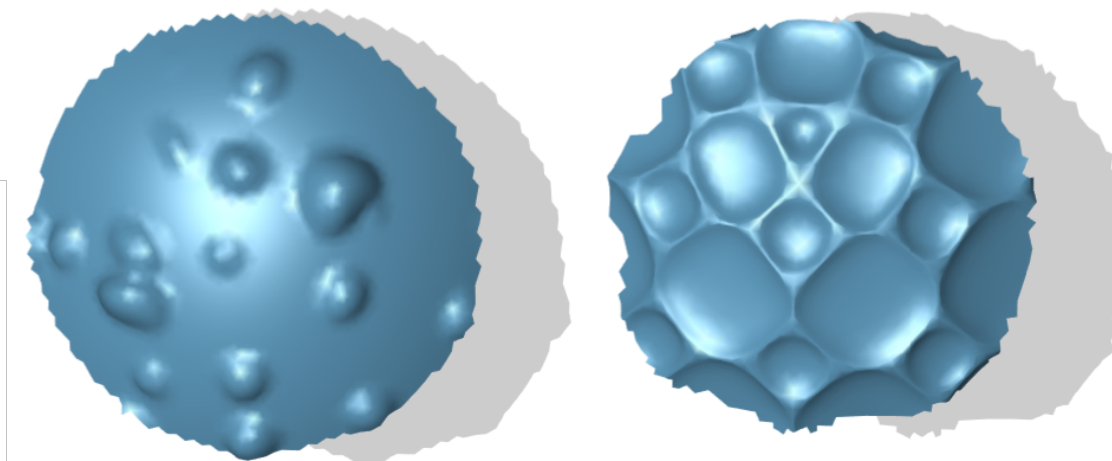
- Traditionally: *intrinsic* only (point-to-point distance)
- Today: *extrinsic* information (bending in space)

**Basic idea:** develop new differential operators

- Instead of standard Laplacian, use *relative Dirac* operator

**Applications:**

- Classification, segmentation, correspondence



# What is Spectral Geometry Processing?

$$L\phi_i = \lambda_i \phi_i$$

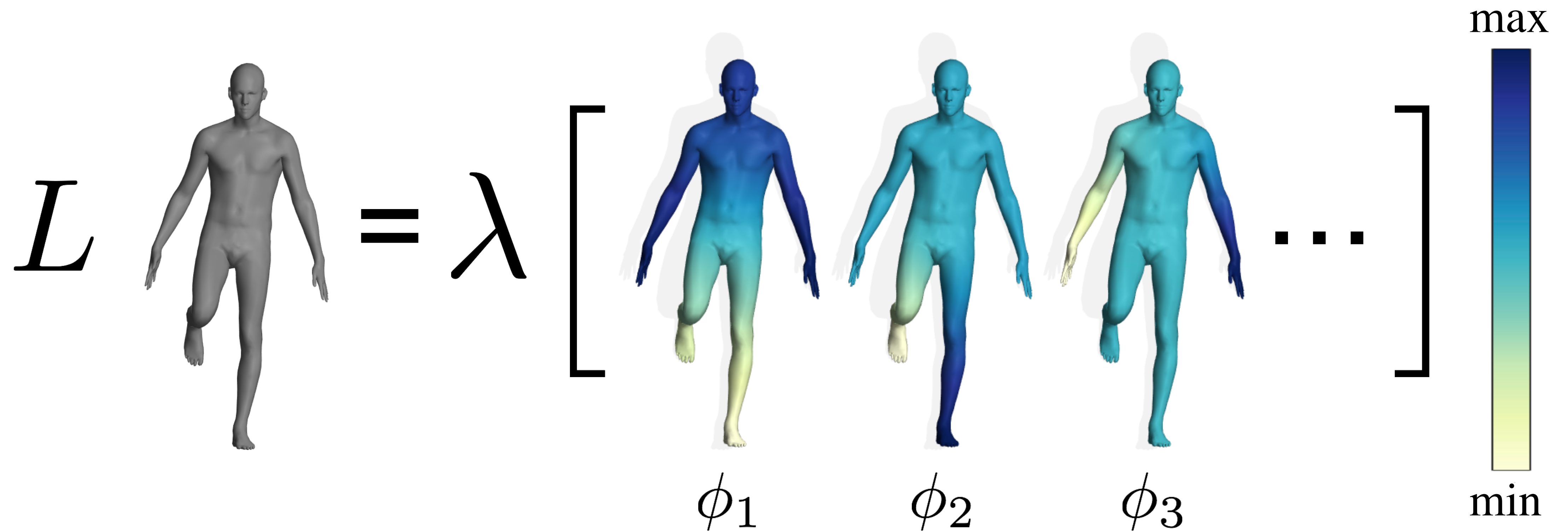
differential operator

eigenvalues

eigenvectors

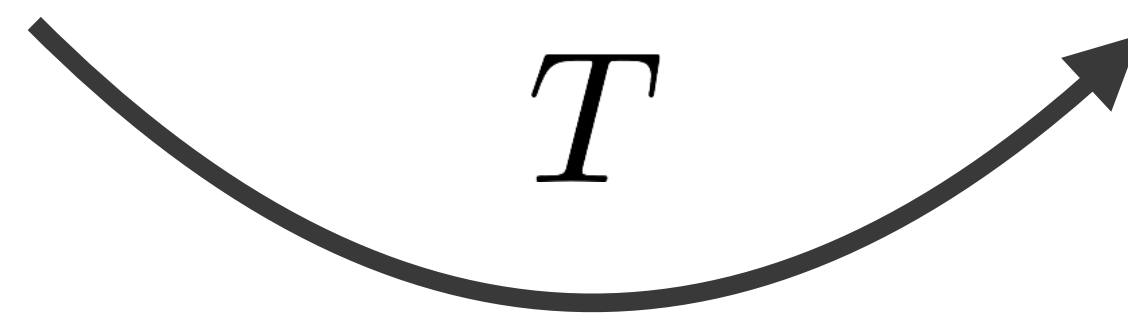
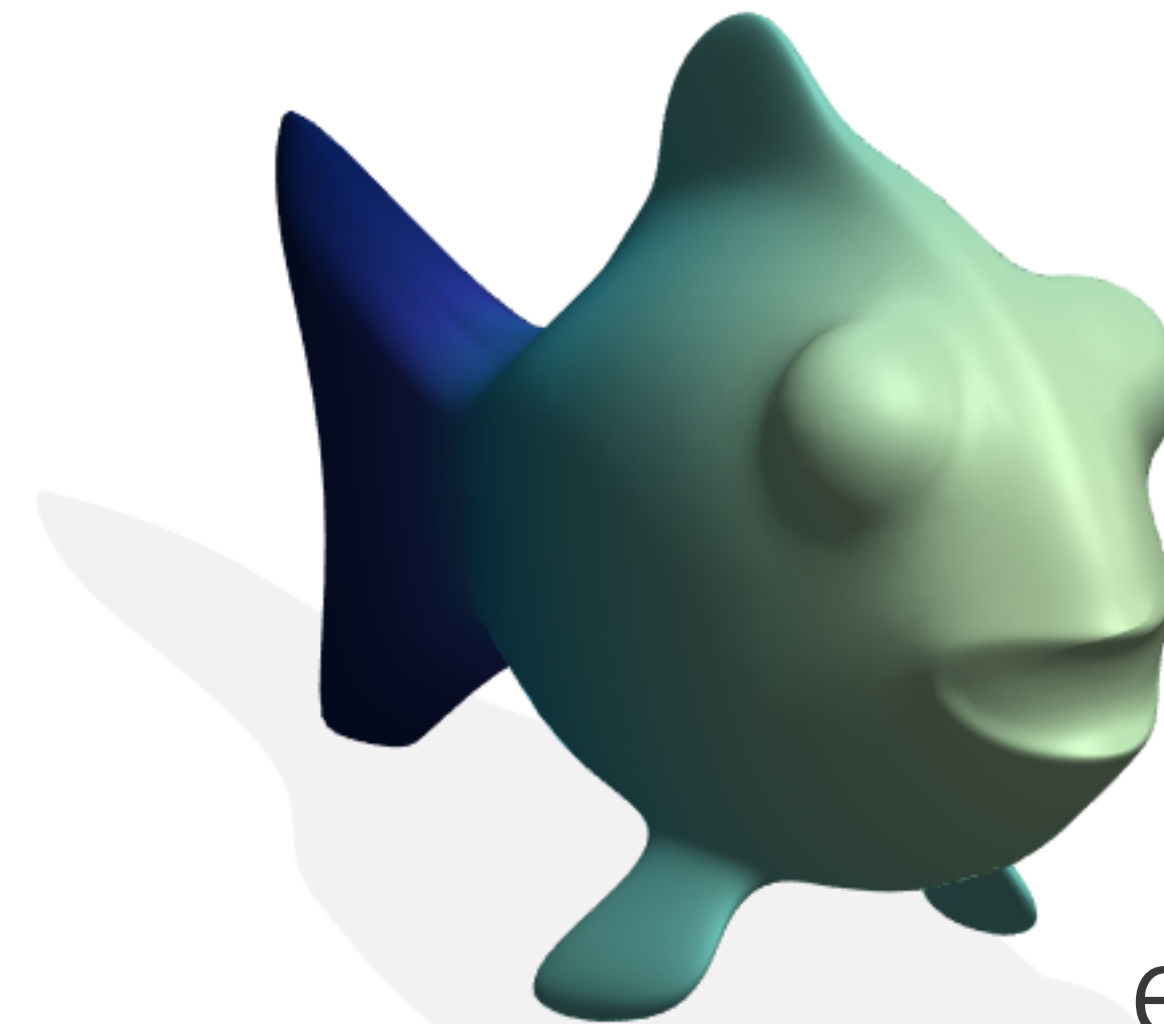
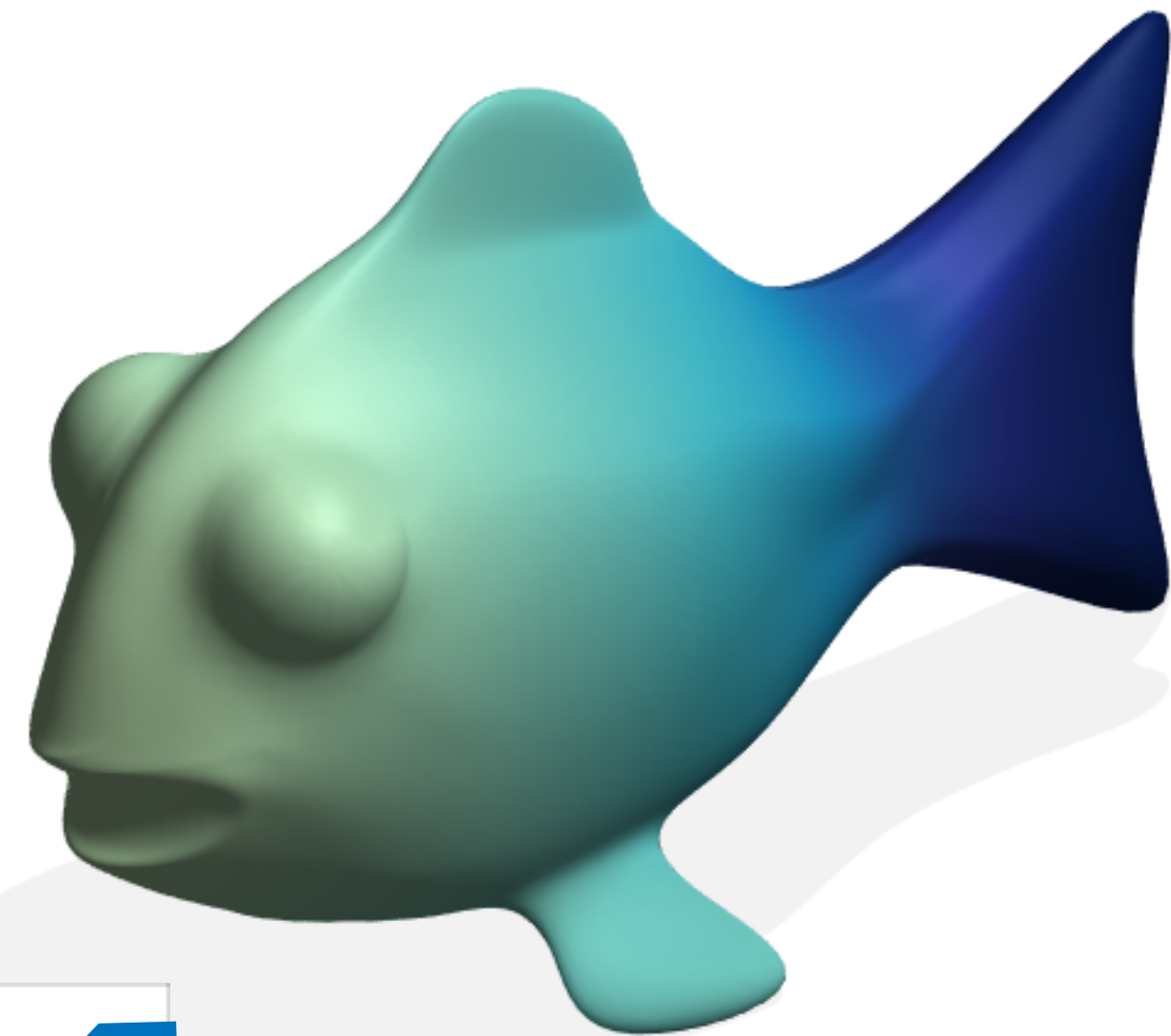
The diagram illustrates the eigenvalue equation  $L\phi_i = \lambda_i \phi_i$ . Three arrows point from labels below to terms in the equation: one from 'differential operator' to  $L$ , one from 'eigenvalues' to  $\lambda_i$ , and one from 'eigenvectors' to  $\phi_i$ .

# What is Spectral Geometry Processing?

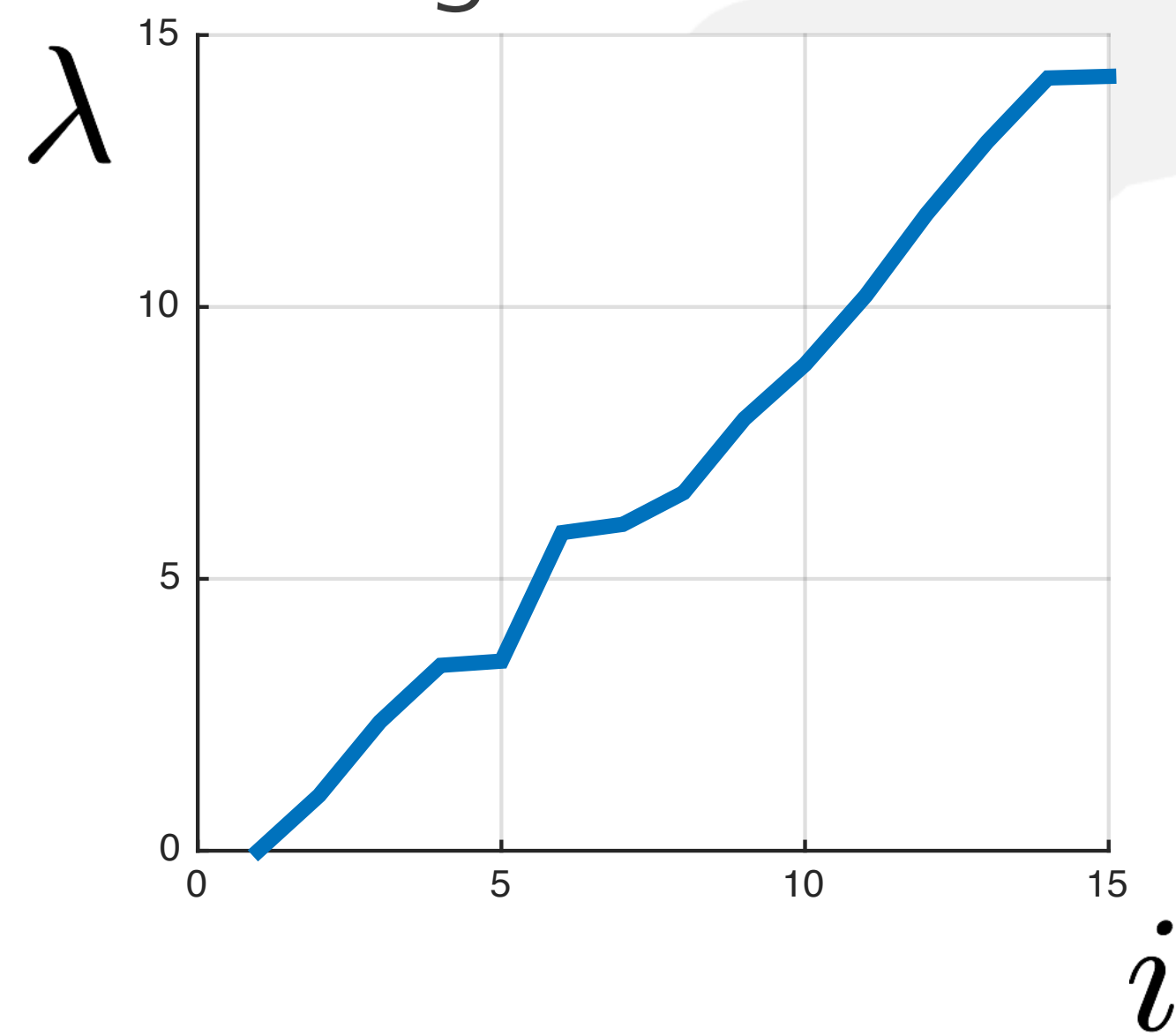


Analogy: "Fourier transform" for surfaces

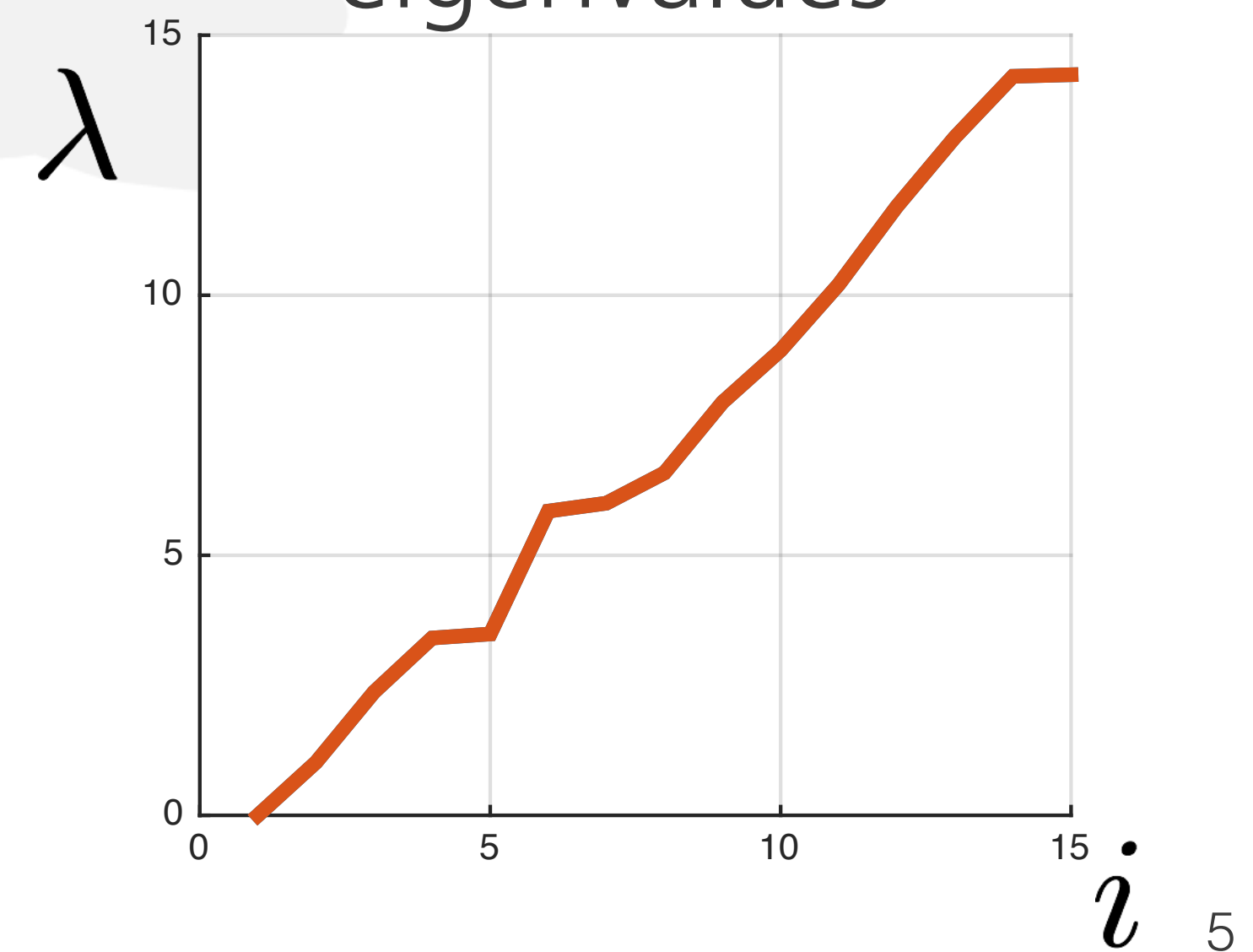
# Why - Coordinate Invariant



eigenvalues

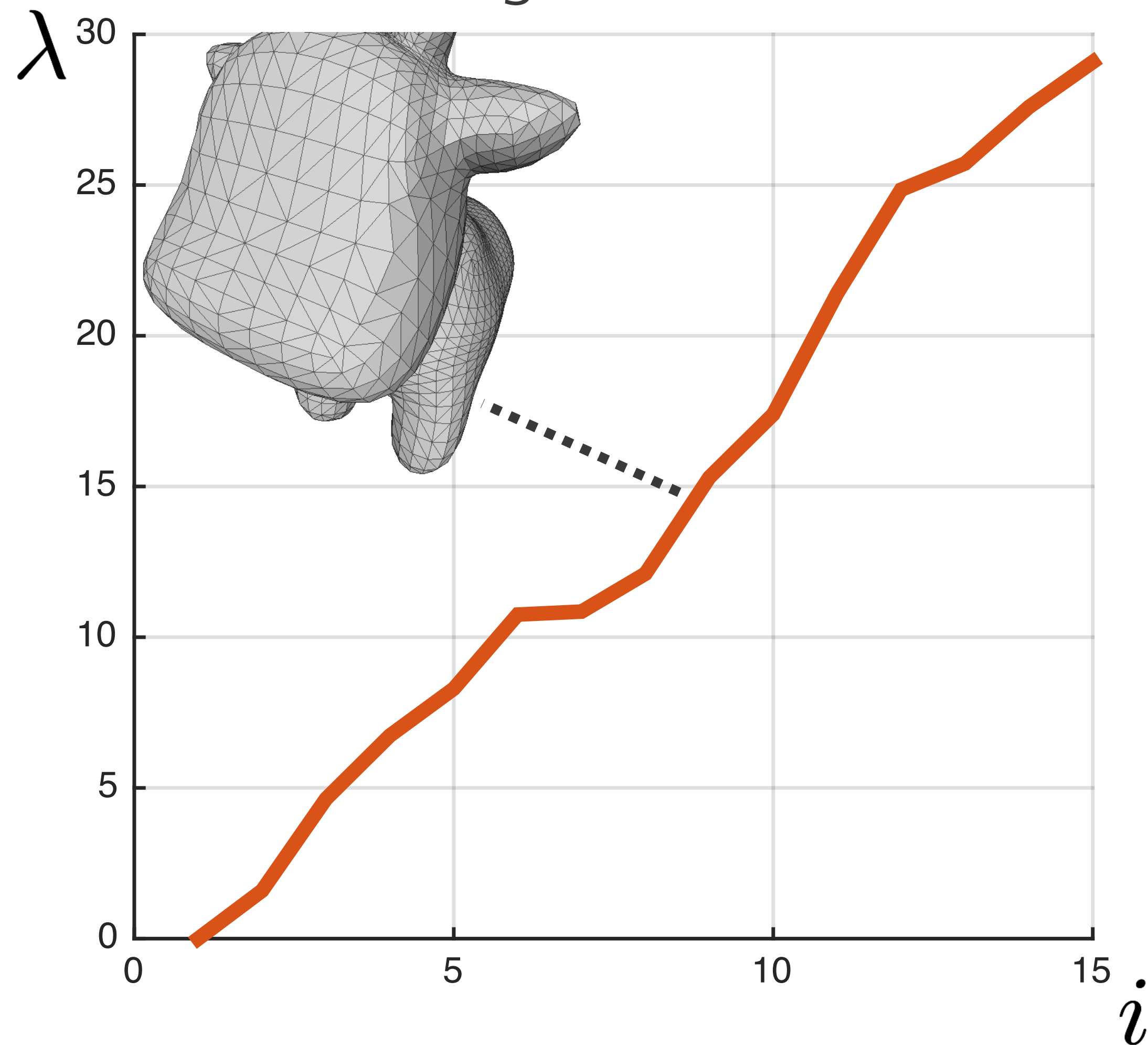


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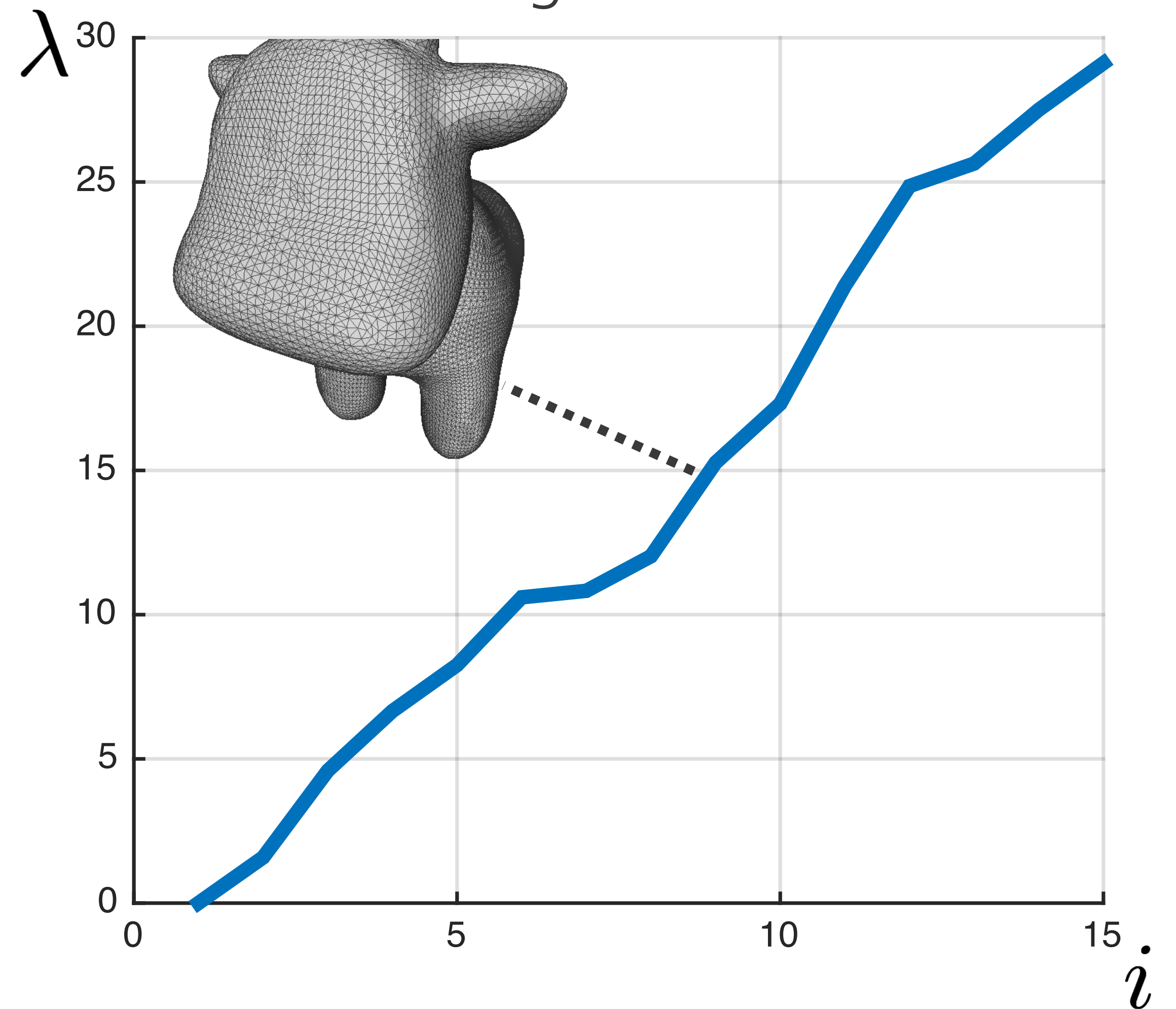


