

Tutorial

Inverse Spectral Geometry

2/4

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EUROGRAPHICS 2021

Outline

- Laplacian eigenvalues and eigenvectors
- Fourier analysis on manifolds
- Functional maps
- Shape difference operators
- Shape-from-operator inverse problems

Entdeckungen
über die
Theorie des Klanges

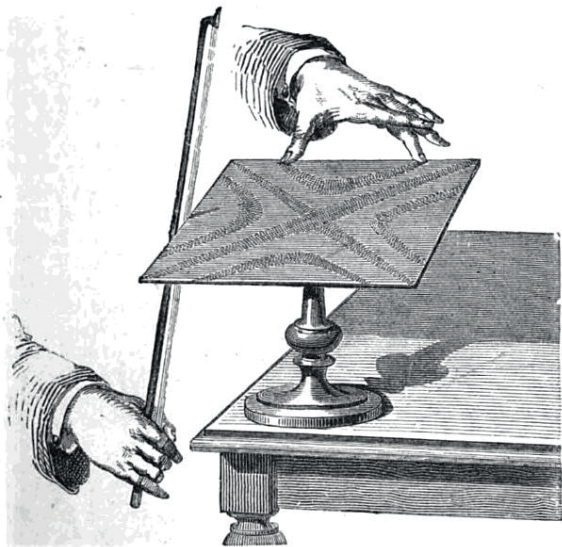
von

Ernst Florens Friedrich Chladni,
der Philosophie und Rechte Doctor zu Wittenberg.

Mit elf Kupfertafeln.

Leipzig,
bey Weidmanns Erben und Reich.

1787.



Ernst Chladni (1756-1827)

Interpretation of Chladni plates

Behavior of waves on the plate is guided by the **wave equation**

$$f_{tt}(x, t) = -\Delta f(x, t)$$

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Spatial part of the solution satisfies the **Helmholtz equation**

$$\Delta\varphi(x) = \lambda\varphi(x)$$

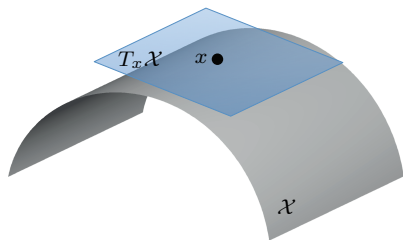
$\lambda =$ vibration frequencies

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$\phi =$ vibration modes

Riemannian geometry in one minute

- Manifold \mathcal{X} = topological space
- No global Euclidean structure
- **Tangent plane** $T_x\mathcal{X}$ = local Euclidean representation of manifold \mathcal{X} around x

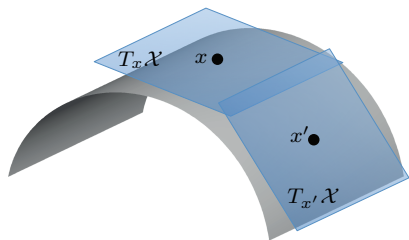


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$$\langle \cdot, \cdot \rangle_x : T_x\mathcal{X} \times T_x\mathcal{X} \rightarrow \mathbb{R}$$

depending smoothly on x



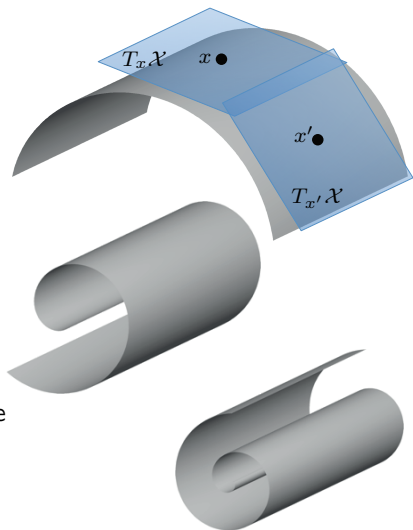
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Isometry = metric-preserving shape deformation