# Why - (Almost) Invariant to Tessellation

eigenvalues

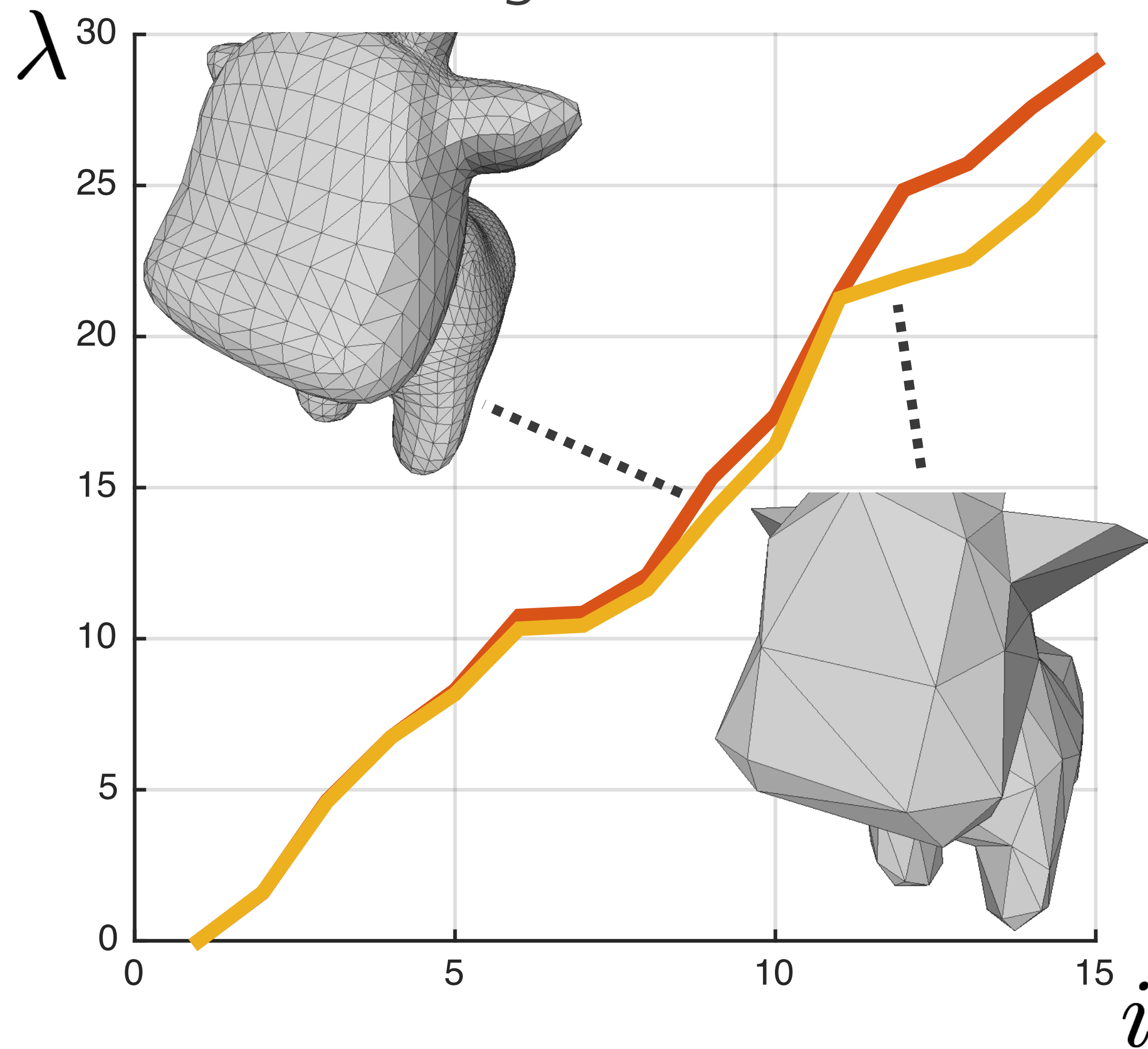


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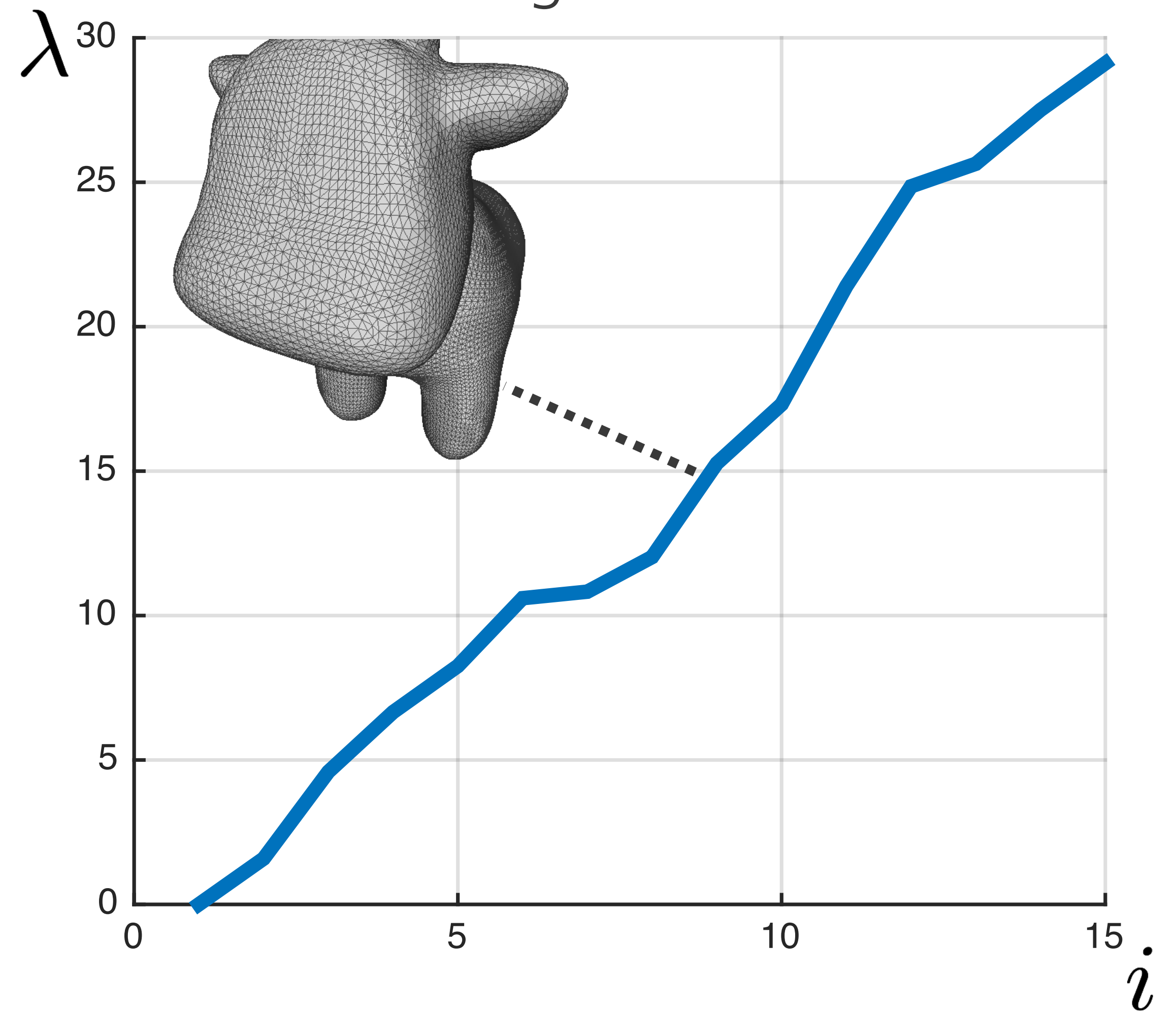


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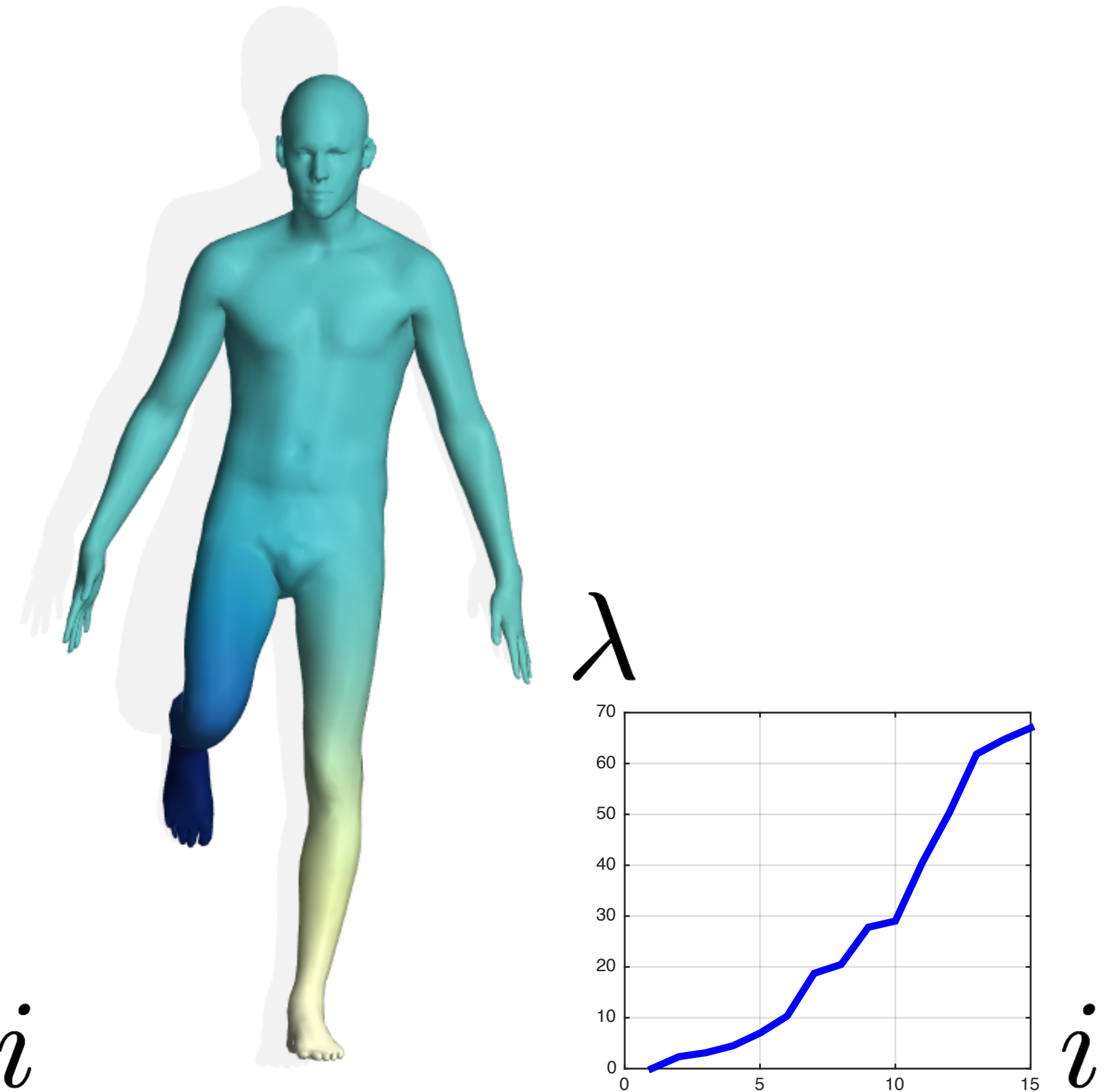
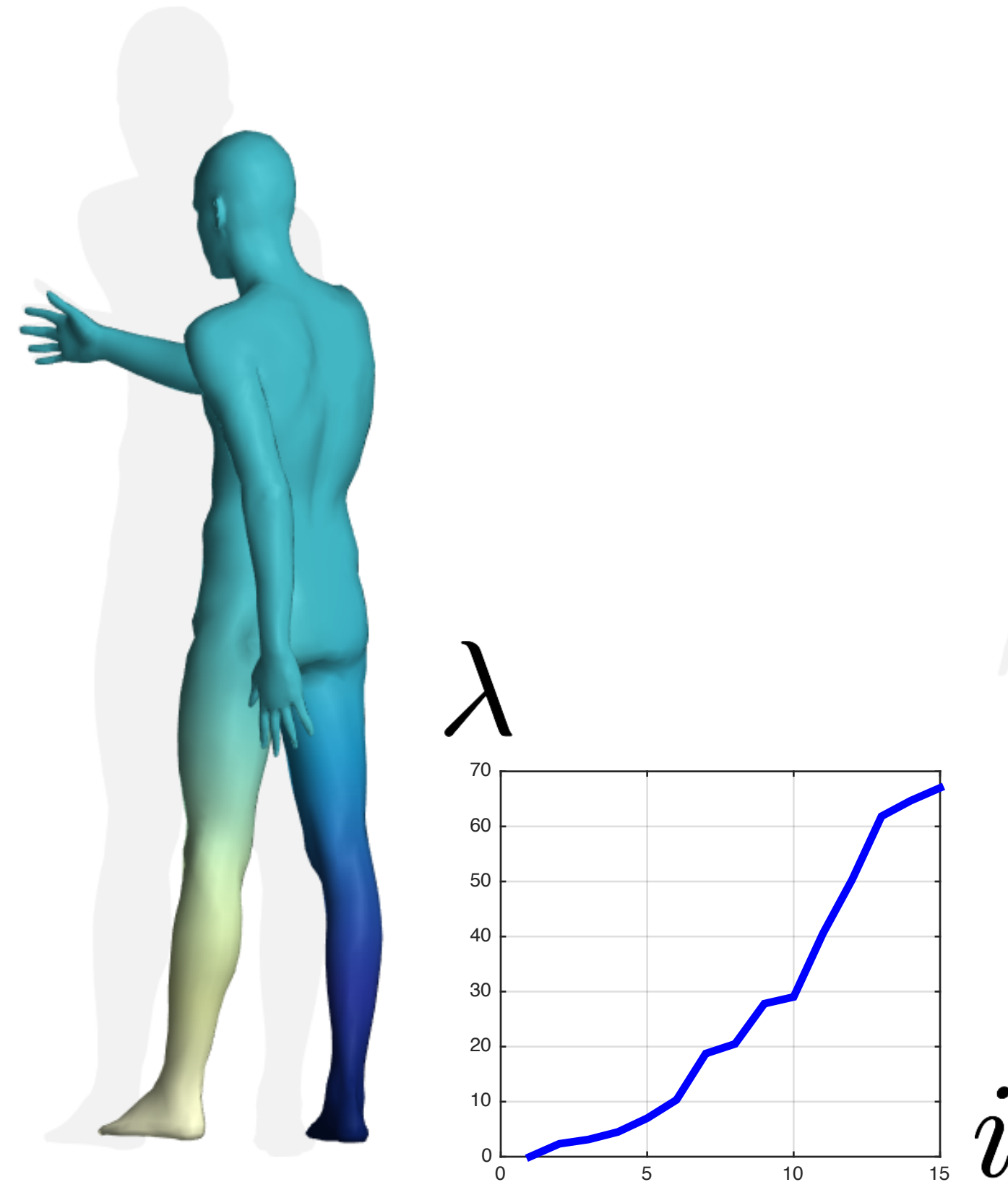
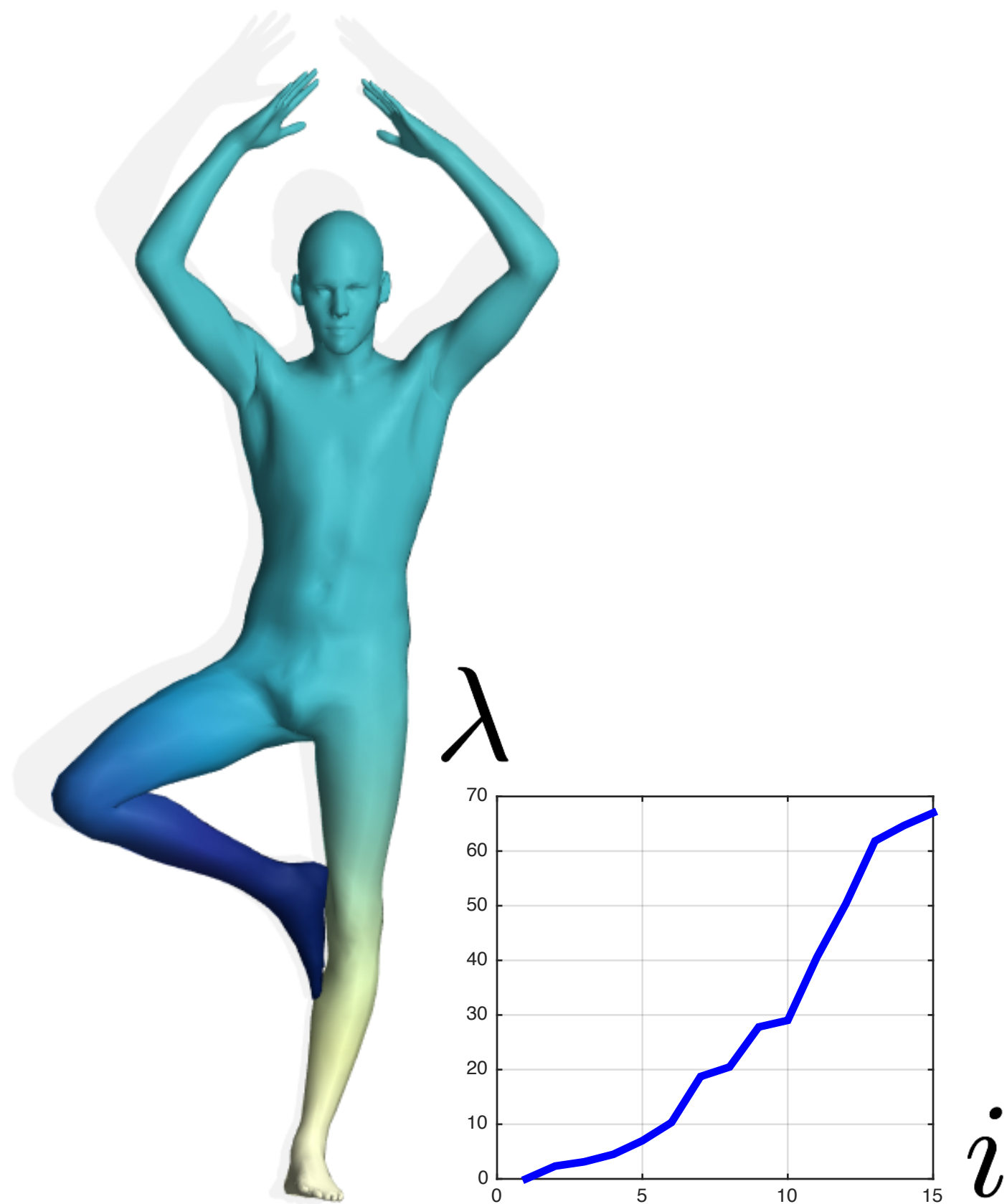


eigenvalues



# Why - Isometry Invariant

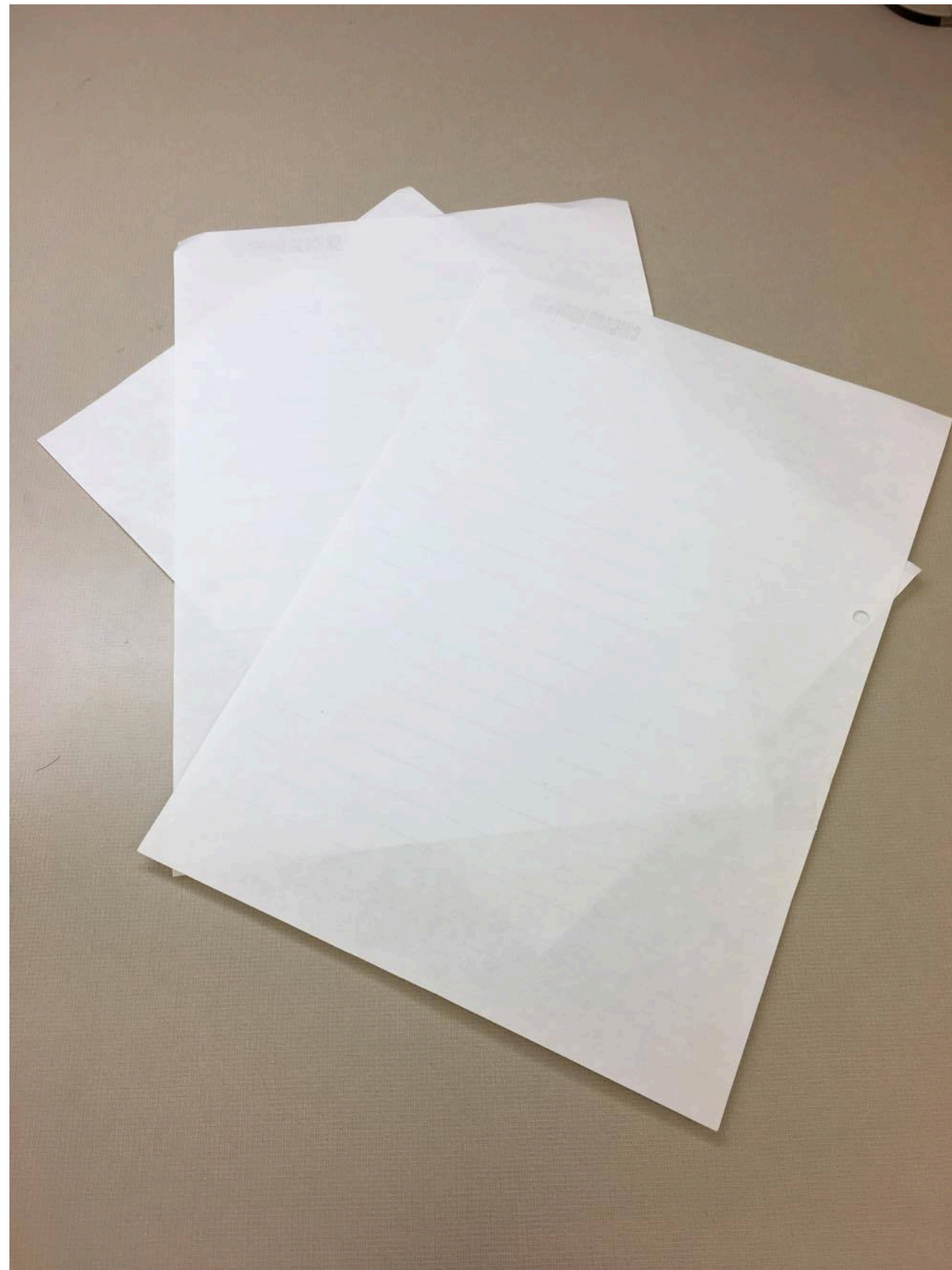
Benefits from Laplace-Beltrami operator  $\Delta$