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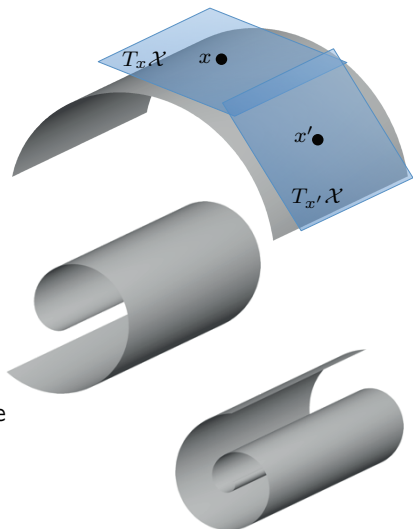
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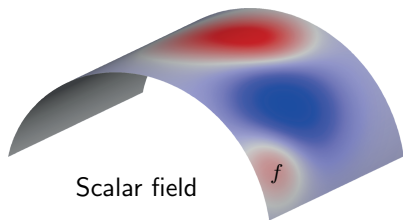
Isometry = metric-preserving shape deformation

Intrinsic = expressible solely in terms of Riemannian metric



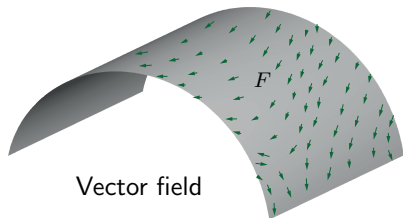
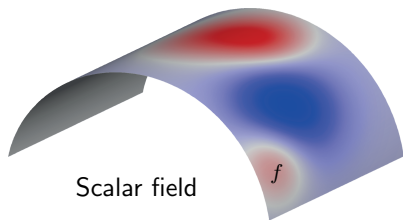
Calculus on manifolds: scalar and vector fields

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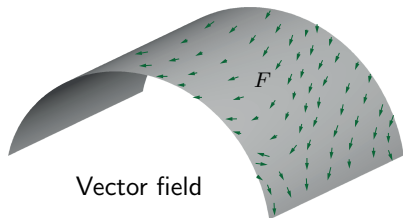
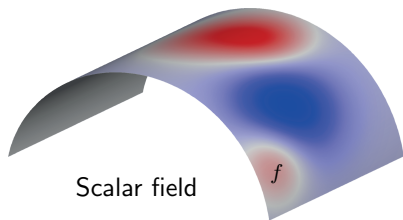
Calculus on manifolds: scalar and vector fields

- **Scalar field** $f : \mathcal{X} \rightarrow \mathbb{R}$
- **Vector field** $F : \mathcal{X} \rightarrow T\mathcal{X}$
- **Hilbert space** with inner products

$$\langle f, g \rangle_{\mathcal{F}(\mathcal{X})} = \int_{\mathcal{X}} f(x)g(x)dx$$

$$\langle F, G \rangle_{\mathcal{F}(T\mathcal{X})} = \int_{\mathcal{X}} \langle F(x), G(x) \rangle_x dx$$

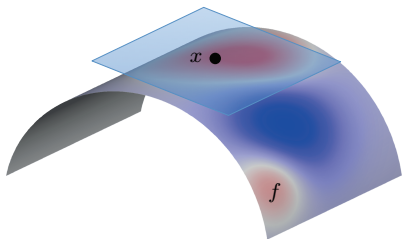
where dx = area element induced by the Riemannian metric



Calculus on manifolds: gradient

- **Differential** $df : T\mathcal{X} \rightarrow \mathbb{R}$ acting on vector fields

$$df(x) = \langle \nabla f(x), \cdot \rangle_x$$



Calculus on manifolds: gradient

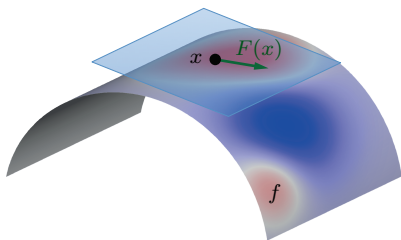
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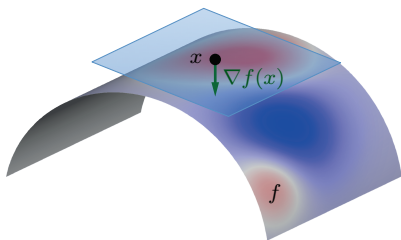
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“how much f changes at x in direction $F(x)$ ”

- **Intrinsic gradient operator**

$$\nabla f : \mathcal{F}(\mathcal{X}) \rightarrow \mathcal{F}(T\mathcal{X})$$

“direction of steepest change of f ”