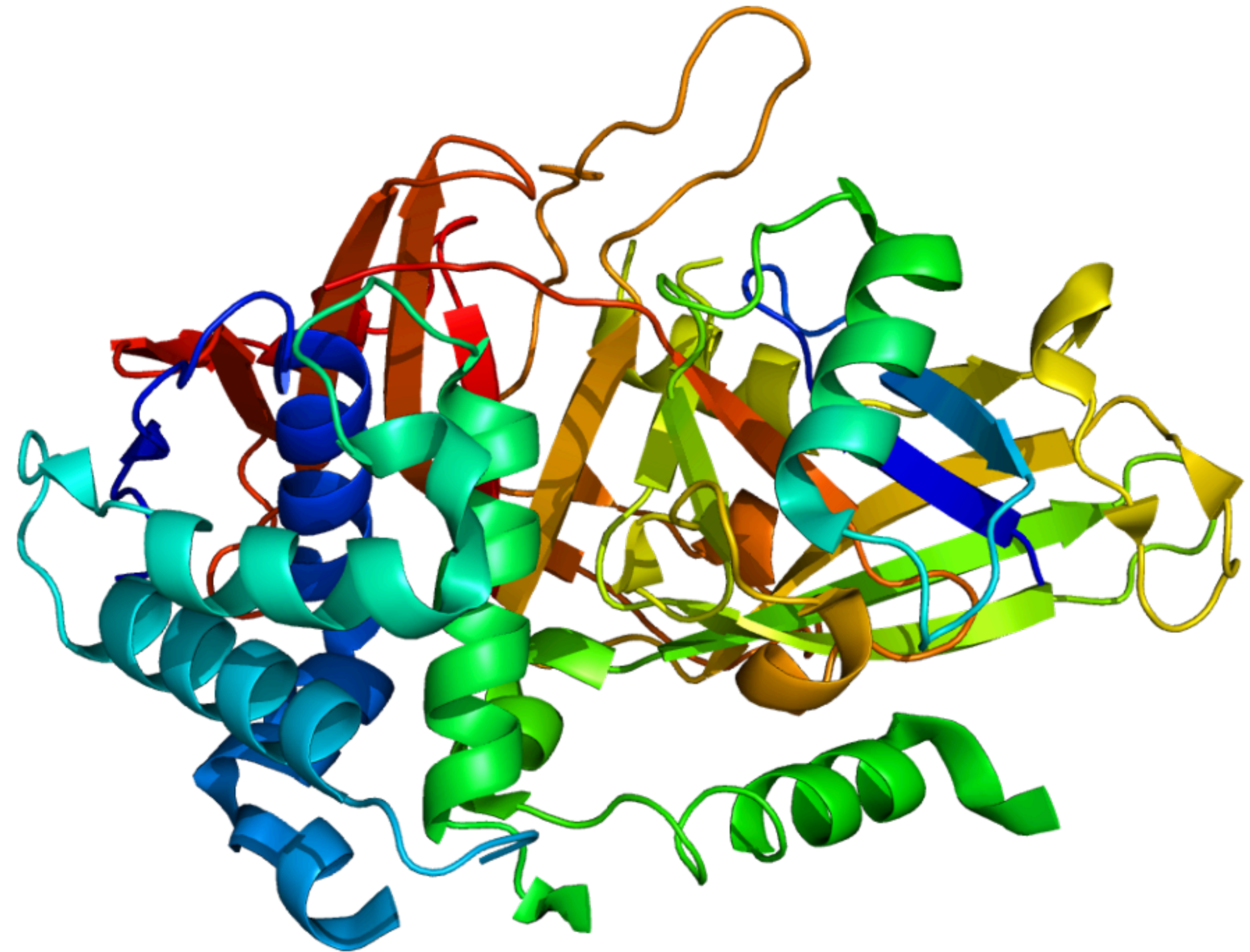
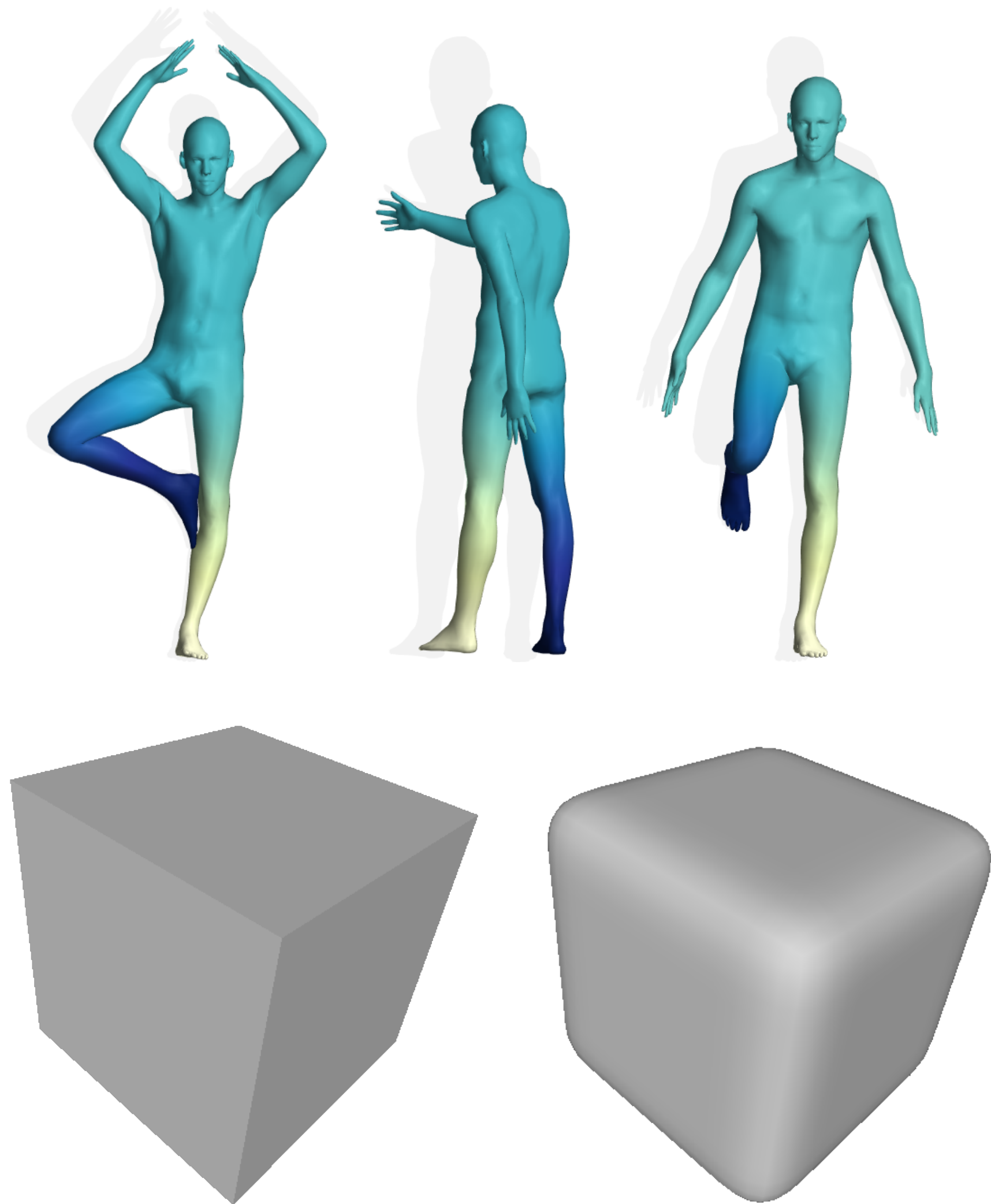


# Isometry Invariance

## Is it a feature or a bug?



# Sensitivity to Metric Distortion

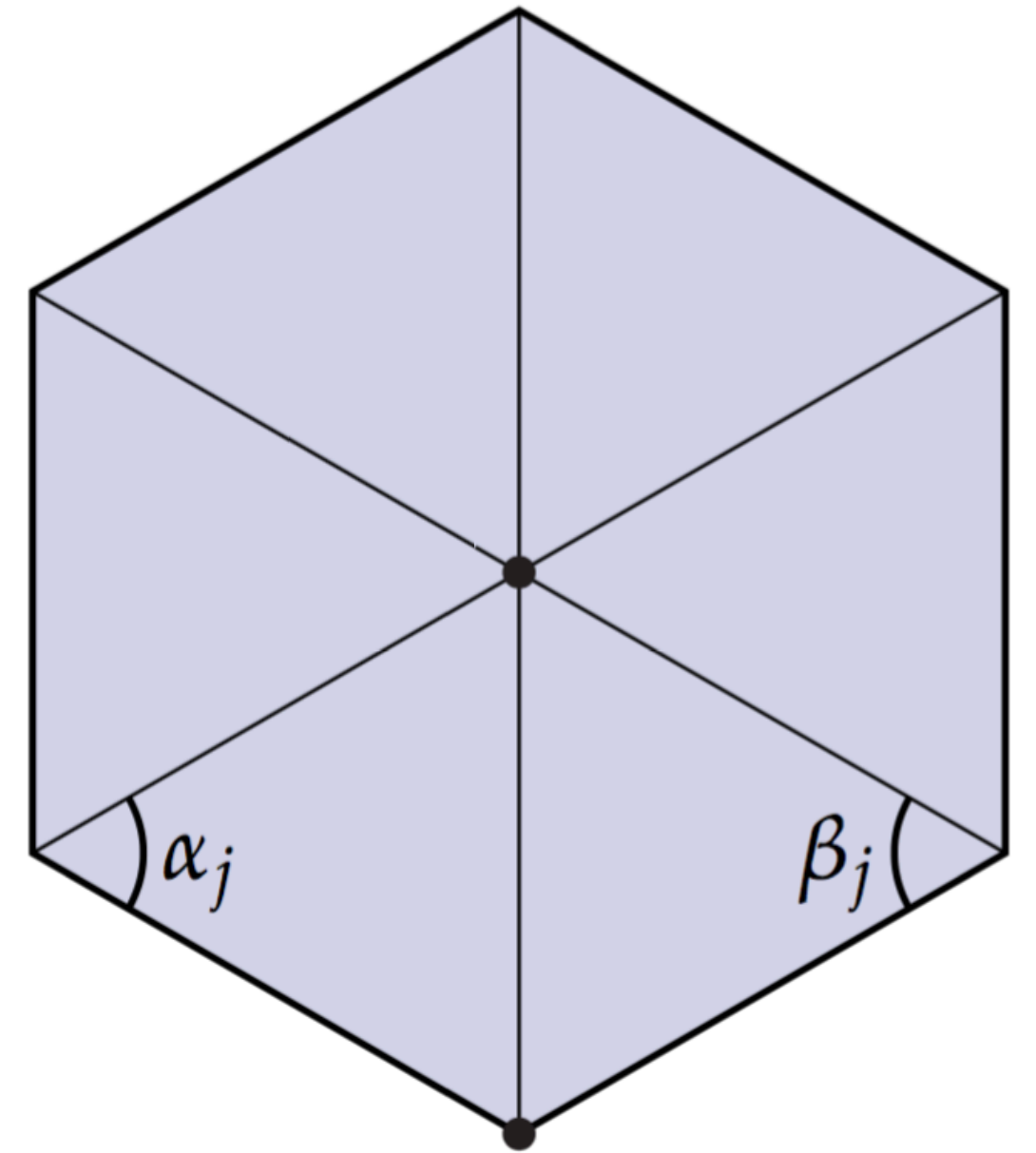


# (Discrete) Differential Operators

# Laplace-Beltrami Operator (intrinsic) $\Delta$

- Discrete cotangent Laplacian

$$\Delta p_i = \frac{1}{2\mathcal{A}_i} \sum_{j \in \mathcal{N}(i)} (\cot \alpha_{ij} + \cot \beta_{ij})(p_i - p_j)$$



# Laplace-Beltrami Operator (intrinsic) $\Delta$

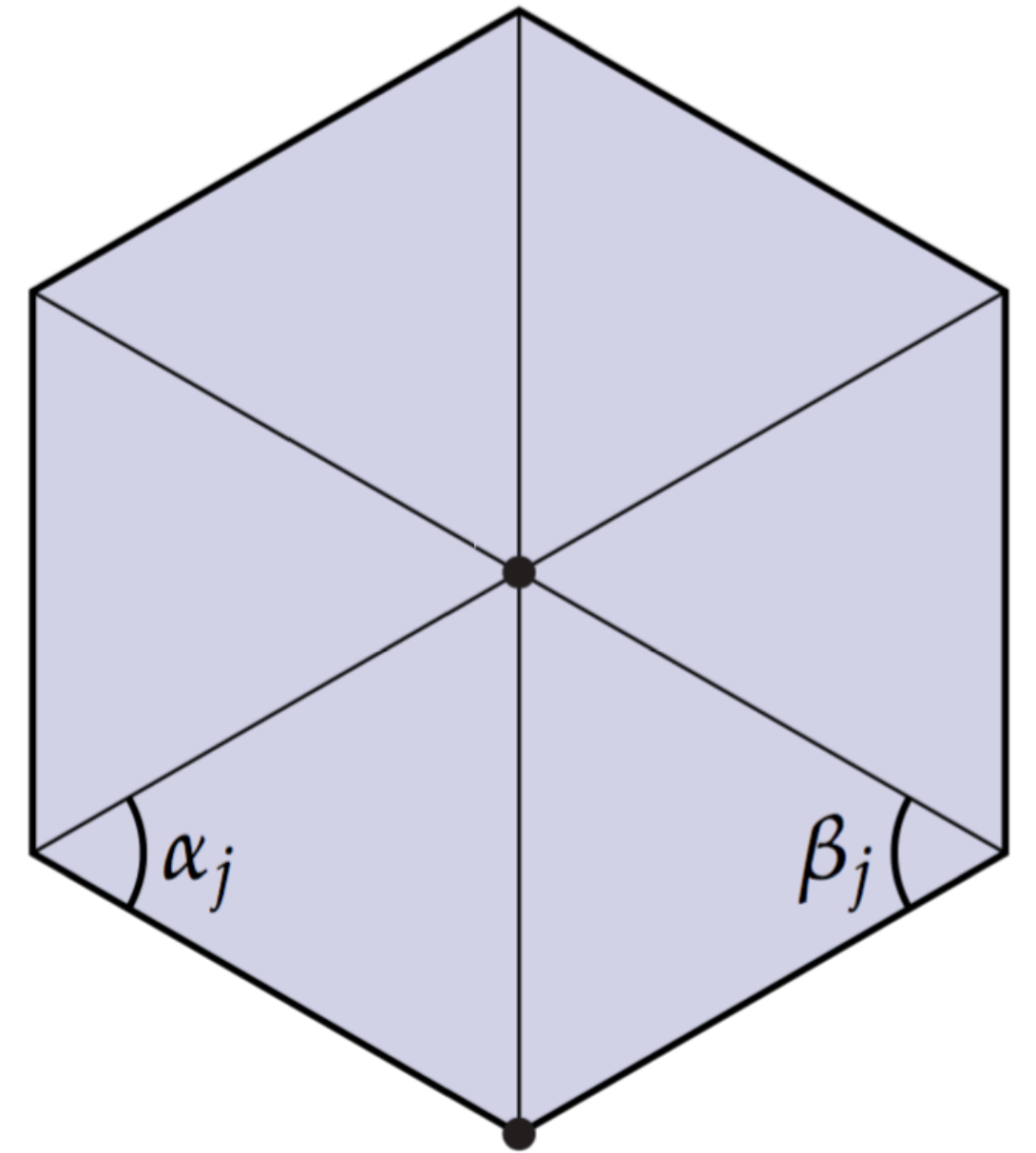
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- Key idea:

Laplace only depends on **edge lengths!**

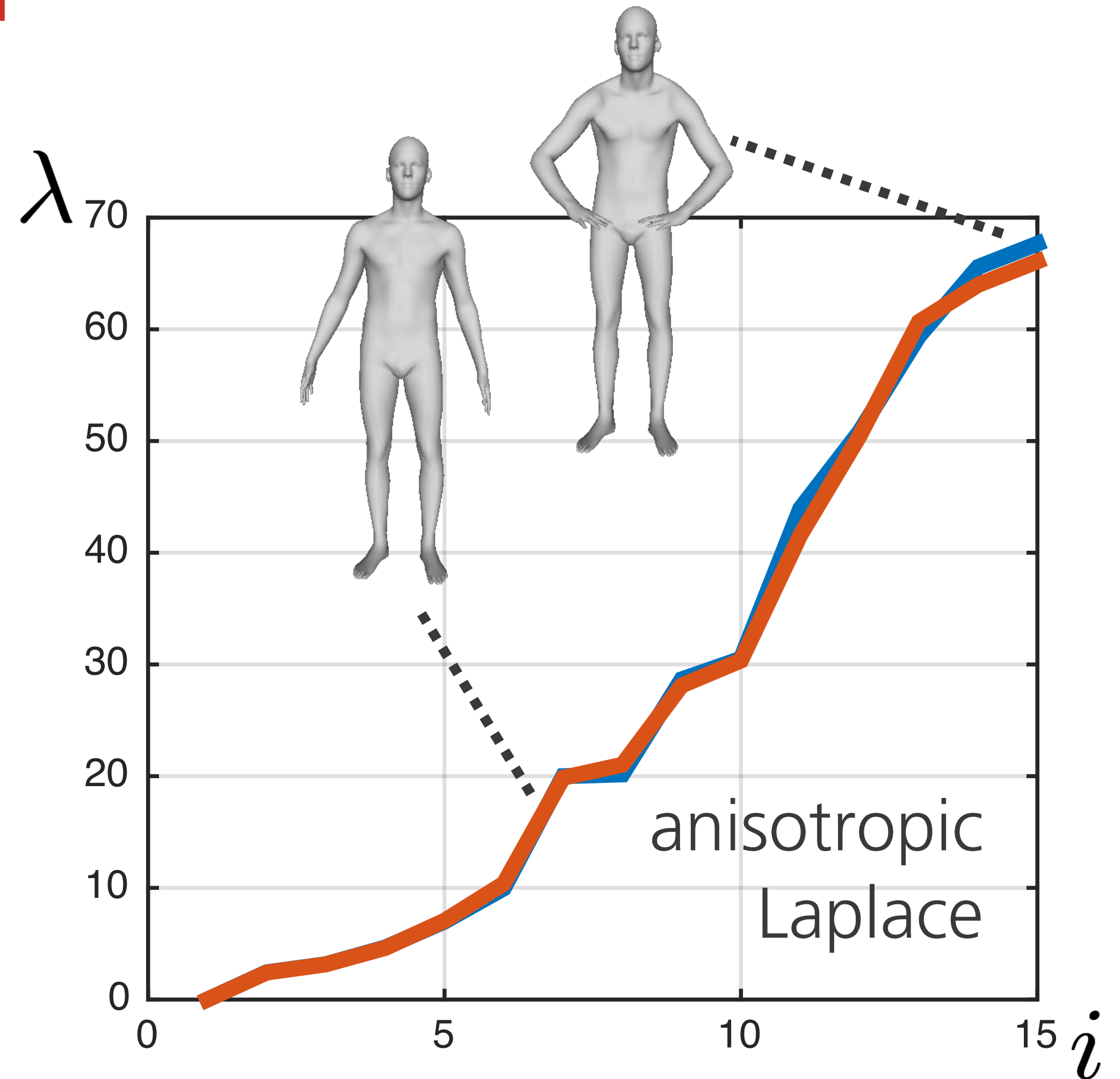
Edge length is an intrinsic quantity



# Not Purely Intrinsic Operators

- Mixture of intrinsic and extrinsic information
- Existing operators:
  - Anisotropic Laplace
  - Modified Dirichlet energy

**How sensitive?**



# Quaternionic Dirac Operator $D$

- Definition:

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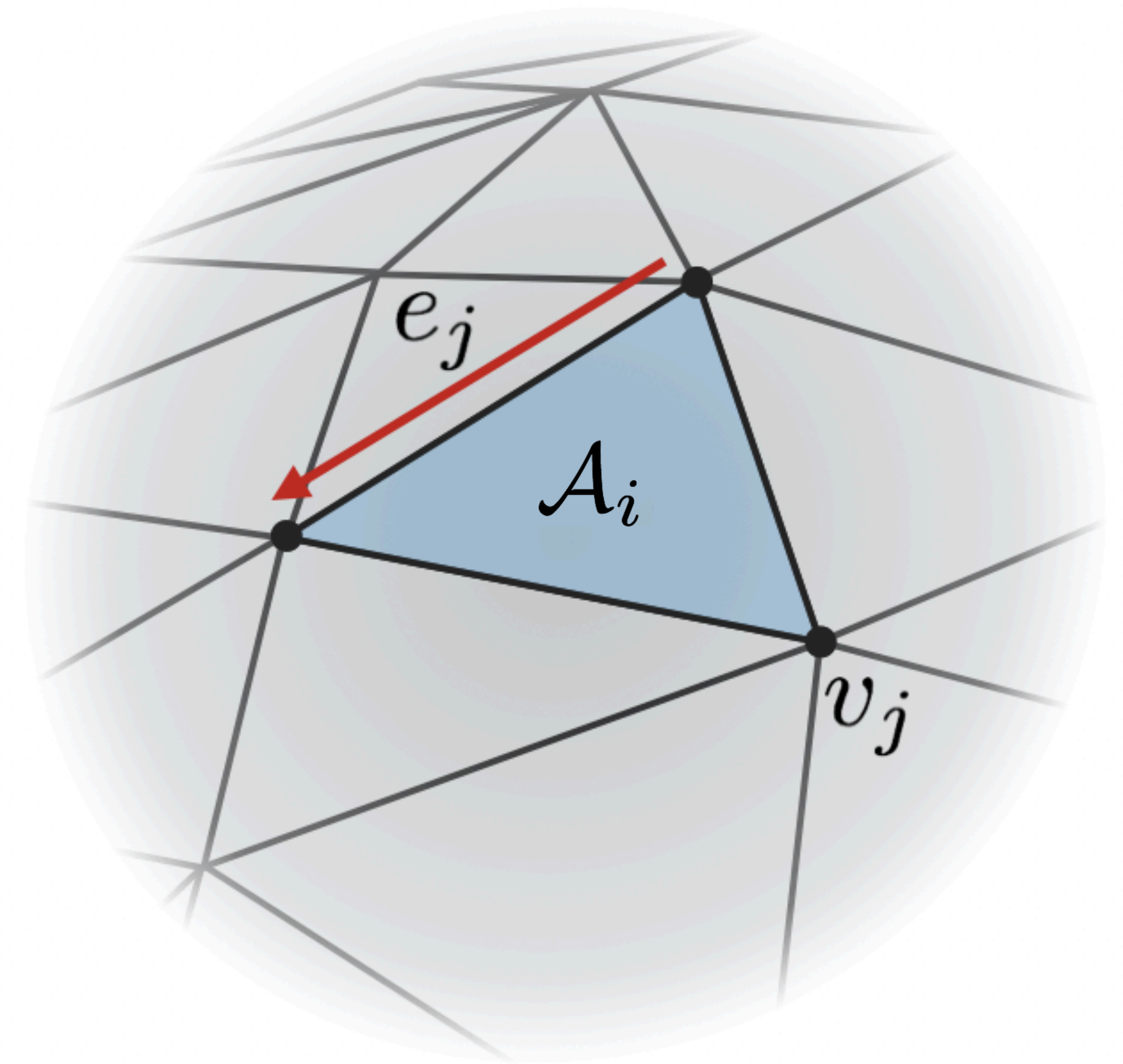


# Quaternionic Dirac Operator $D$

- Discrete Dirac:

$$D_{ij} = \frac{-1}{2\mathcal{A}_i} e_j$$

- Key idea:  
depends on **edge vectors**  
(rather than edge length)



# Square of Dirac Operator

$$D^2\psi = \underset{\substack{\text{Laplace} \\ \downarrow}}{\Delta}\psi + \underset{\substack{\text{Normal} \\ \downarrow}}{\boxed{\frac{dN \wedge d\psi}{|df|^2}}}$$