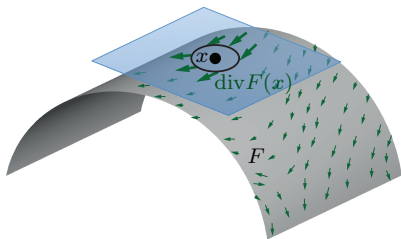


Calculus on manifolds: divergence

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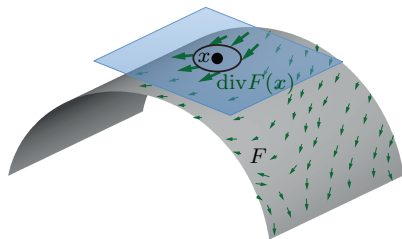
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Formal adjoint of the gradient

$$\langle F, \nabla f \rangle_{\mathcal{F}(T\mathcal{X})} = \langle \nabla^* F, f \rangle_{\mathcal{F}(\mathcal{X})} = \langle -\operatorname{div} F, f \rangle_{\mathcal{F}(\mathcal{X})}$$

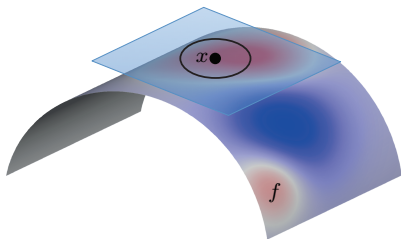


Calculus on manifolds: Laplacian

- Laplacian $\Delta : \mathcal{F}(\mathcal{X}) \rightarrow \mathcal{F}(\mathcal{X})$

$$\Delta f = -\operatorname{div}(\nabla f)$$

“difference between $f(x)$ and average value of f around x ”

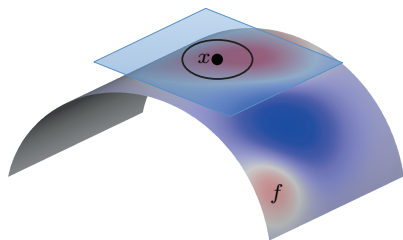


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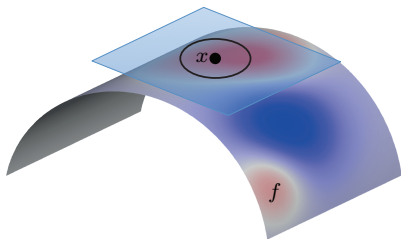
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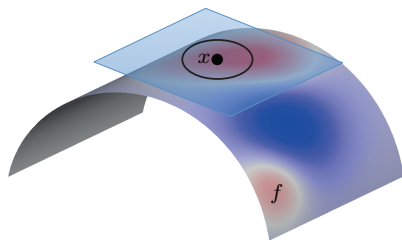
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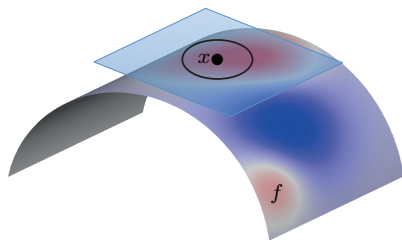
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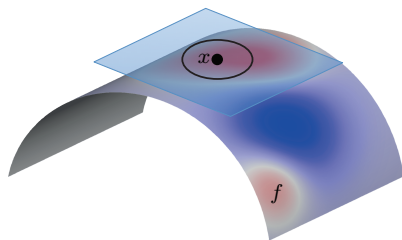
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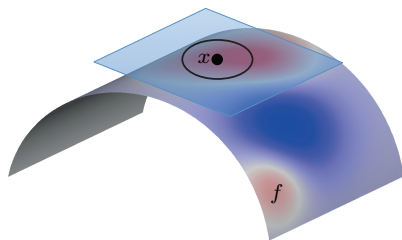
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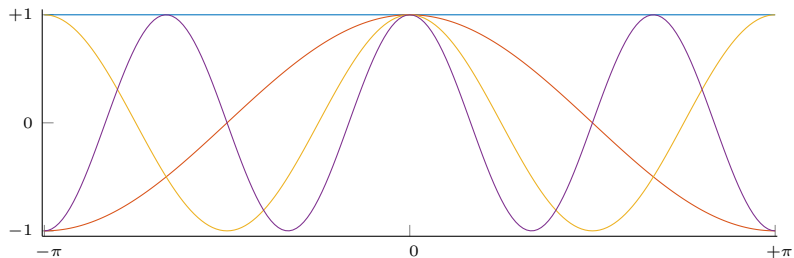
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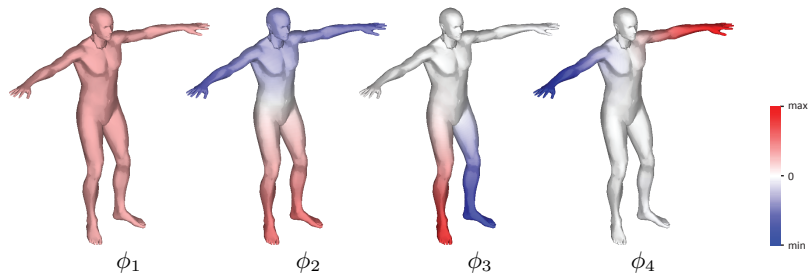
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- **Positive semidefinite** \Rightarrow non-negative eigenvalues