**intrinsic**                      **extrinsic**

**Relative Dirac Operator**  
 $D_N$

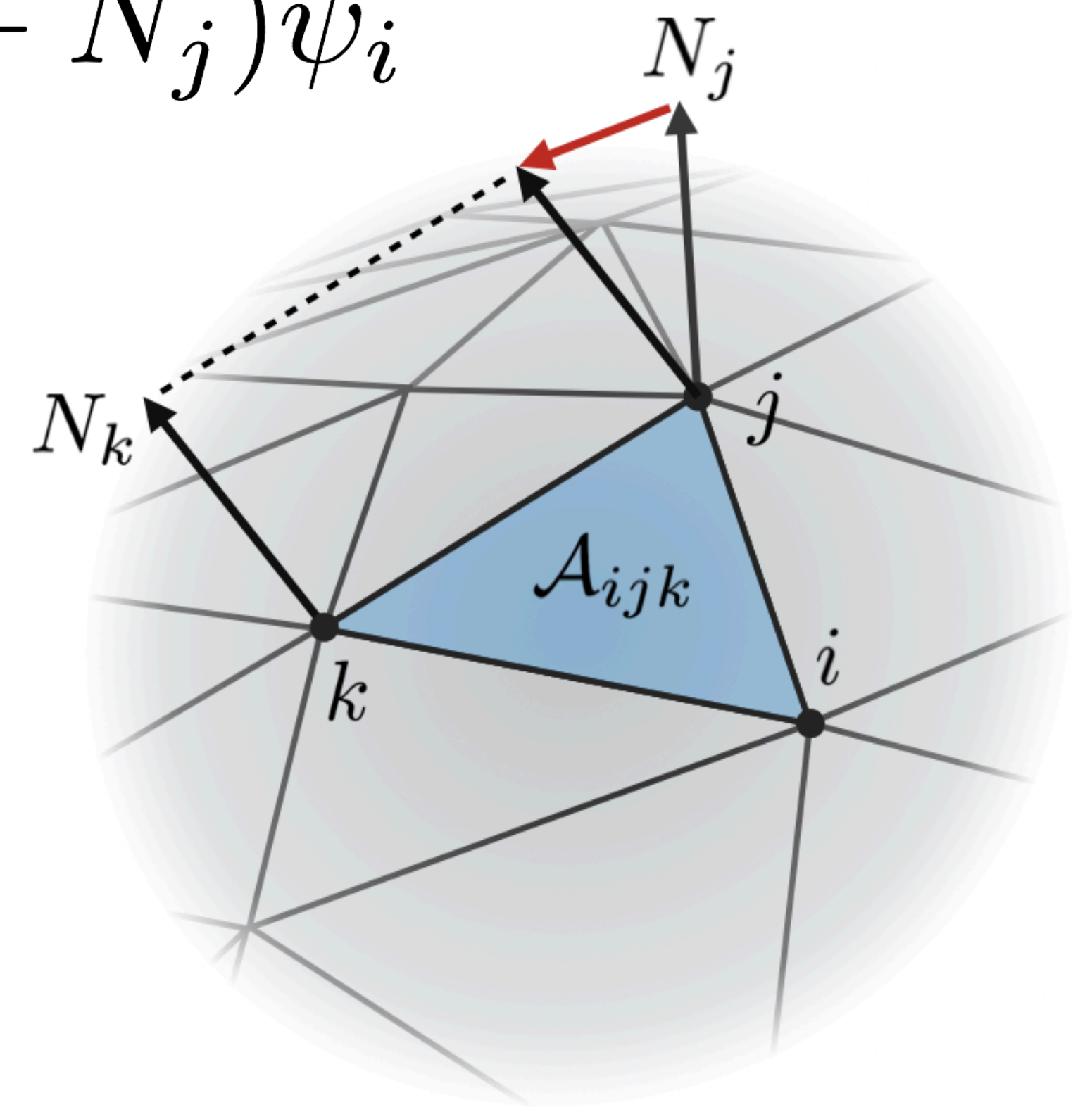
# Discretization

- Discrete relative Dirac

$$(D_N \psi)_{ijk} = -\frac{1}{2\mathcal{A}_{ijk}} \sum_{pqr \in \mathcal{C}(ijk)} (N_k - N_j) \psi_i$$

- Matrix form

$$D_{N_{ijk}, i} = -\frac{N_k - N_j}{2\mathcal{A}_{ijk}}$$



# Basic Properties of Relative Dirac

$$\begin{array}{ccc} \text{(Dirac)} & & \text{(relative Dirac)} \\ D\psi = -\frac{df \wedge d\psi}{|df|^2} & \longrightarrow & D_N\psi = -\frac{dN \wedge d\psi}{|df|^2} \end{array}$$

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  - eigenvalues **are** coordinate invariant
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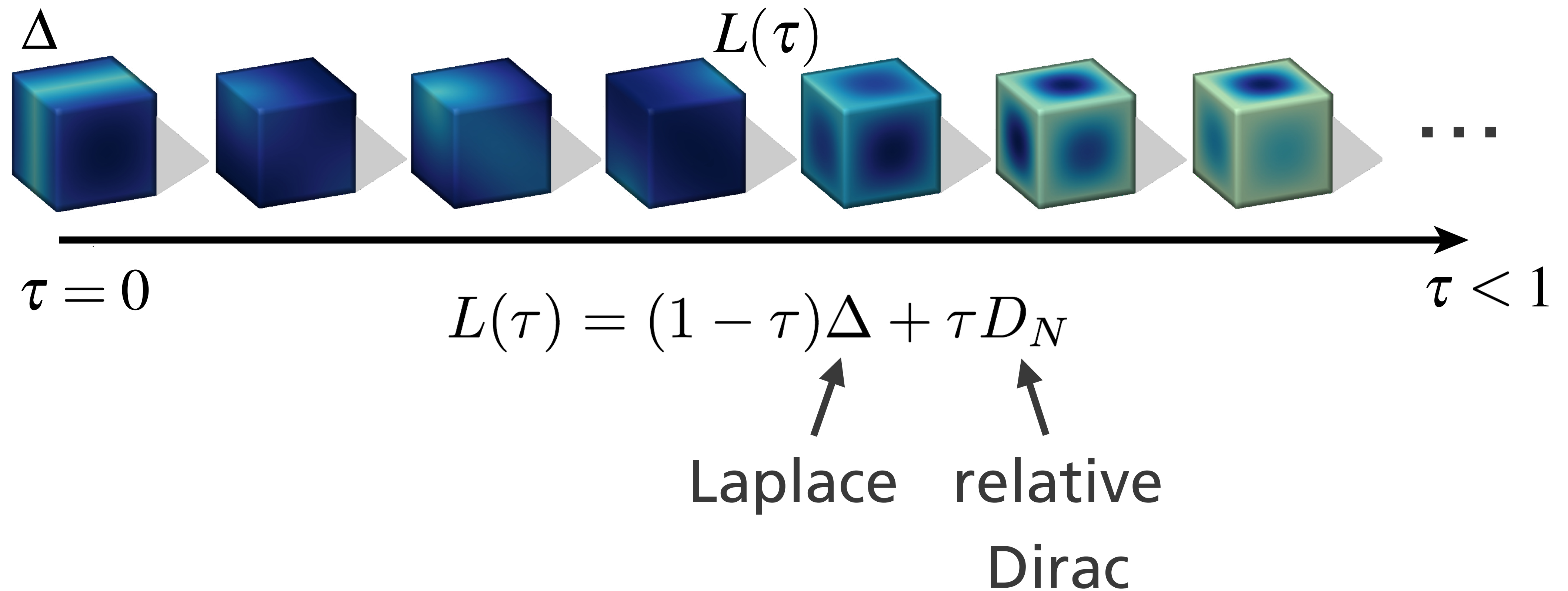
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**Not isometry invariant!!**

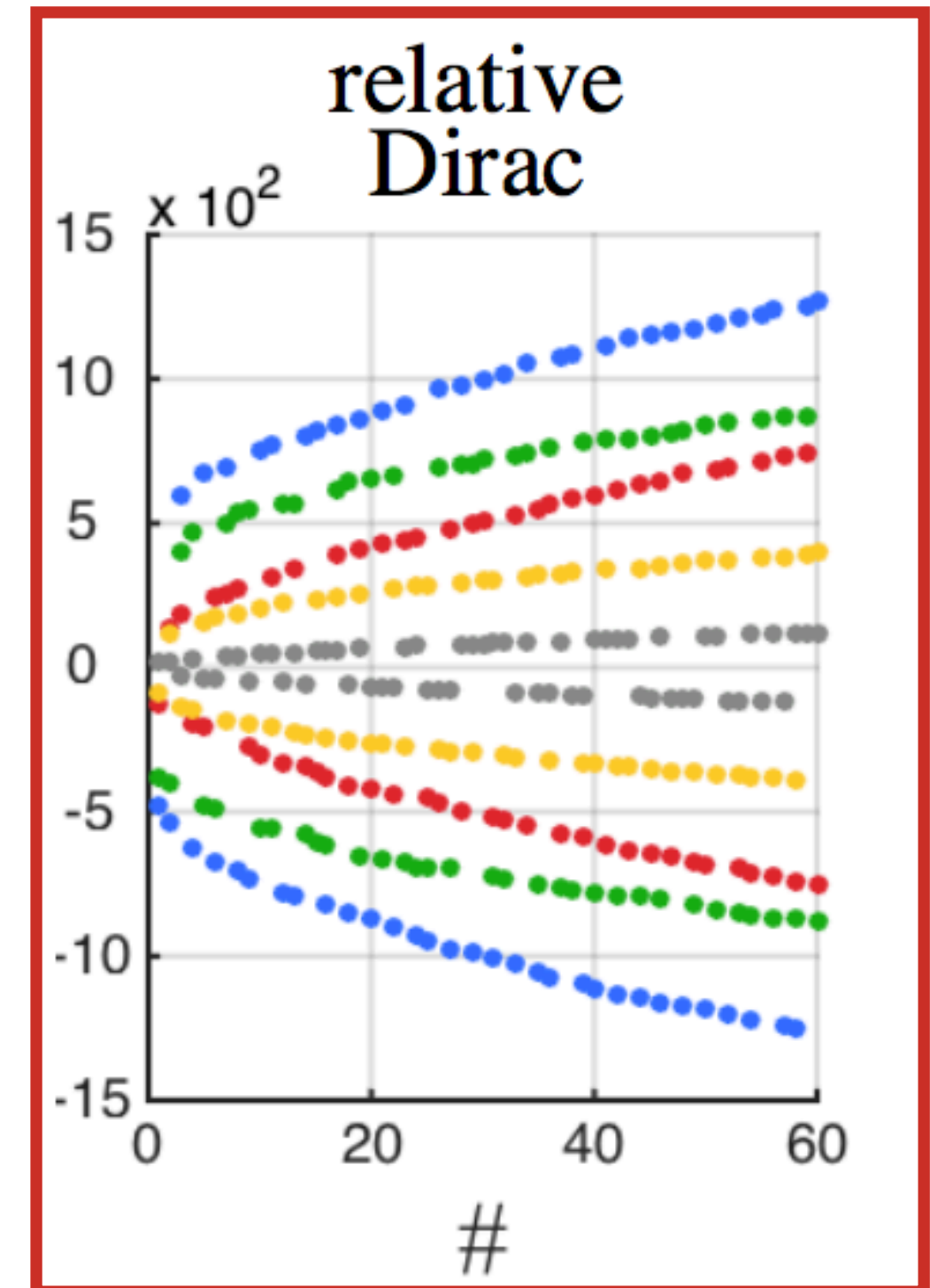
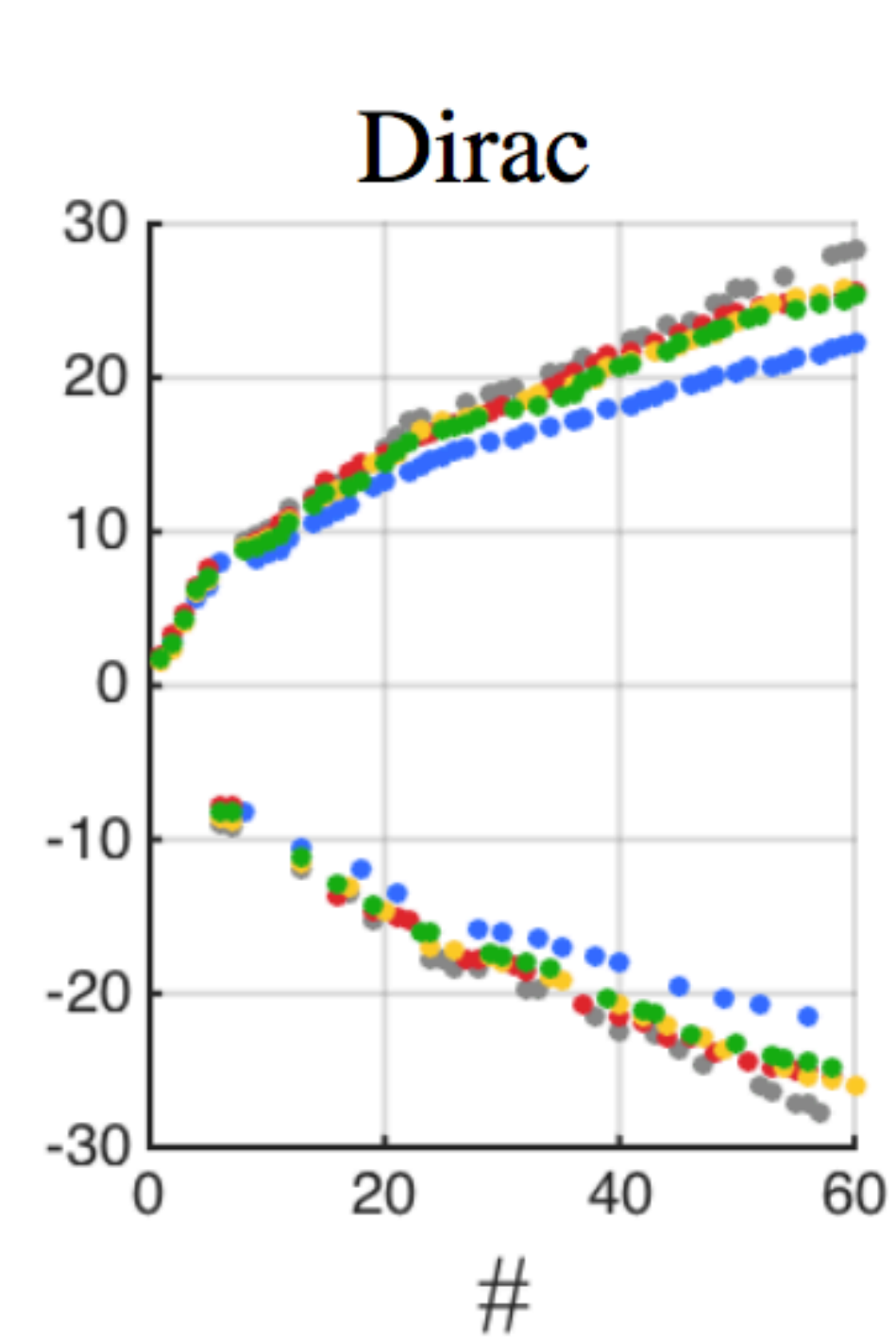
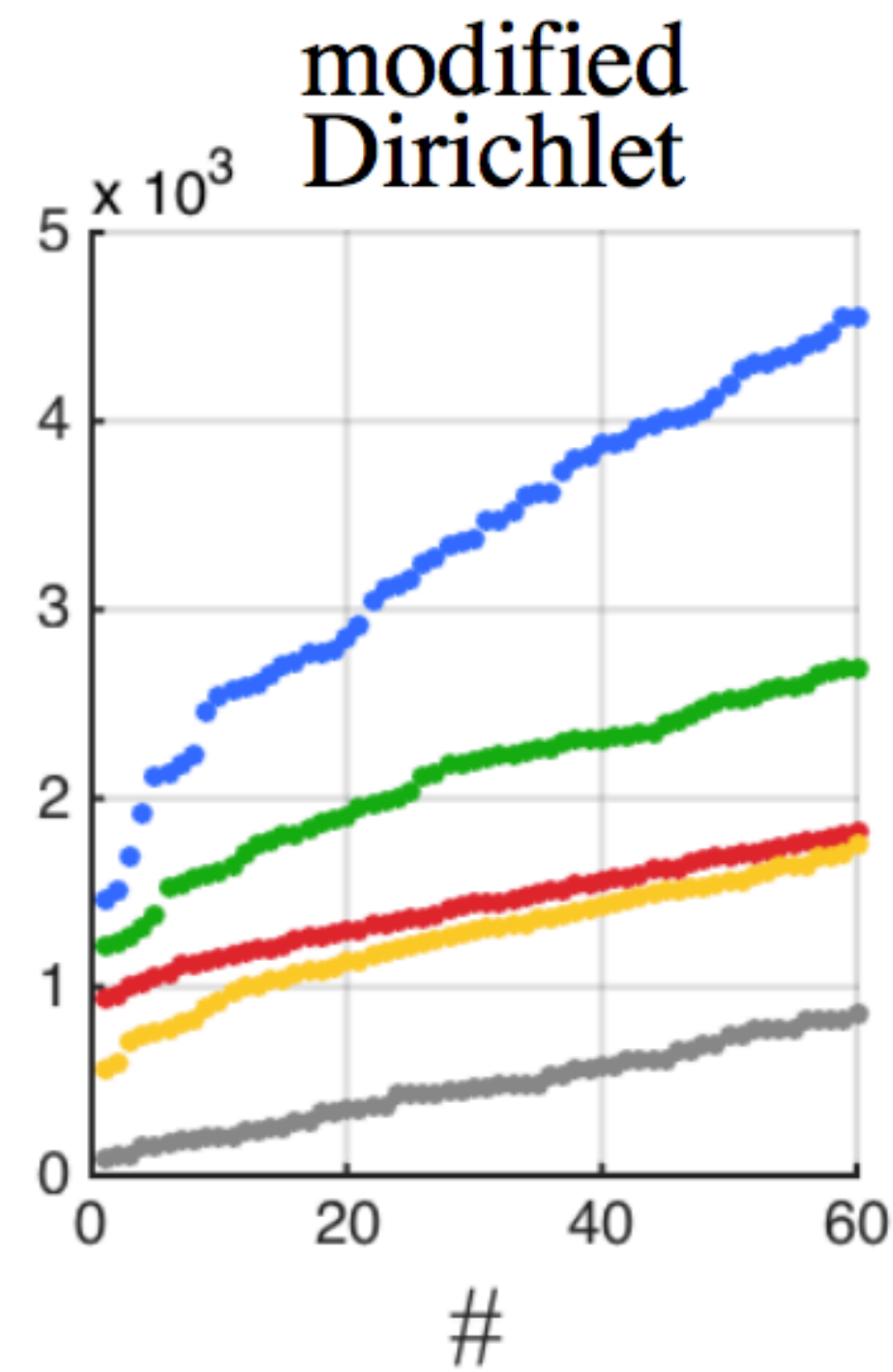
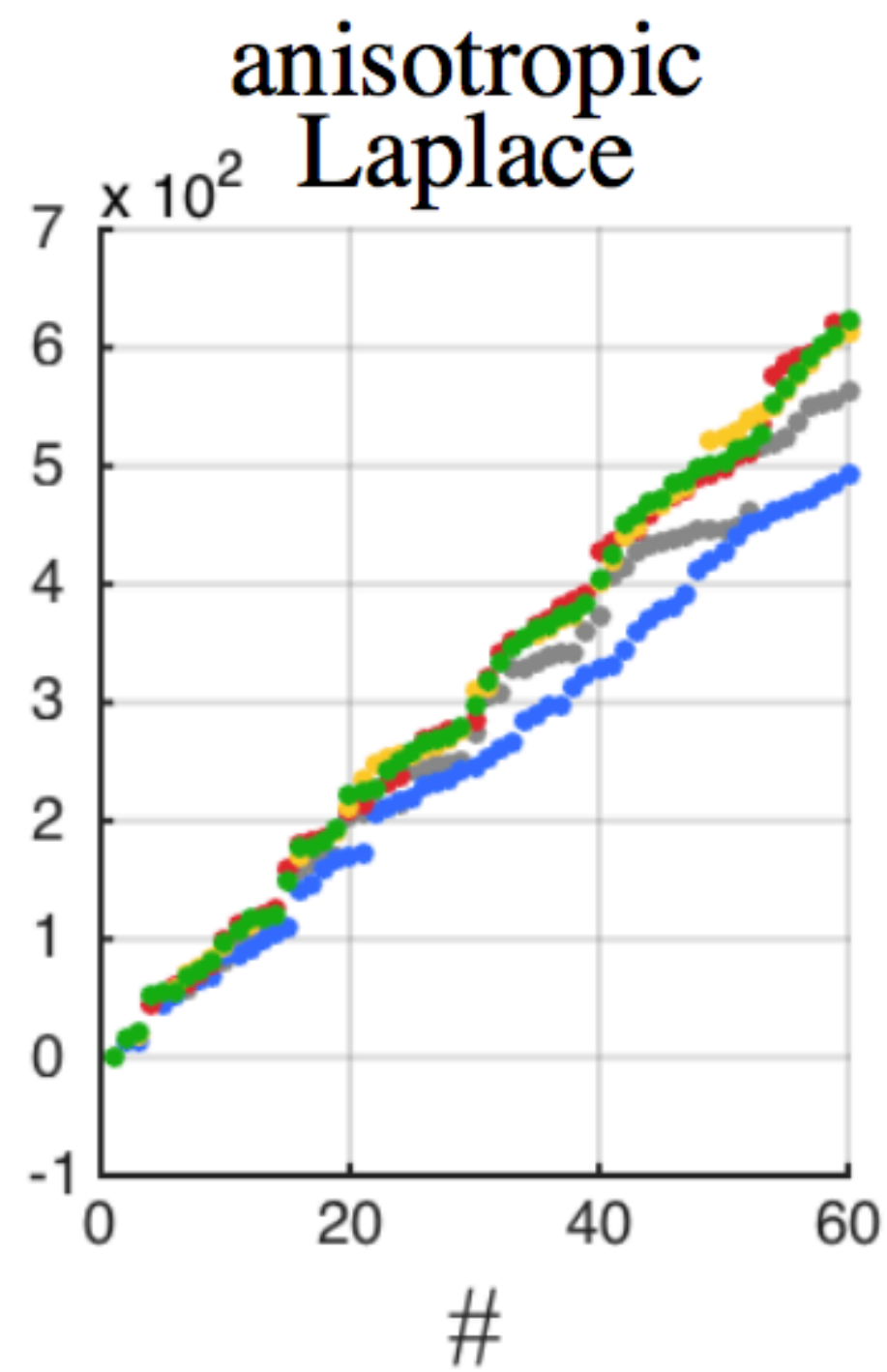
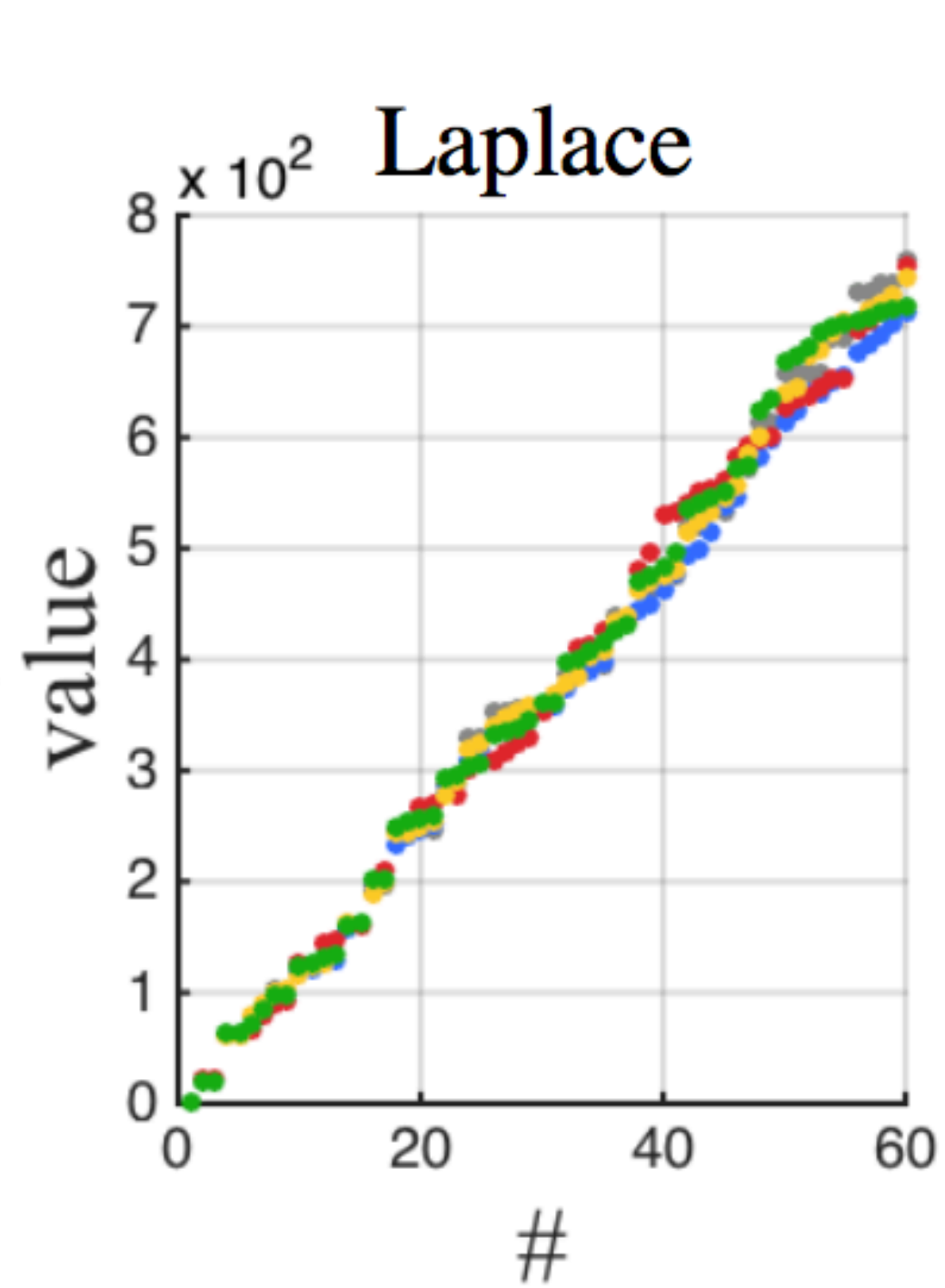
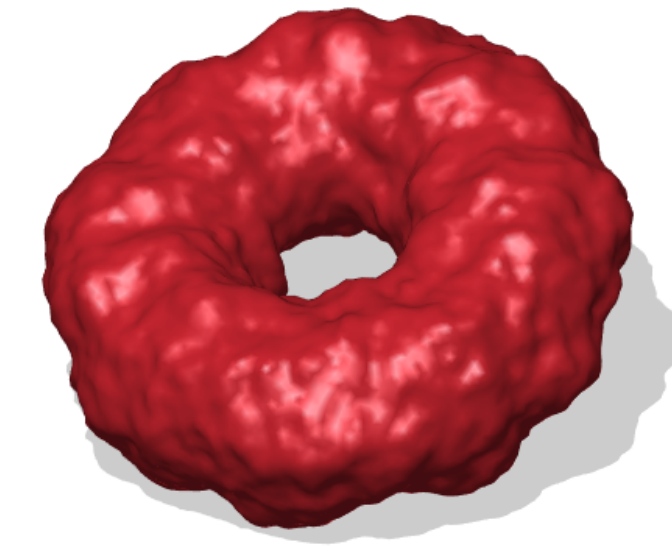
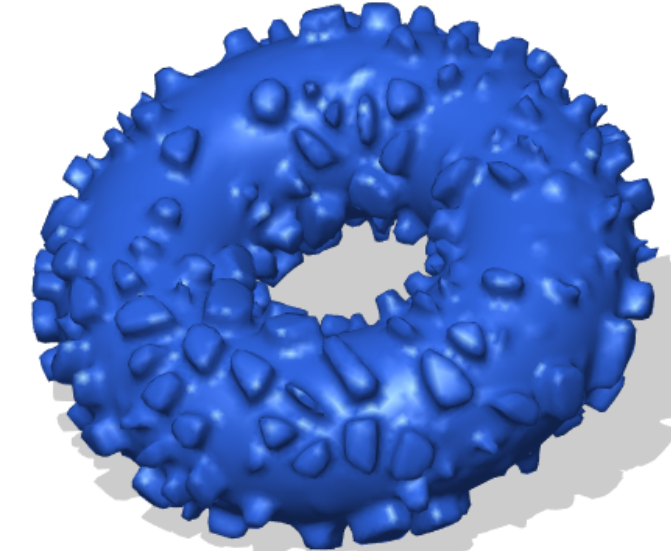
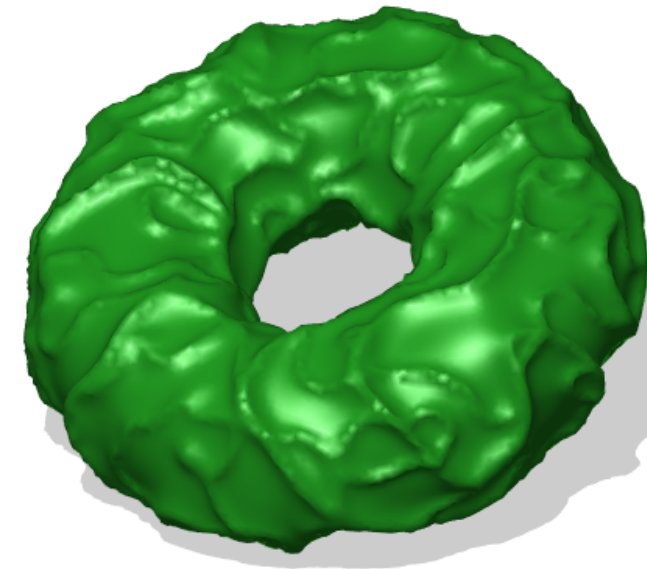
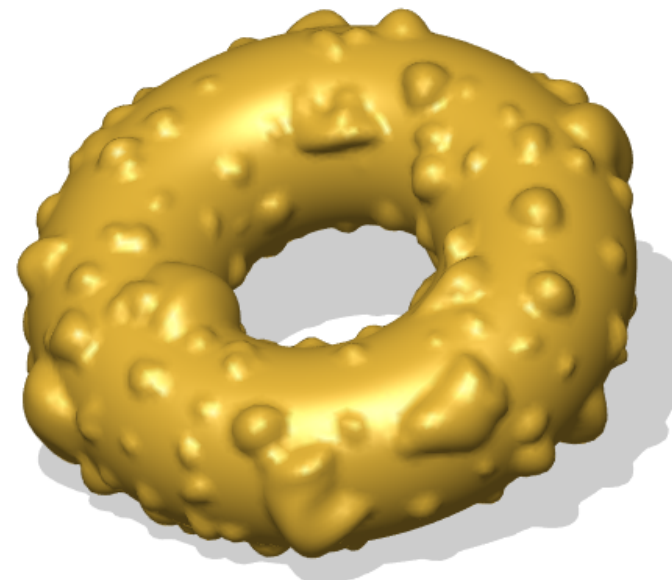
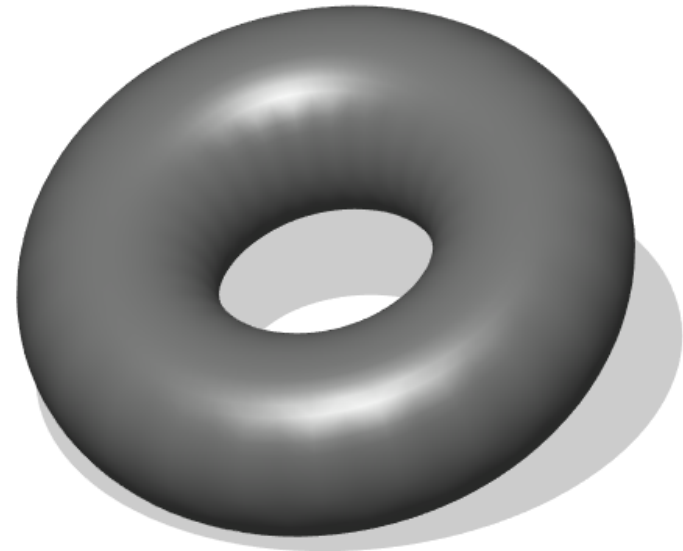