Laplacian eigenfunctions: Euclidean



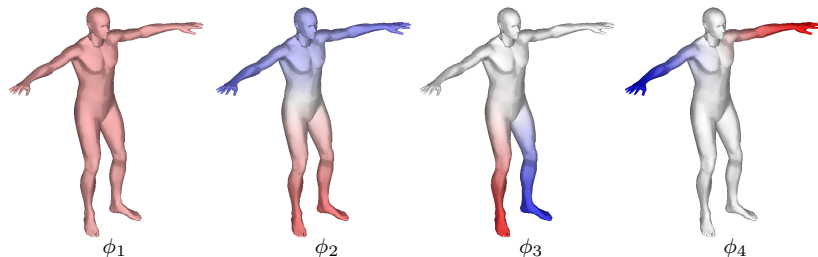
First eigenfunctions of 1D Euclidean Laplacian = standard Fourier basis

Laplacian eigenfunctions: non-Euclidean



First eigenfunctions of a manifold Laplacian = Fourier basis on manifolds

Laplacian eigenfunctions: non-Euclidean



For shapes with simple spectrum, Laplacian eigenfunctions are invariant (up to sign) to isometric deformations, $\psi_i = \pm T\phi_i$



Laplacian eigenbasis

- ϕ_1, ϕ_2, \dots is an **orthogonal basis** on $L^2(\mathcal{X})$, i.e. $\langle \phi_i, \phi_j \rangle_{L^2(\mathcal{X})} = \delta_{ij}$
- Smoothest orthogonal basis, due to minimization of the **Dirichlet energy**

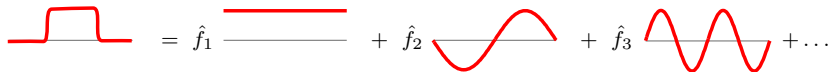
$$\min_{\phi_i} \|\nabla \phi_i\|^2 \quad \text{s.t.} \quad \|\phi_i\| = 1, \quad i = 1, 2, \dots$$
$$\phi_i \perp \text{span}\{\phi_1, \dots, \phi_{i-1}\}$$

- Optimal basis for smooth signals
- **Intrinsic**, hence invariant under inelastic deformations (isometries)
- Non-Euclidean analogy of the **Fourier transform**

Fourier analysis (Euclidean spaces)

A (square-integrable) function $f : [-\pi, \pi] \rightarrow \mathbb{R}$ can be written as **Fourier series**

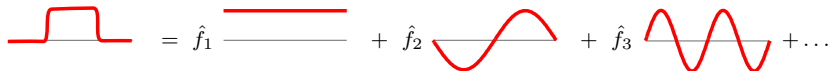
$$f(x) = \sum_{\omega} \frac{1}{2\pi} \int_{-\pi}^{\pi} f(\xi) e^{i\omega\xi} d\xi e^{-i\omega x}$$



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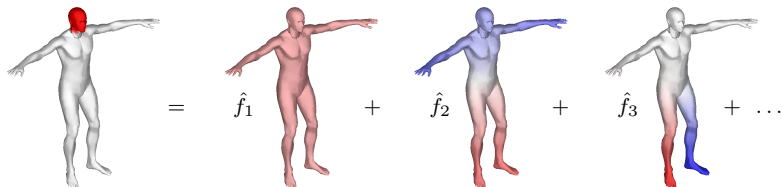
$$f(x) = \sum_{\omega} \underbrace{\frac{1}{2\pi} \int_{-\pi}^{\pi} f(\xi) e^{i\omega\xi} d\xi}_{\hat{f}_{\omega} = \langle f, e^{-i\omega x} \rangle_{L^2([-\pi, \pi])}} e^{-i\omega x}$$



Fourier analysis (non-Euclidean spaces)

A (square-integrable) function $f : \mathcal{X} \rightarrow \mathbb{R}$ can be written as **Fourier series**

$$f(x) = \sum_{k \geq 1} \underbrace{\int_{\mathcal{X}} f(\xi) \phi_k(\xi) d\xi}_{\hat{f}_k = \langle f, \phi_k \rangle_{L^2(\mathcal{X})}} \phi_k(x)$$



Heat diffusion on manifolds

Heat equation governs diffusion processes on manifolds:

$$\begin{cases} f_t(x, t) = -\Delta f(x, t) \\ f(x, 0) = f_0(x) \quad \text{initial condition} \end{cases}$$

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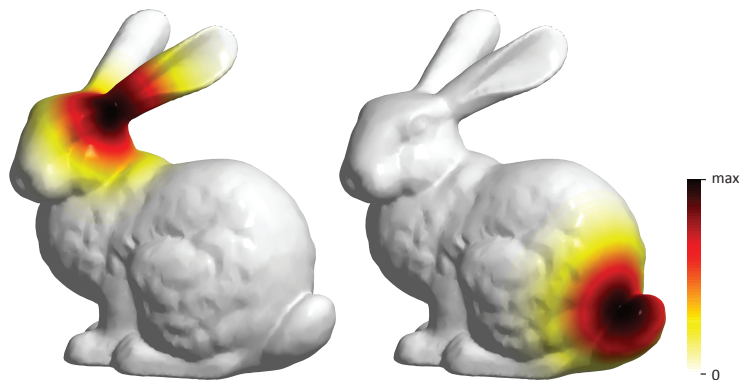
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Heat kernel



Heat kernel $h_t(x, \cdot)$ at different points on a manifold

$\lambda = \text{frequency}$

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What can be recovered from the Laplacian spectrum?

Heat trace expansion

$$\text{trace}(e^{-t\Delta}) = \sum_{k \geq 1} e^{-t\lambda_k} = \sum_{k \geq 1} a_k t^k$$

for a 2-manifold (without boundary) allows to recover the following properties from the Laplacian spectrum:

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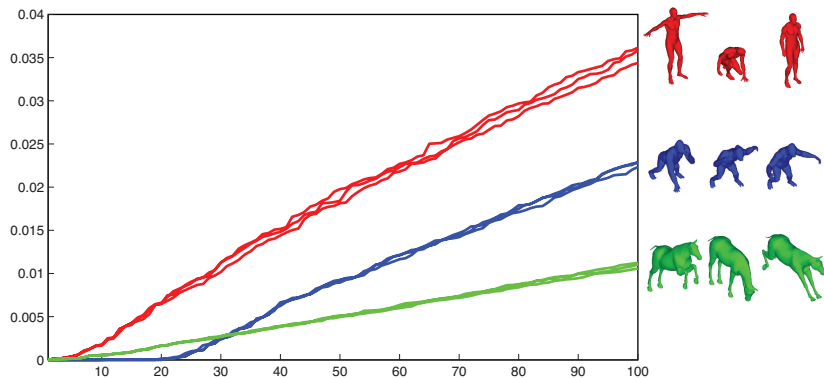
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- ...

Shape DNA



Laplacian spectrum used as deformation-invariant shape descriptor

isometric \implies isospectral

isometric $\overset{?}{\iff}$ **isospectral**

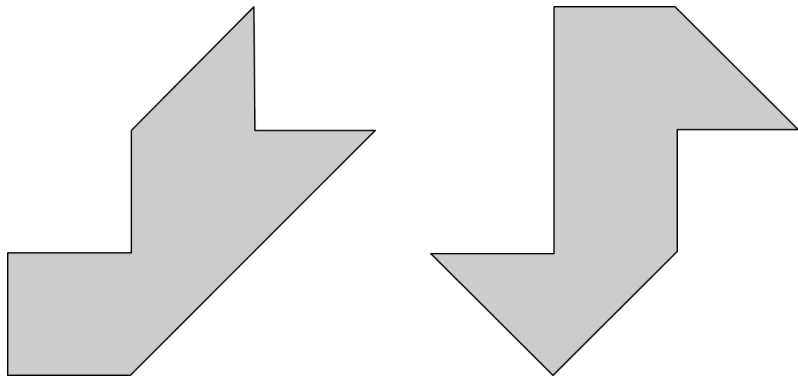
CAN ONE HEAR THE SHAPE OF A DRUM?