# From Intrinsic to Extrinsic

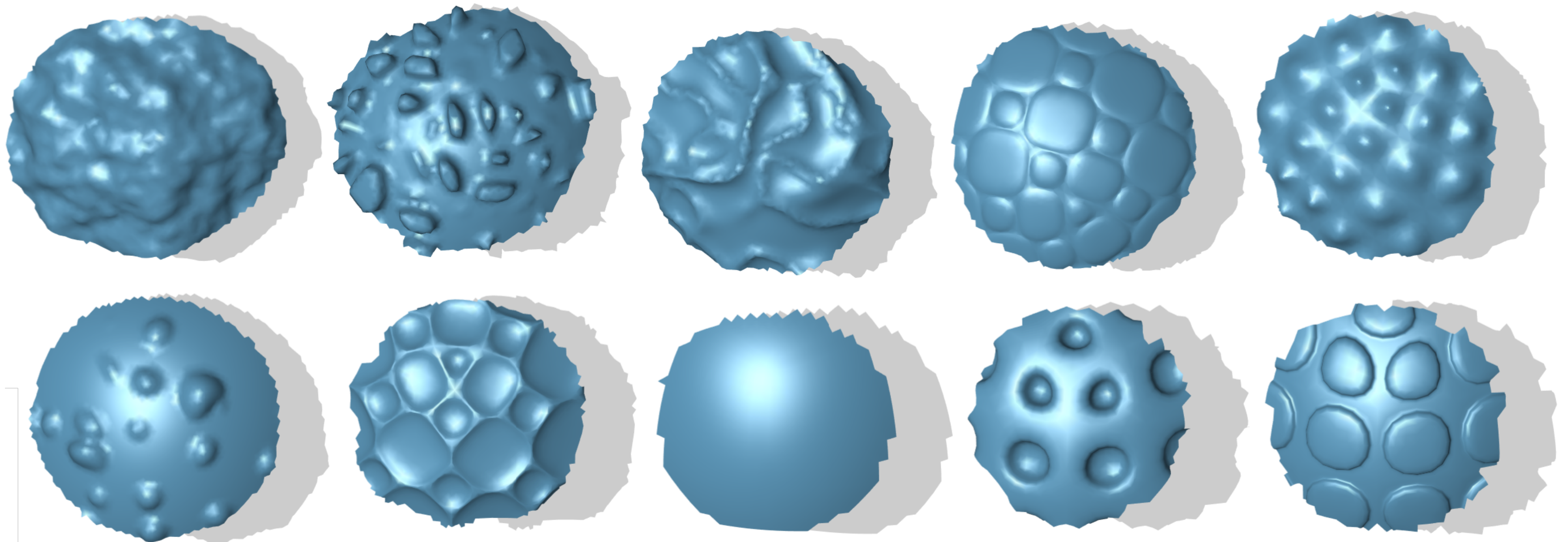


# Applications

# Surface Texture Classification

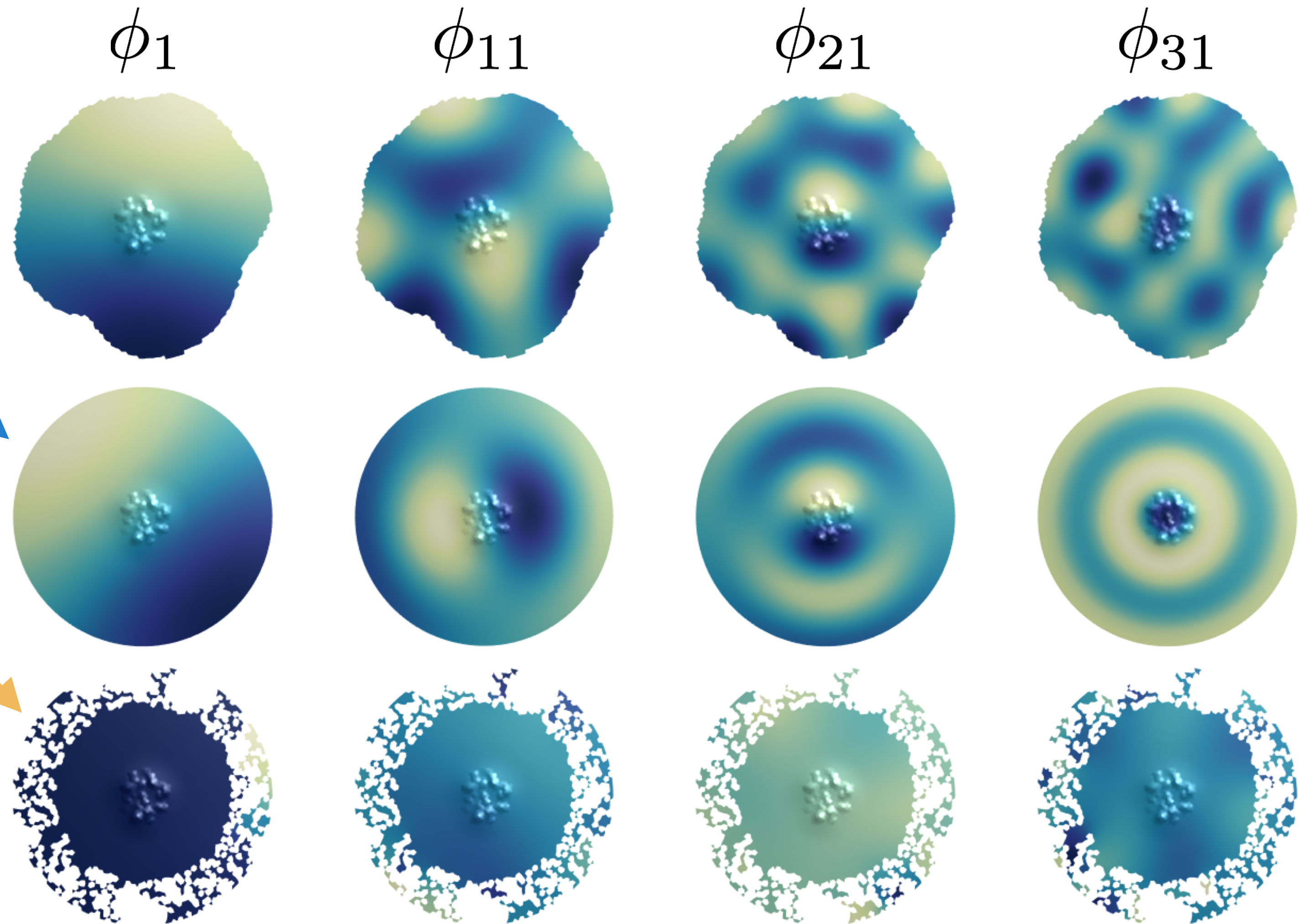
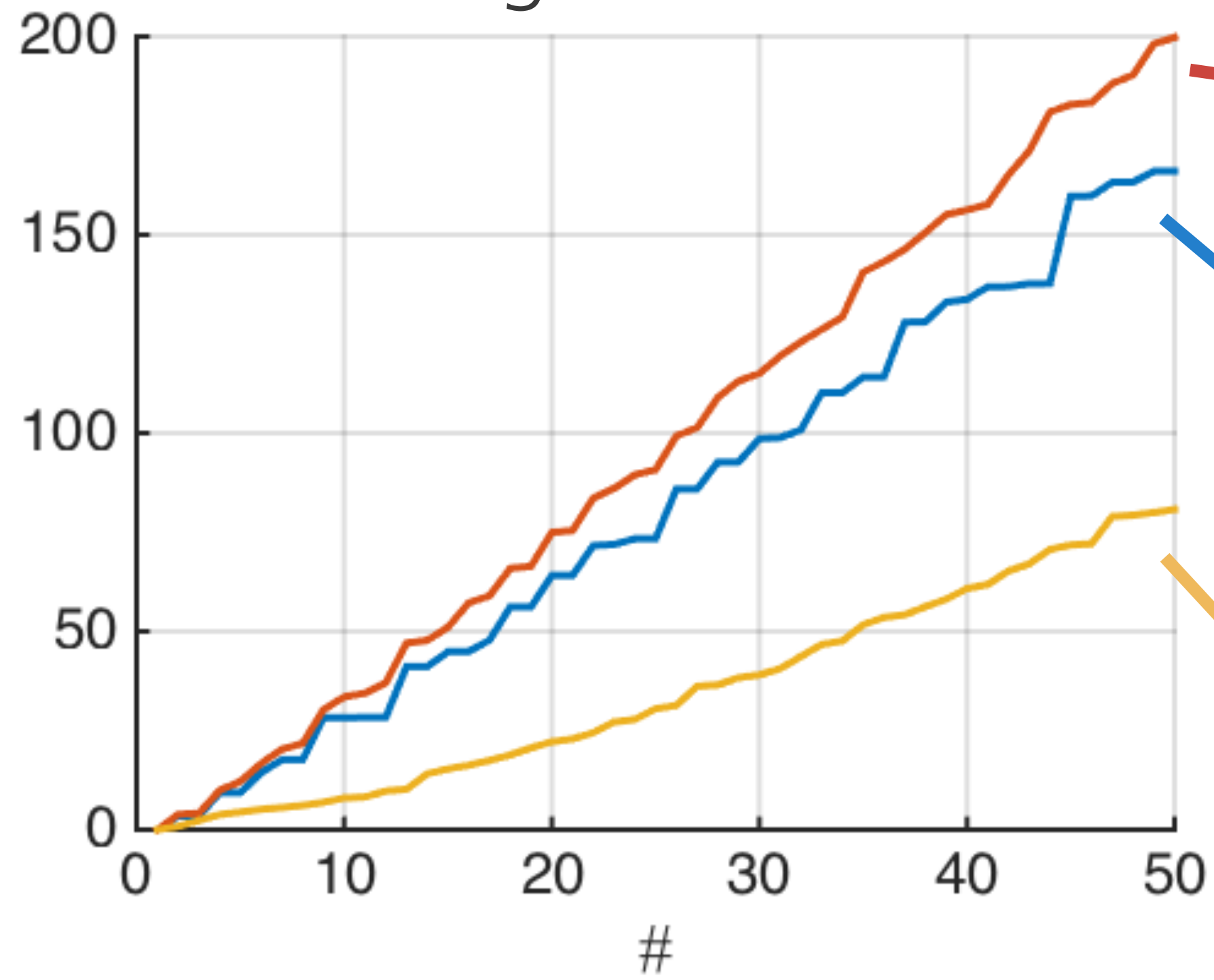


# Patch Classification



# Laplace

eigenvalues



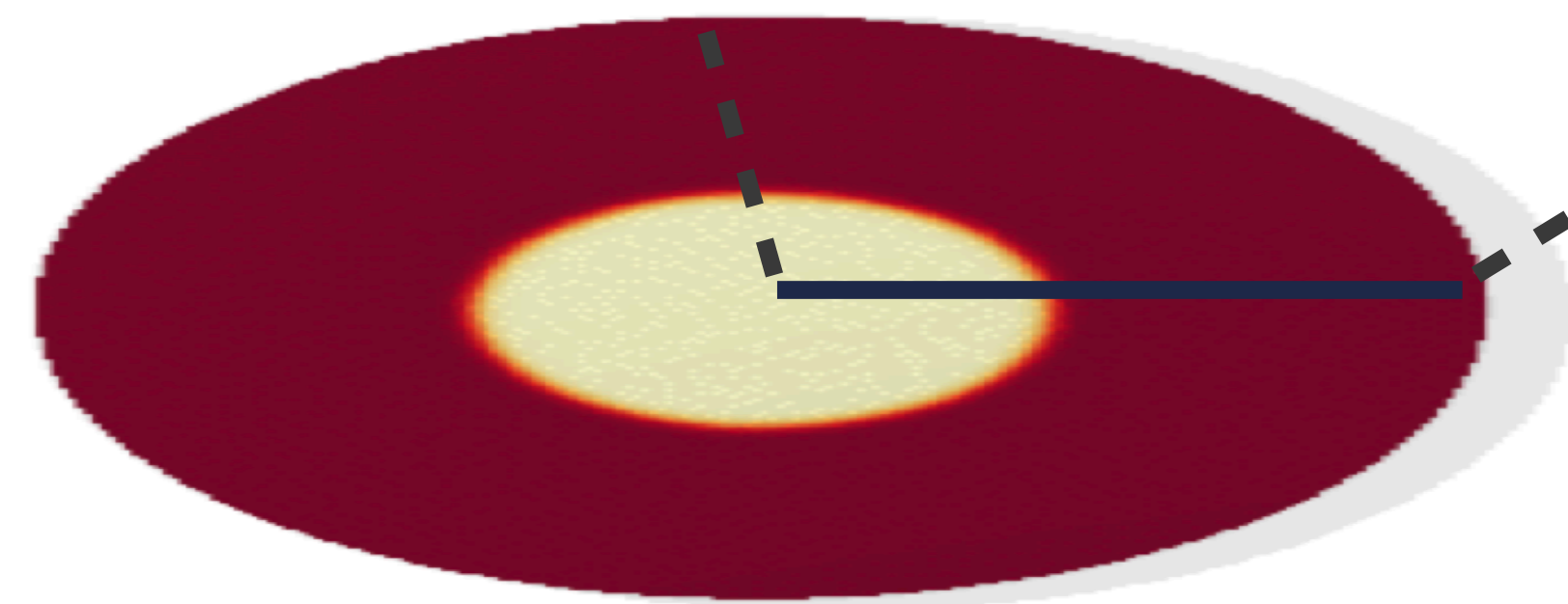
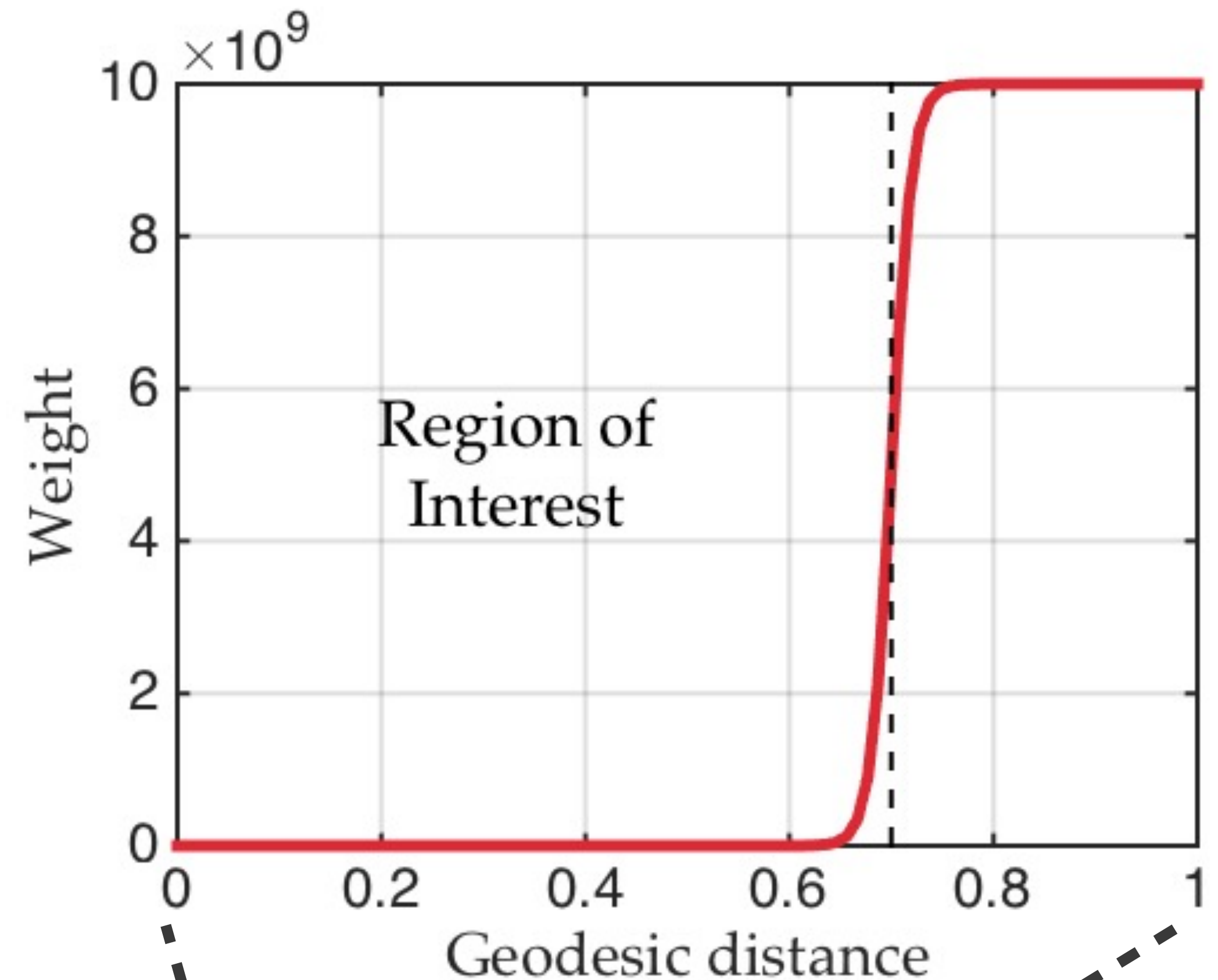
# Infinite Potential Well

- Modified sigmoid function

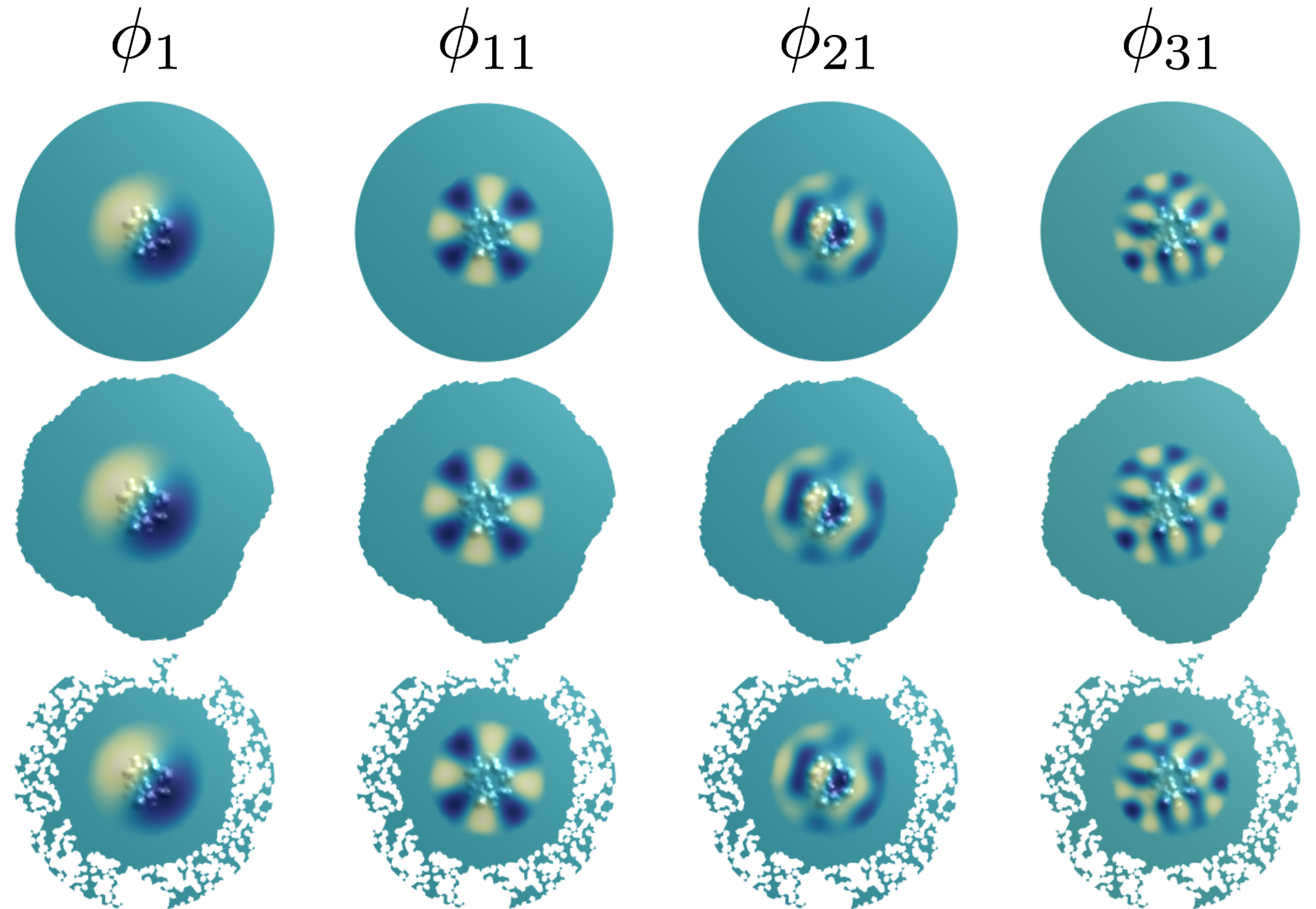
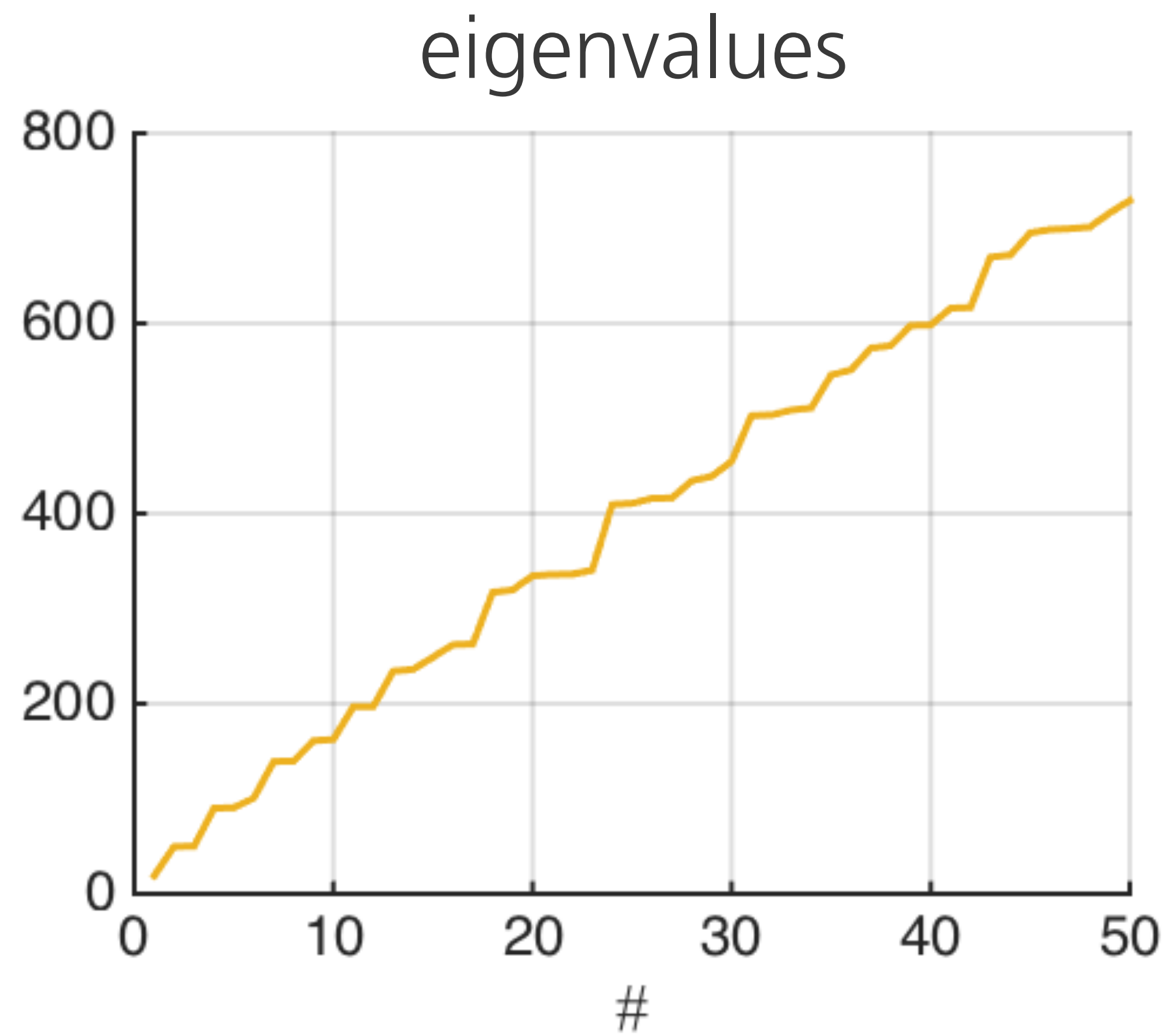
$$U(p) = \frac{c}{1 + (e^{-(d(p,q)-\beta)})^\gamma}$$

- Operator with potential well

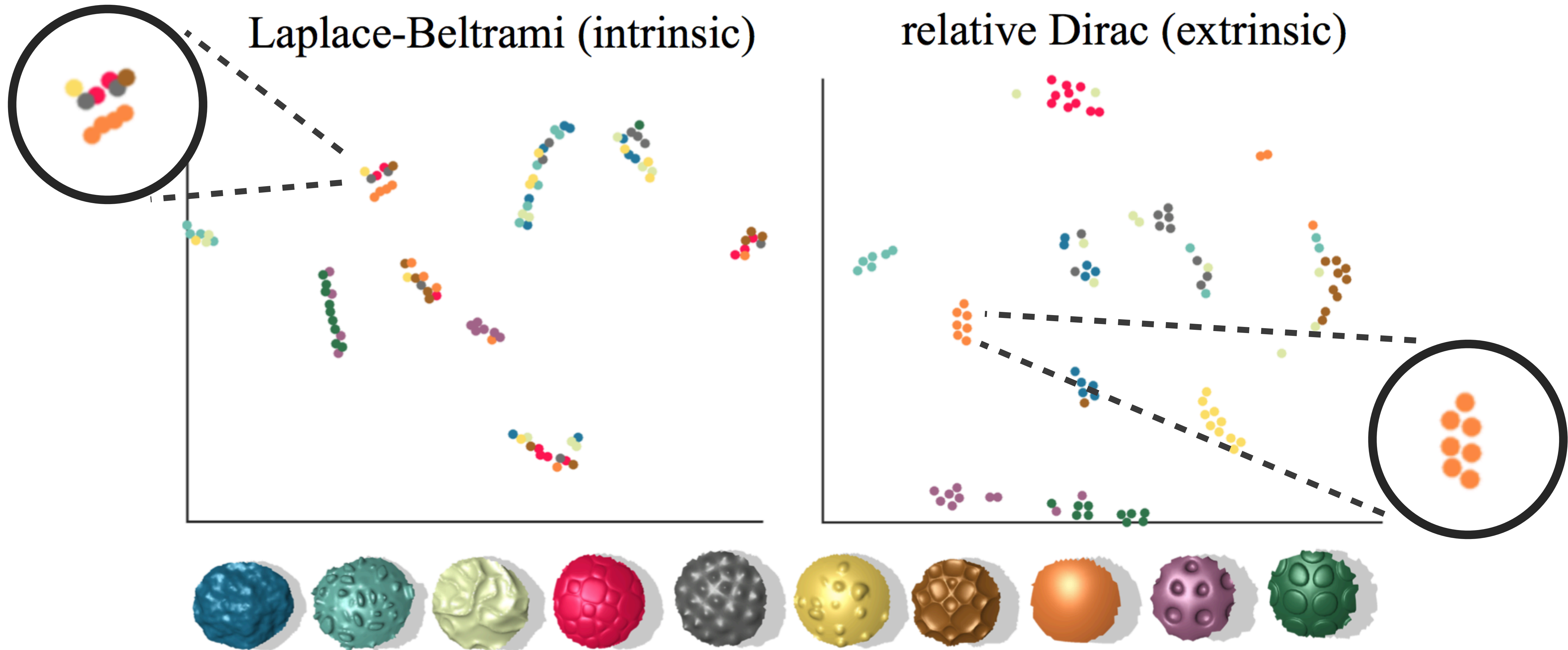
Ex:  $\Delta \rightarrow \Delta + U$



# Laplacian with Infinite Potential Well



# Patch Classification





# Segmentation

Step 1: Adapt **global point signature** to the magnitude of Dirac

eigenvectors

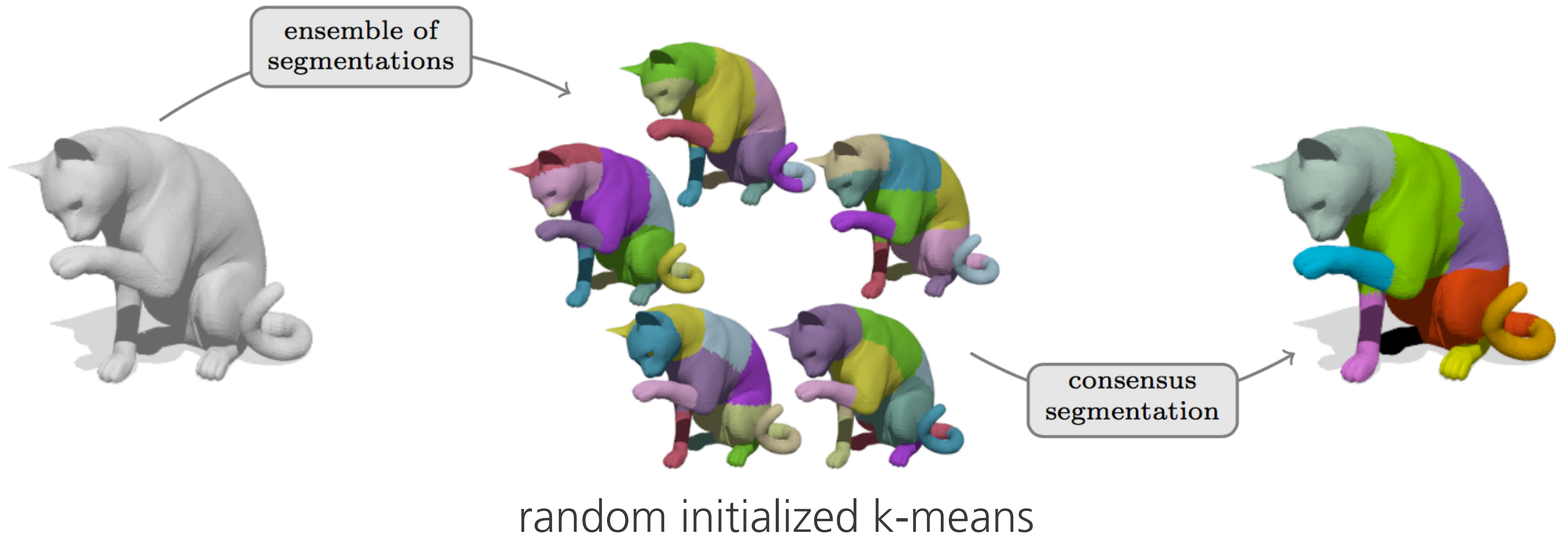
$$v \mapsto \left( \frac{|\phi_1(v)|}{\sqrt{\lambda_1}}, \frac{|\phi_2(v)|}{\sqrt{\lambda_2}}, \frac{|\phi_3(v)|}{\sqrt{\lambda_3}}, \dots \right)$$

point on the surface

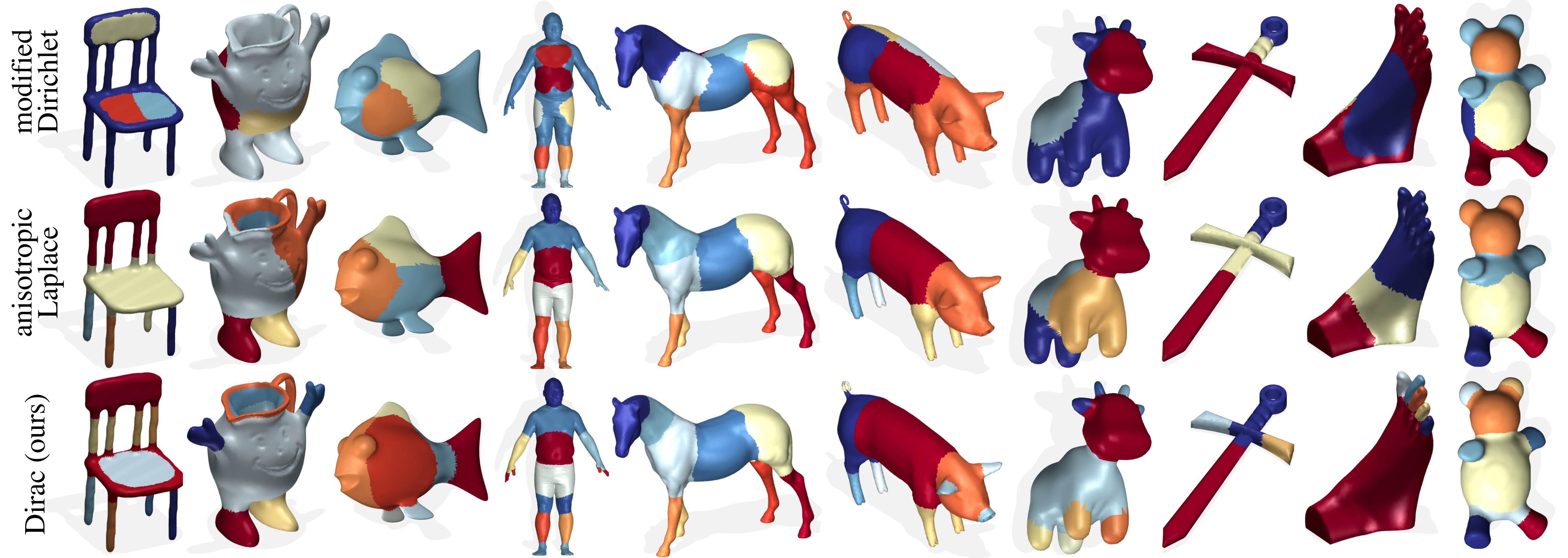
eigenvalues

# Segmentation

Step 2: Apply the **consensus segmentation** algorithm

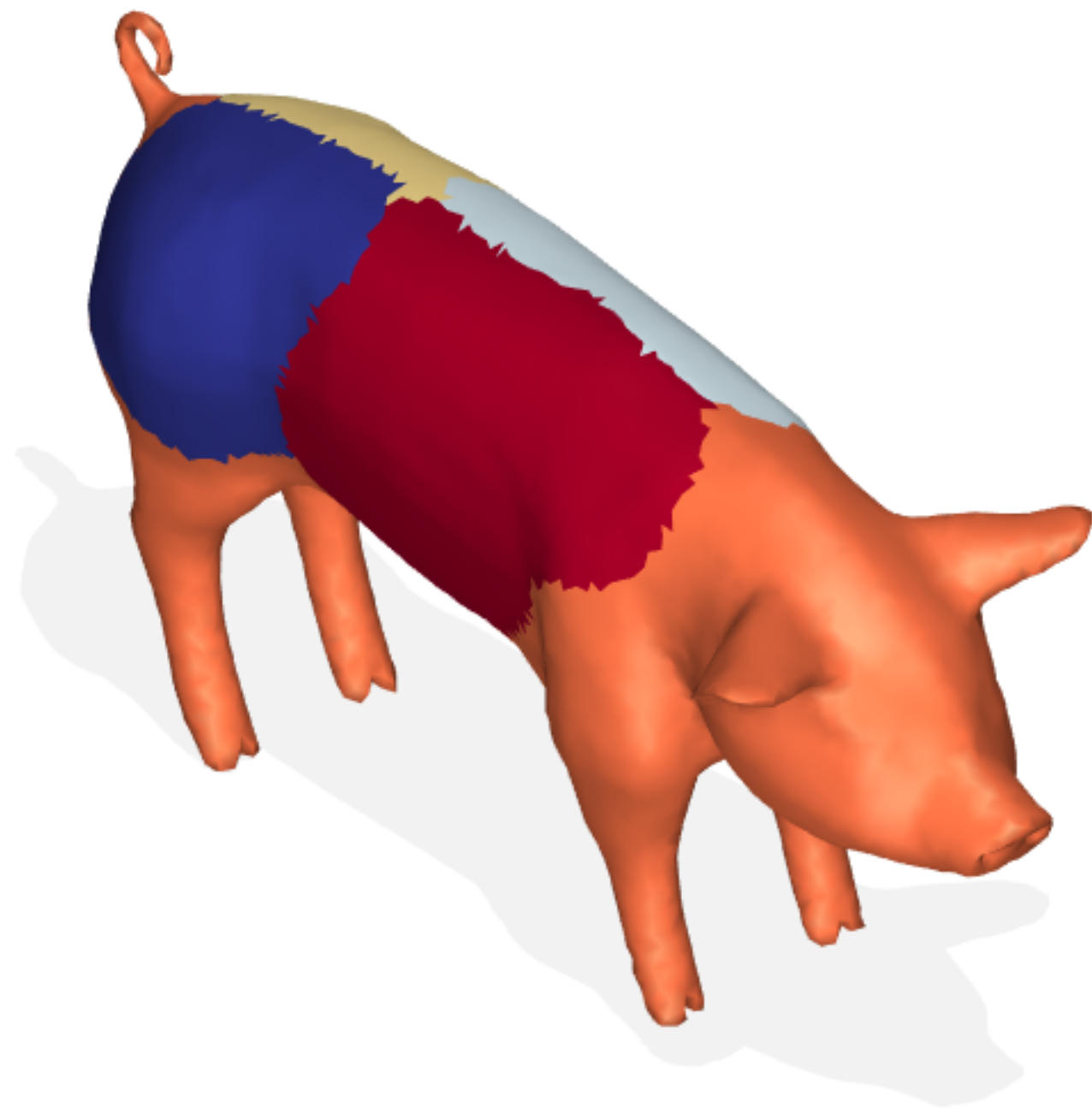


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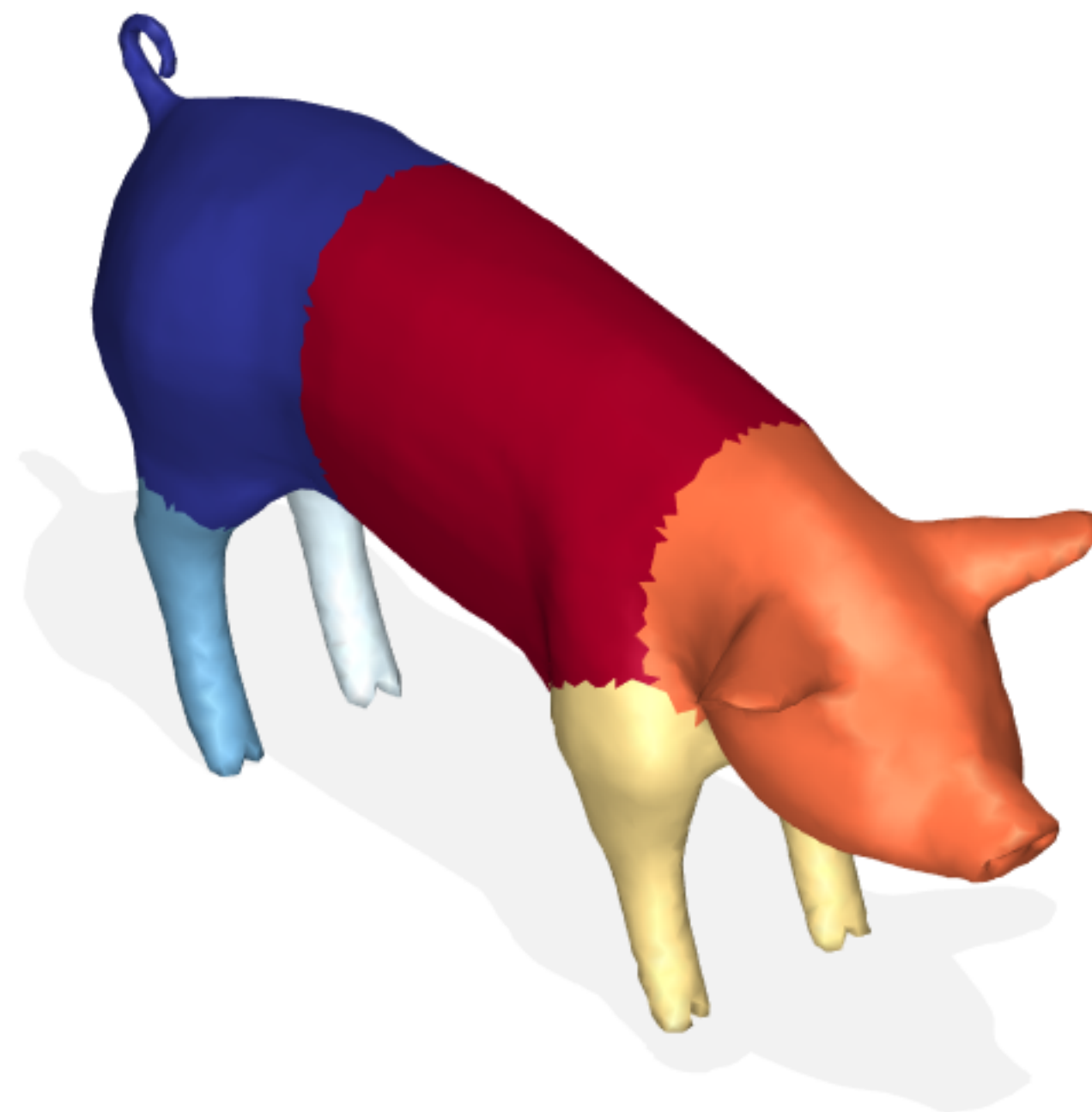


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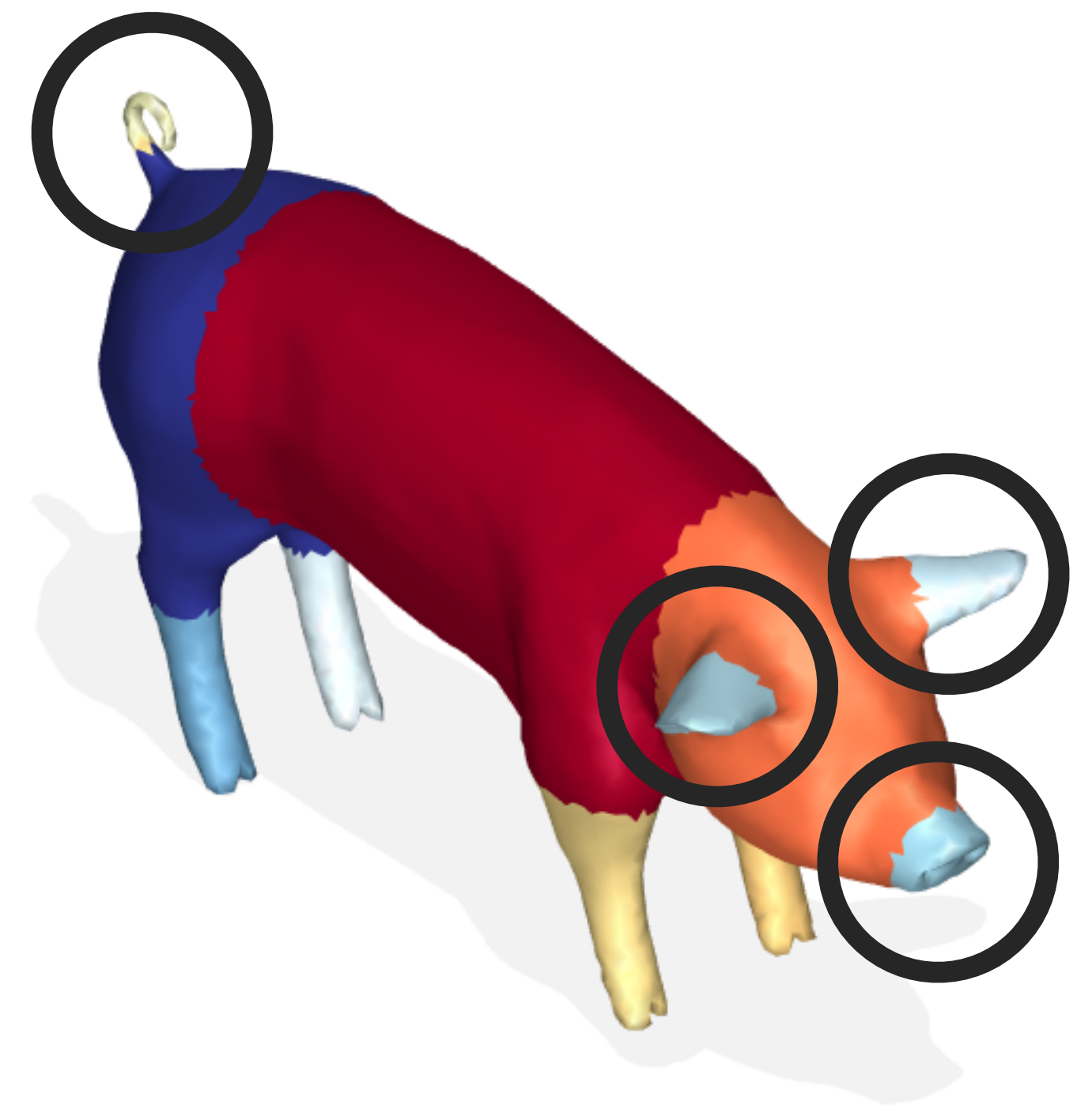
Modified Dirichlet



Anisotropic Laplace

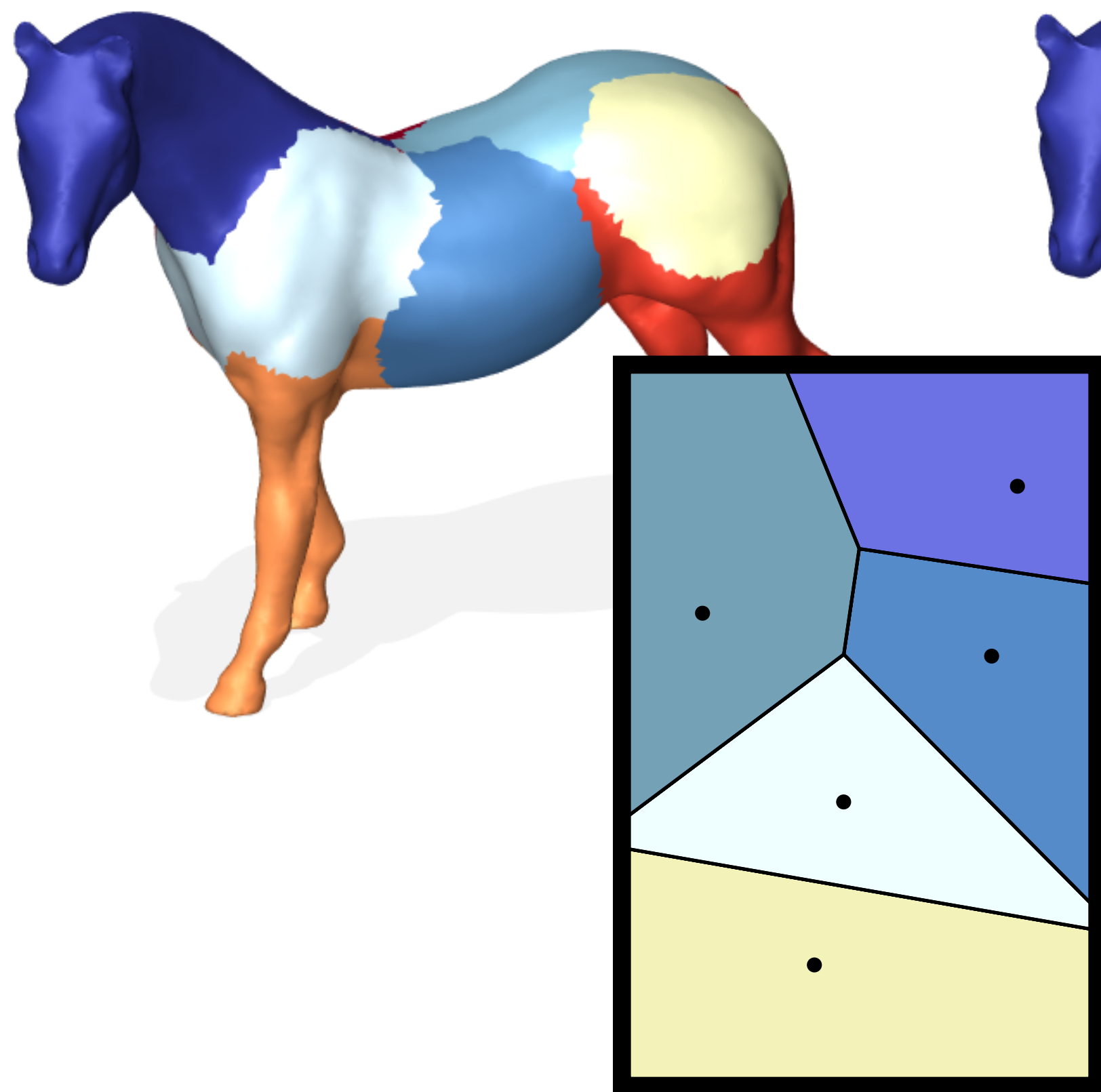


Dirac (ours)

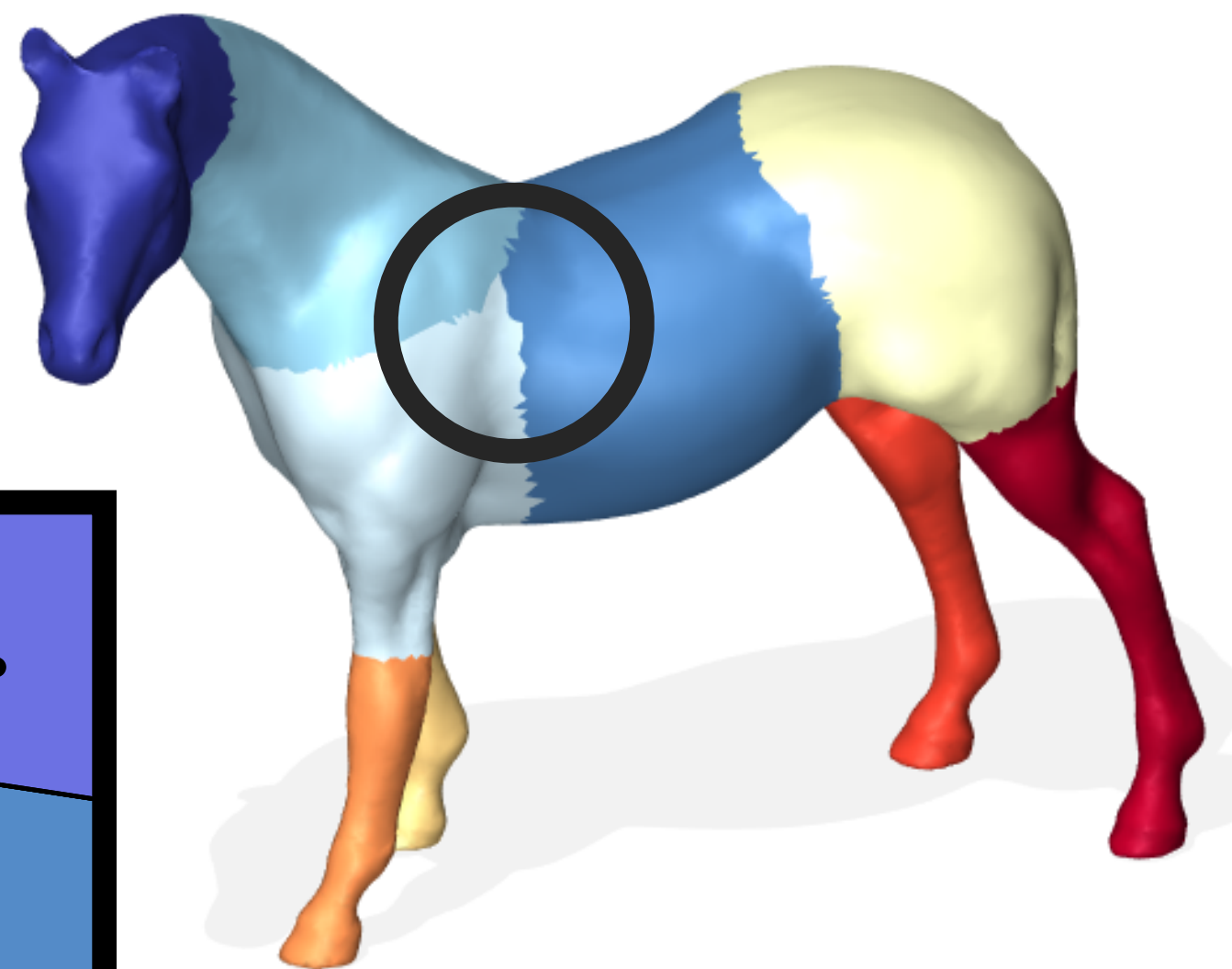


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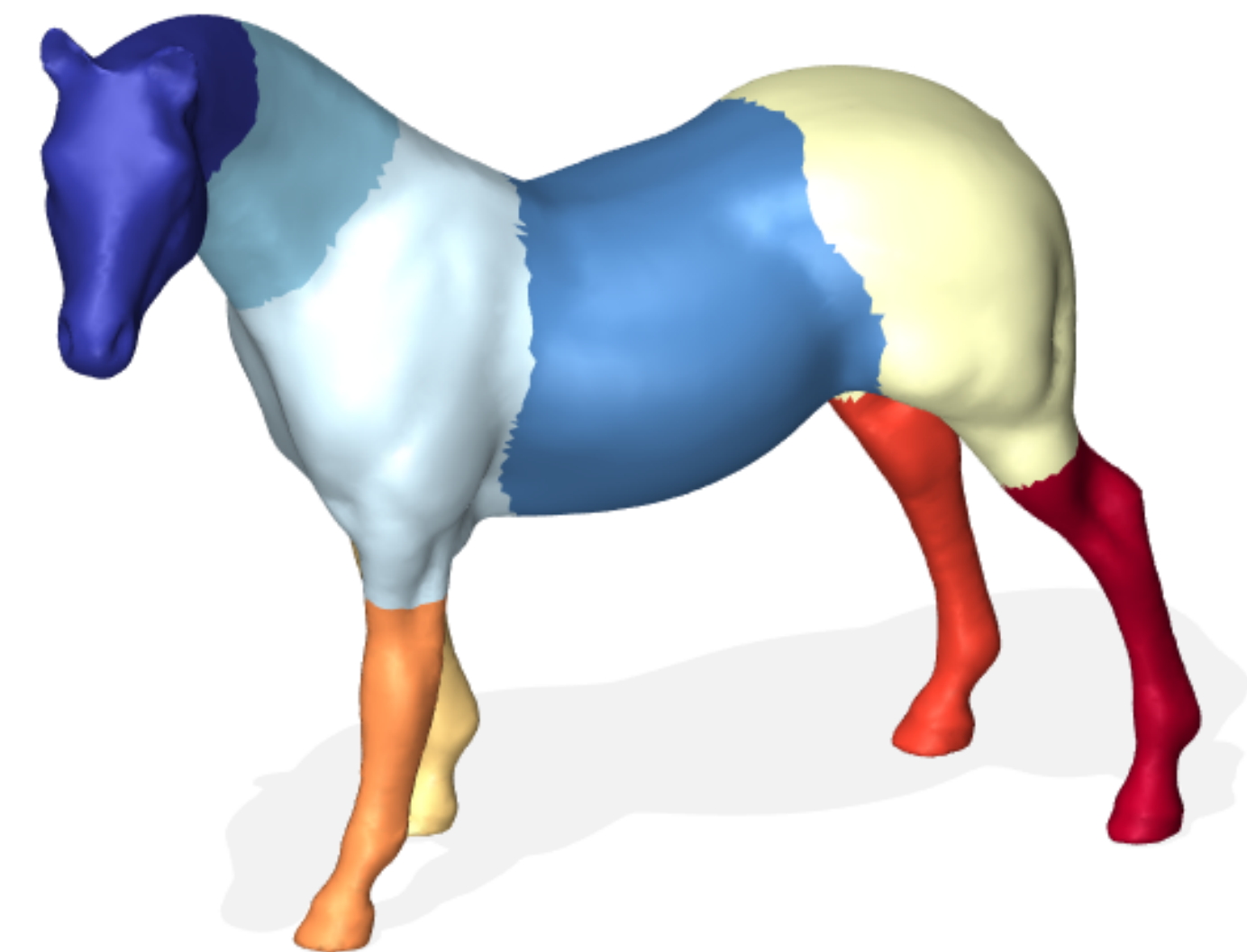
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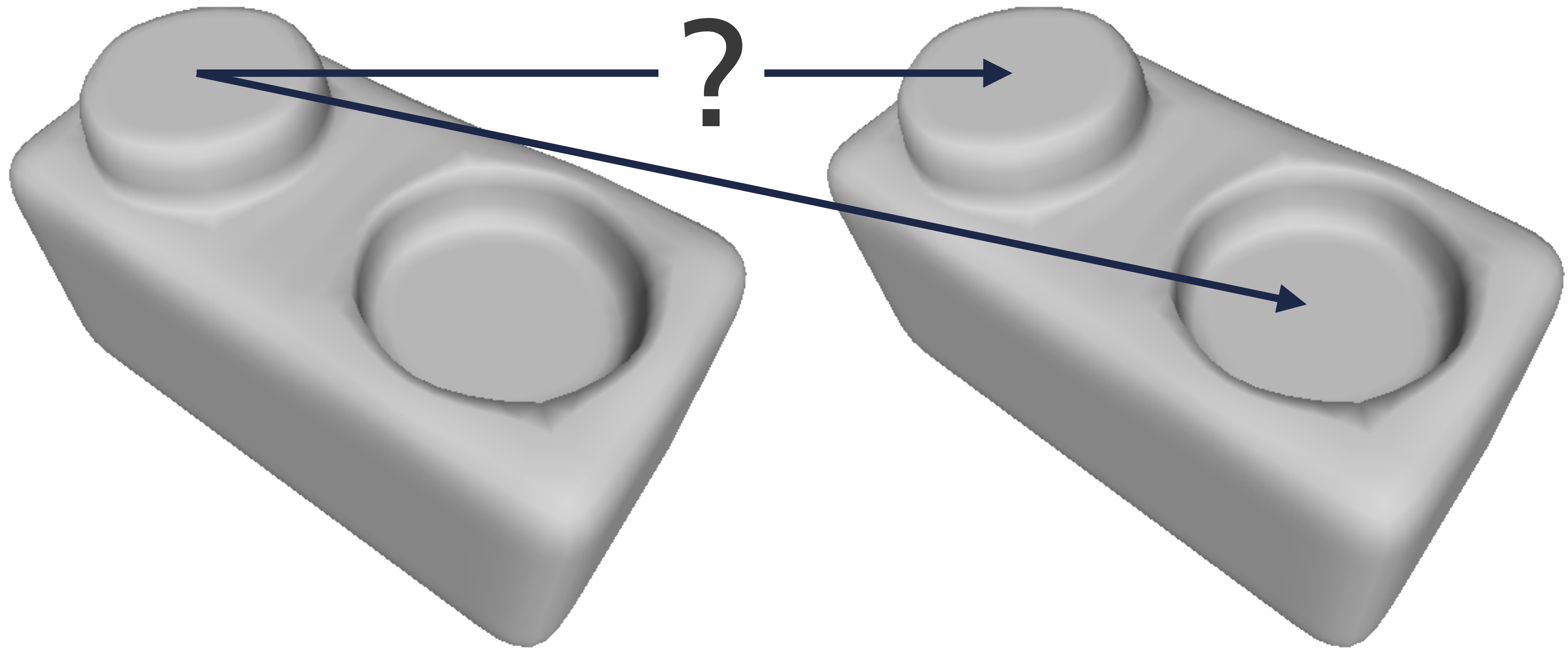


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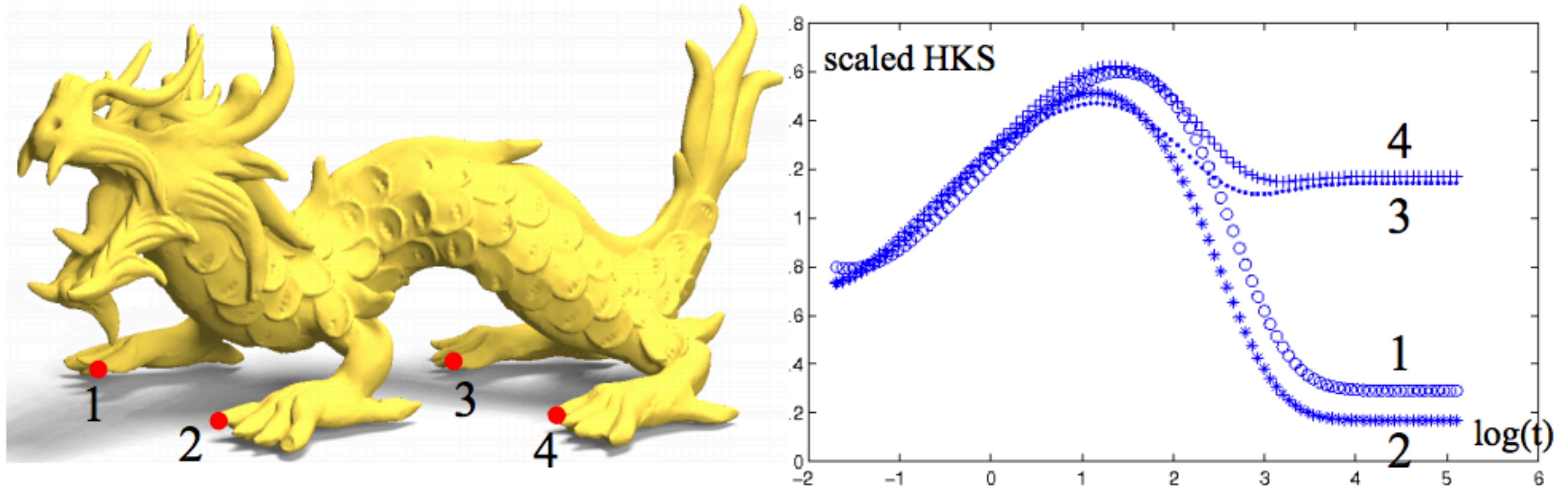


# Correspondence

- Laplace cannot differentiate between bumped out/in



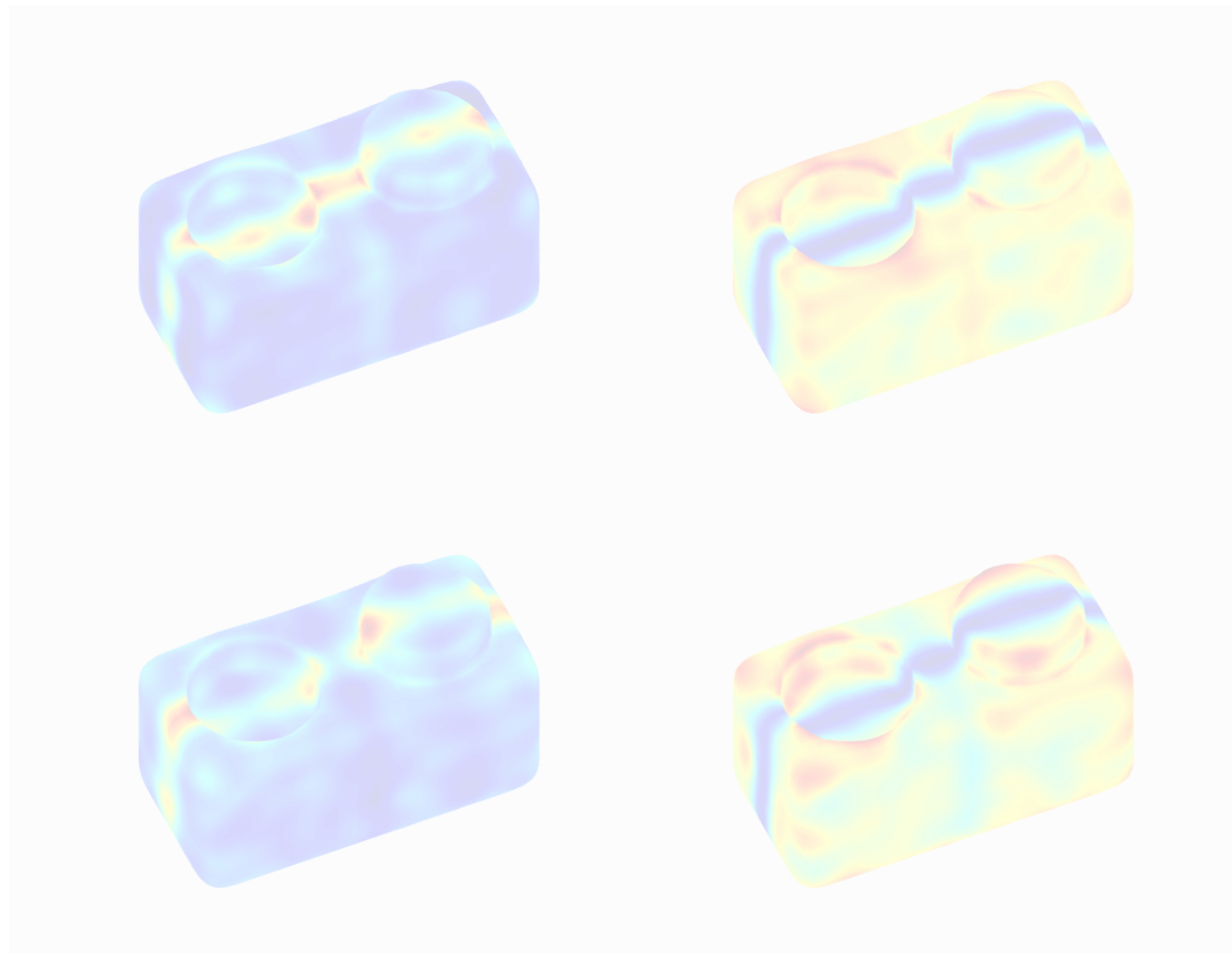
# Heat Kernel Signature



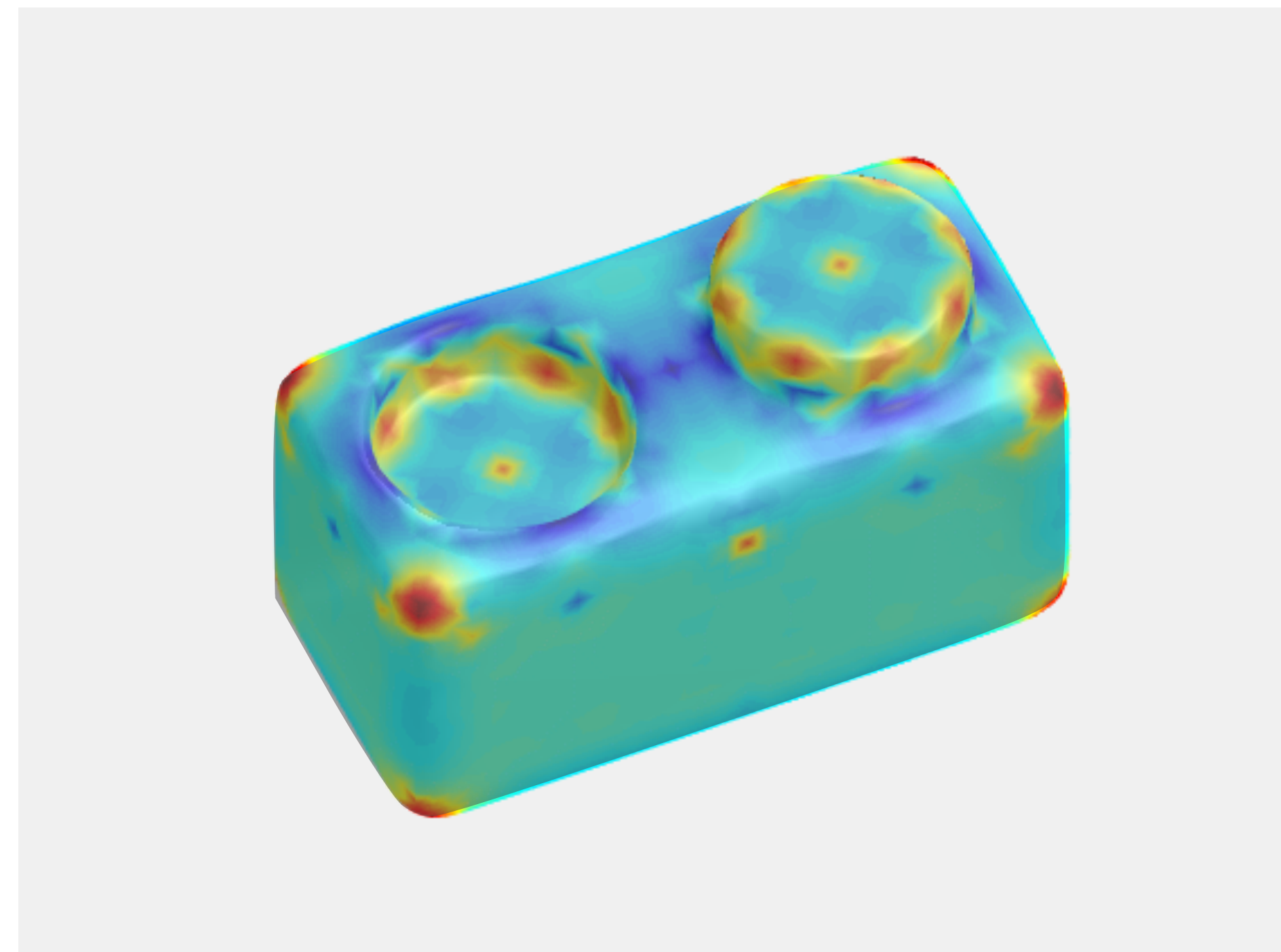
# Correspondence

- Adapt heat kernel signature to the Dirac operator

Dirac kernel signature  $\mathbb{H}$



heat kernel signature  $\mathbb{R}$

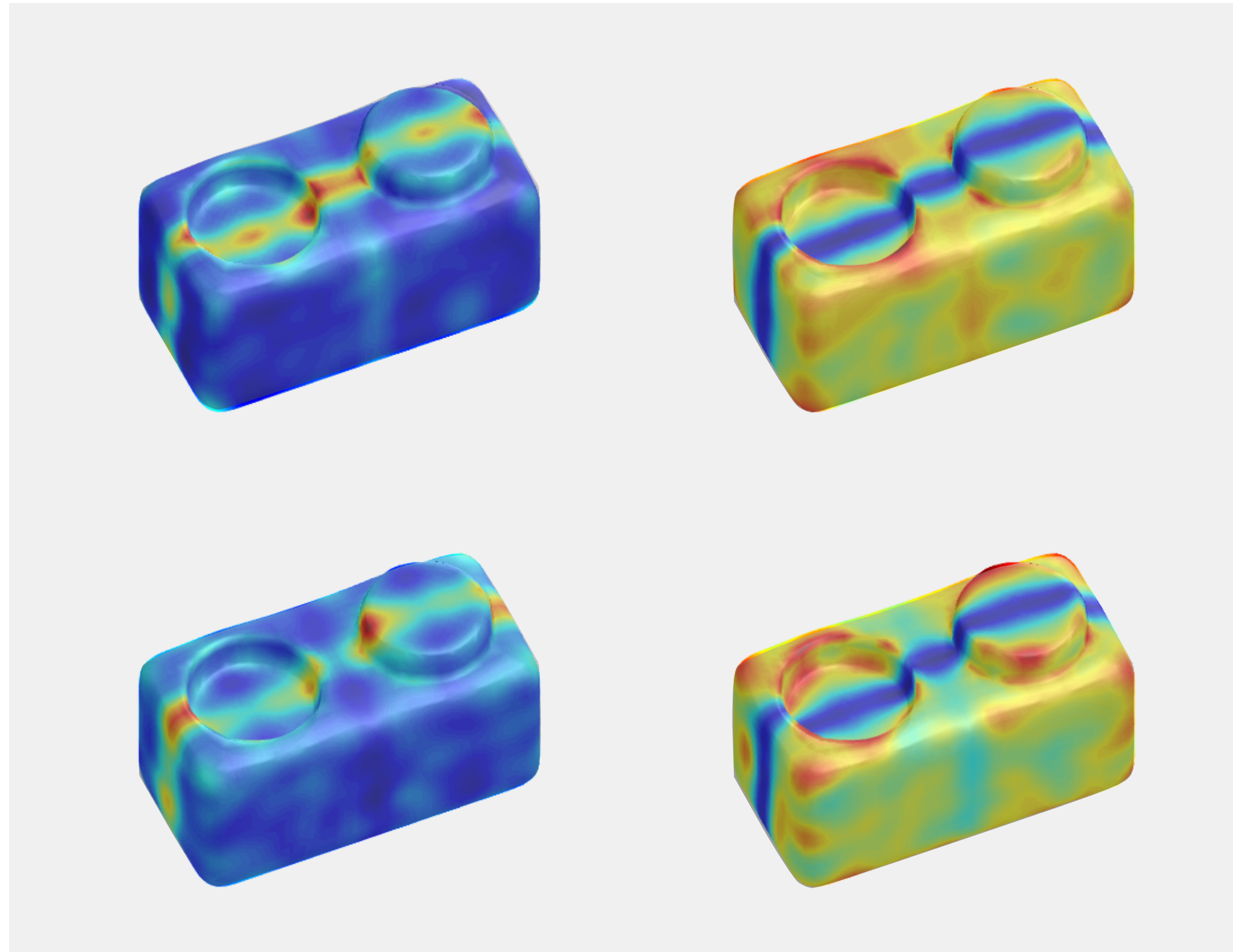




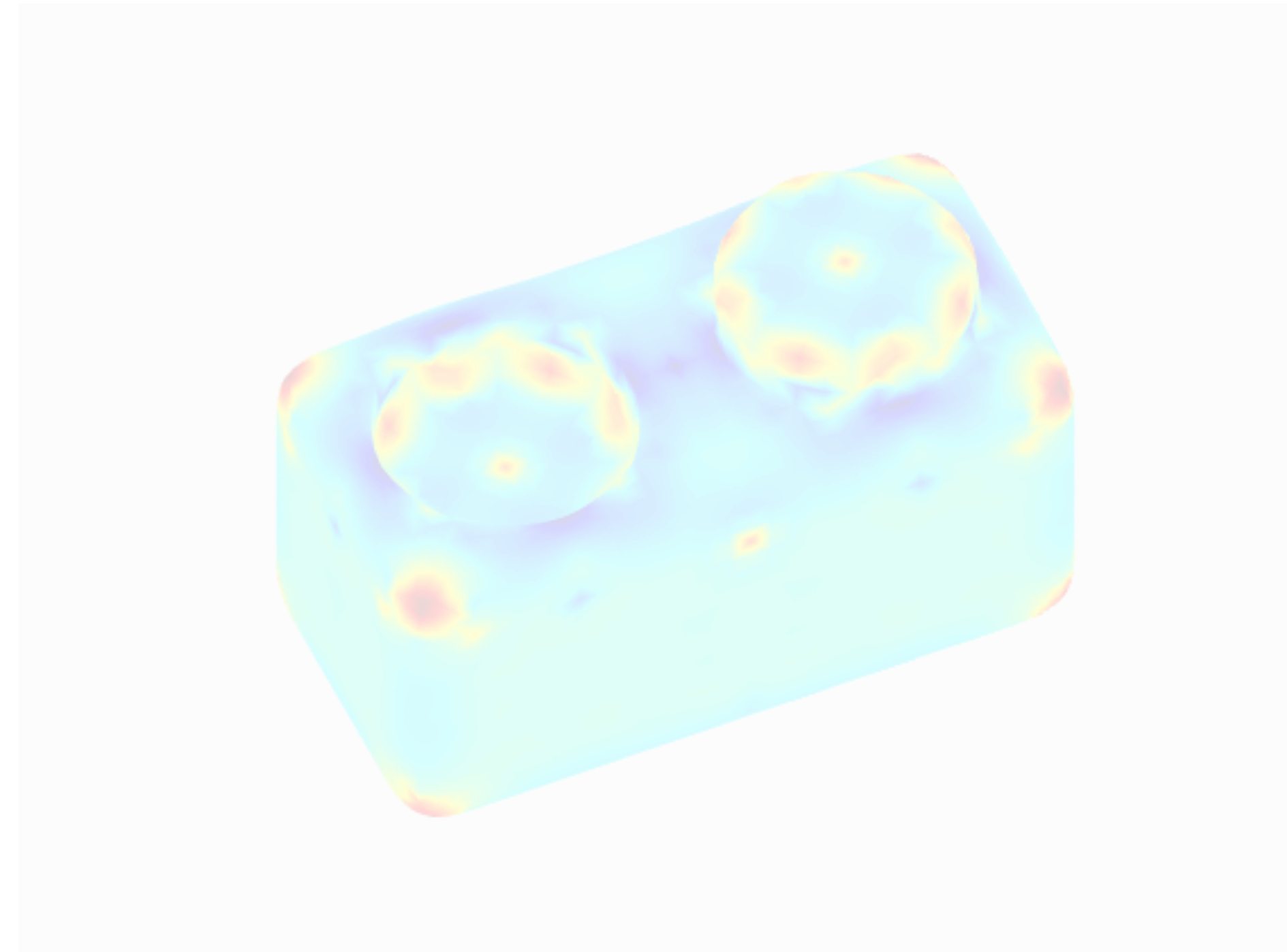
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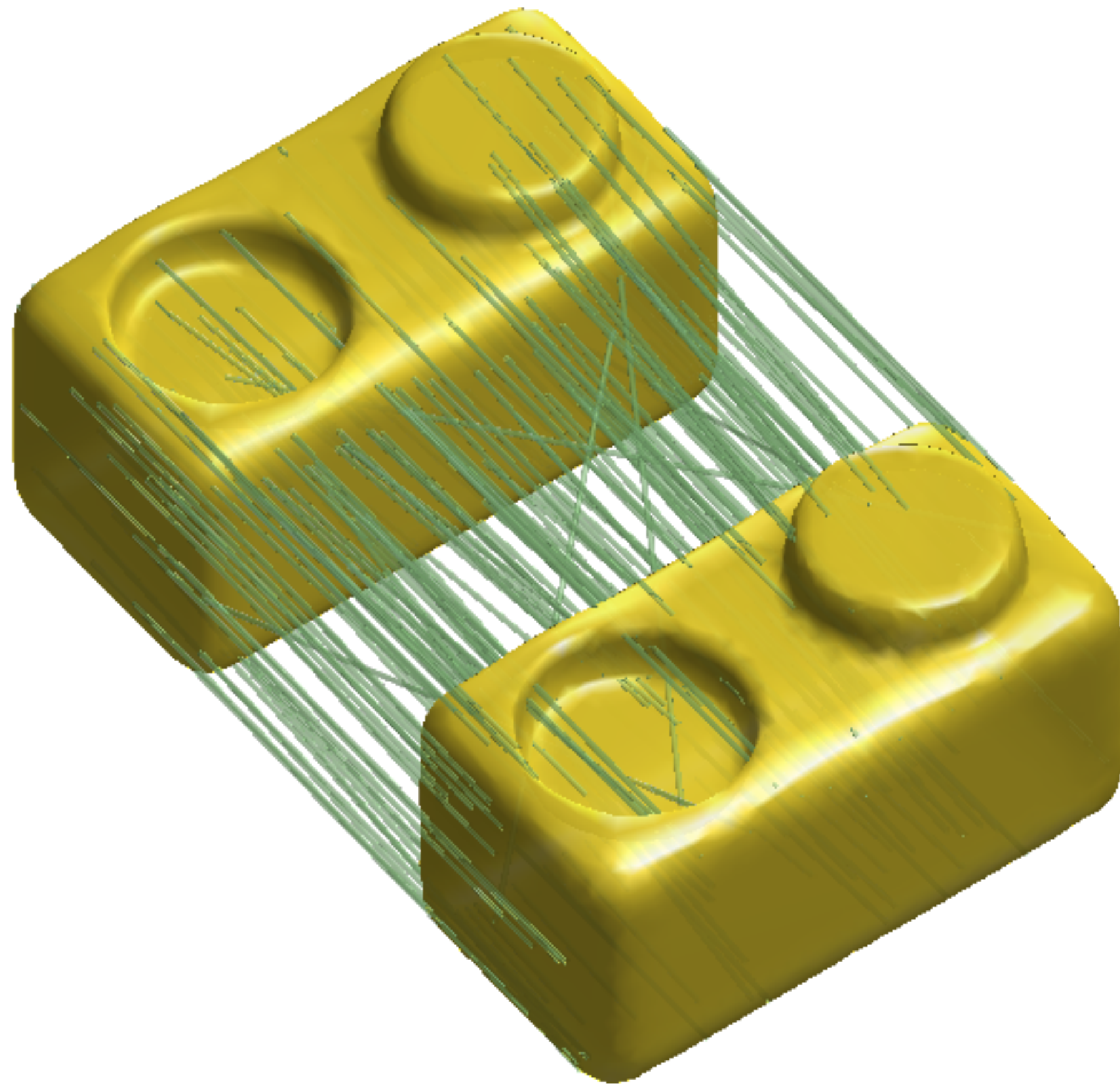


heat kernel signature  $\mathbb{R}$

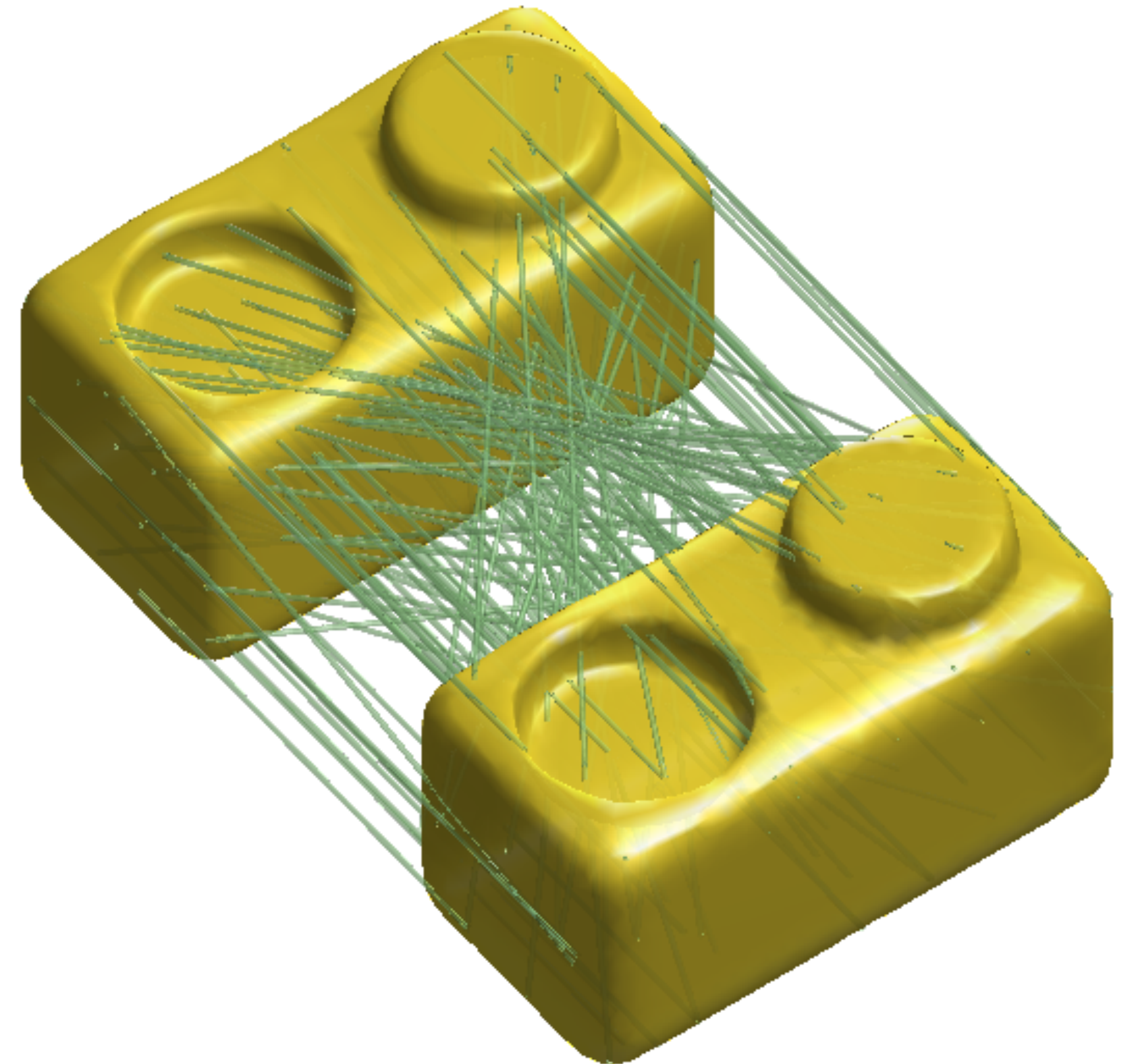


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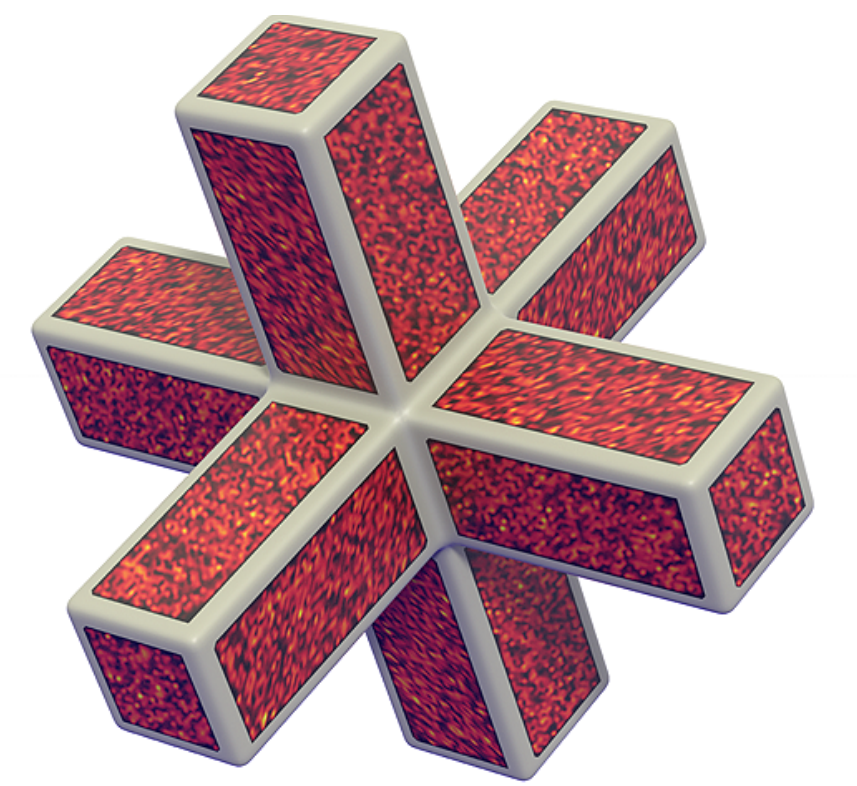
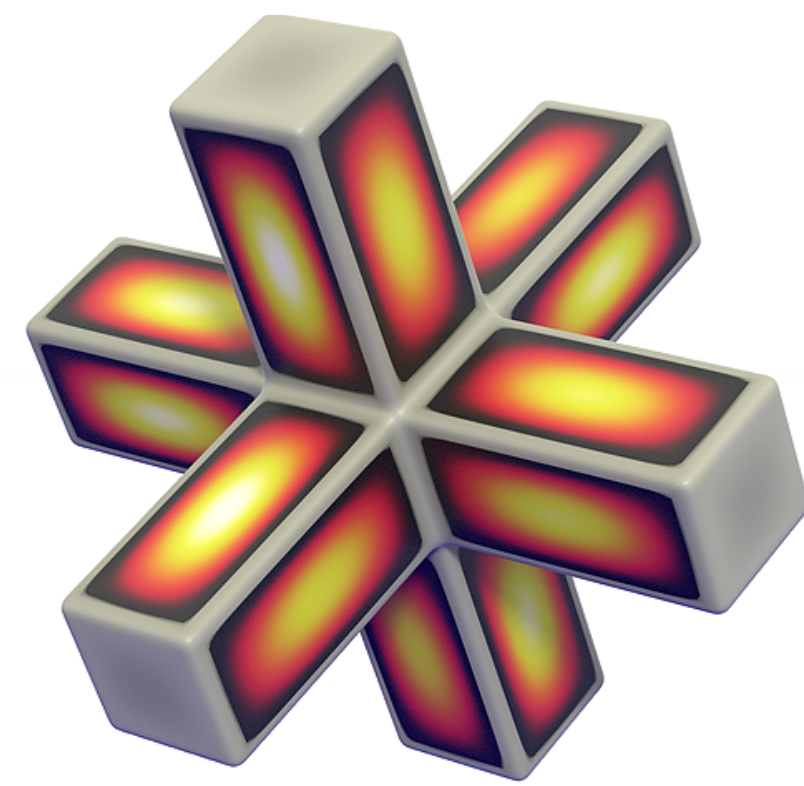
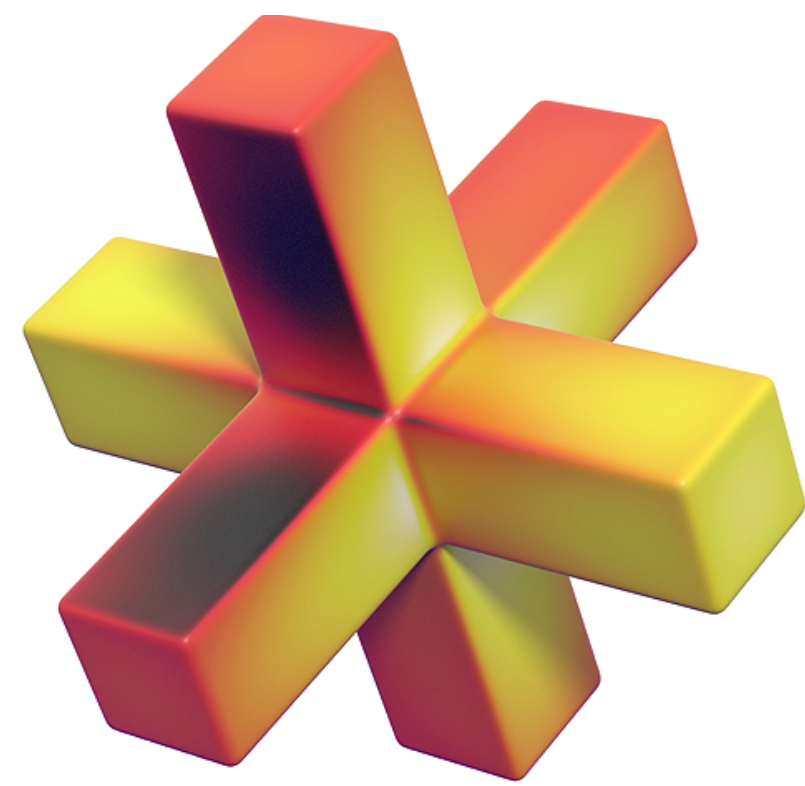
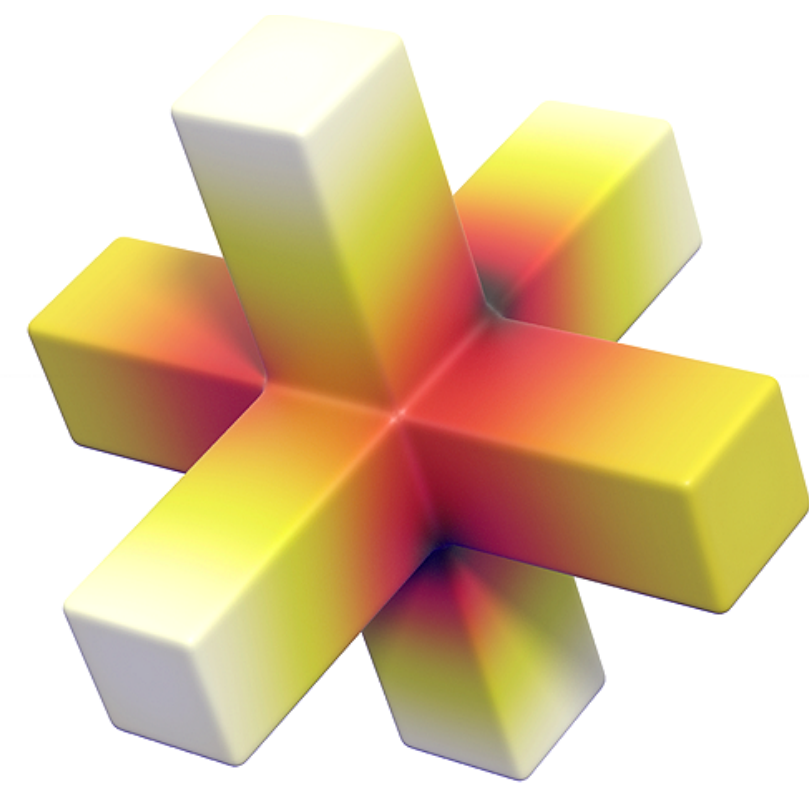
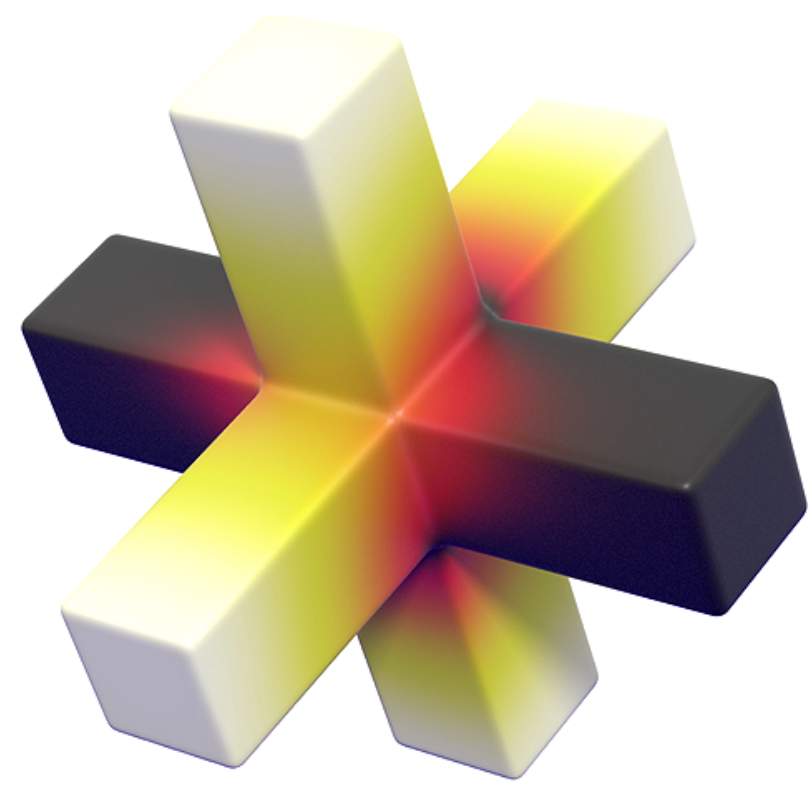
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- What are other operators can we use to capture different geometric quantities?

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<https://github.com/alecjacobson/gptoolbox.git>



Thank you!

Hsueh-Ti Derek Liu, [hsuehtil@andrew.cmu.edu](mailto:hsuehtil@andrew.cmu.edu)