MARK KAC, The Rockefeller University, New York

To George Eugene Uhlenbeck on the occasion of his sixty-fifth birthday

“La Physique ne nous donne pas seulement l’occasion de résoudre des problèmes . . . , elle nous fait sentir la solution.” H. POINCARÉ.

Counter-examples

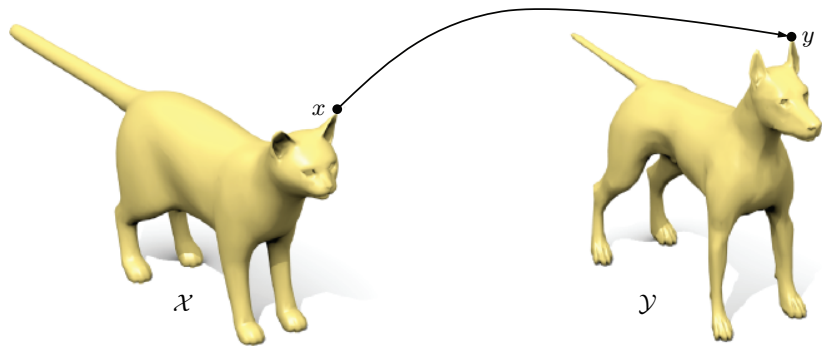


One cannot hear the shape of the drum!

$\lambda = \text{frequency}$

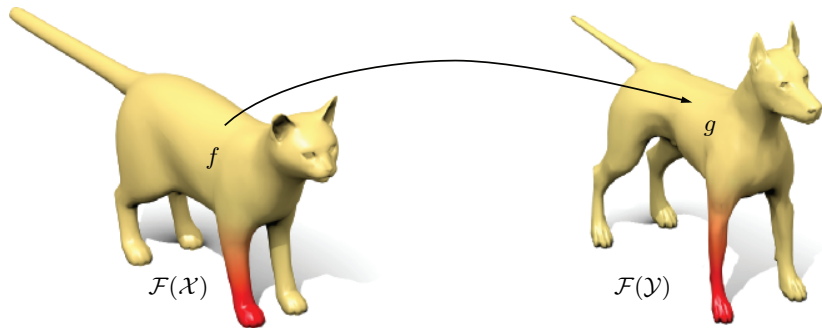
$\phi = \text{Fourier atoms}$

Pointwise correspondence



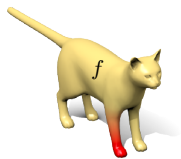
Point-wise map $\tau: \mathcal{X} \rightarrow \mathcal{Y}$

Functional correspondence

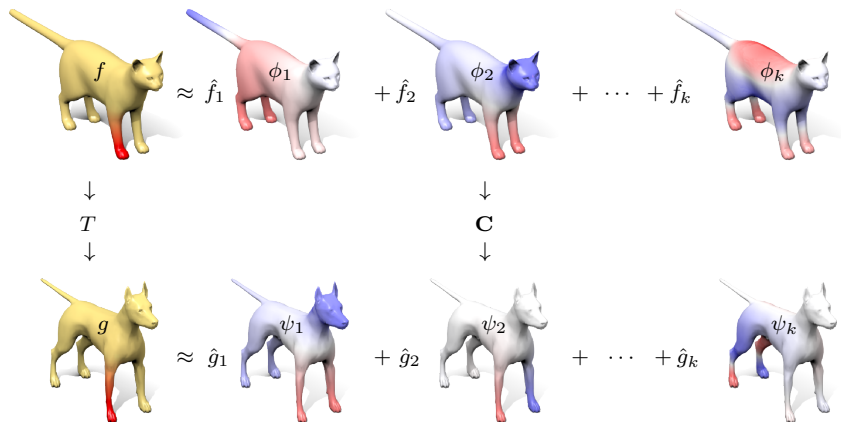


Functional map $T: \mathcal{F}(\mathcal{X}) \rightarrow \mathcal{F}(\mathcal{Y})$

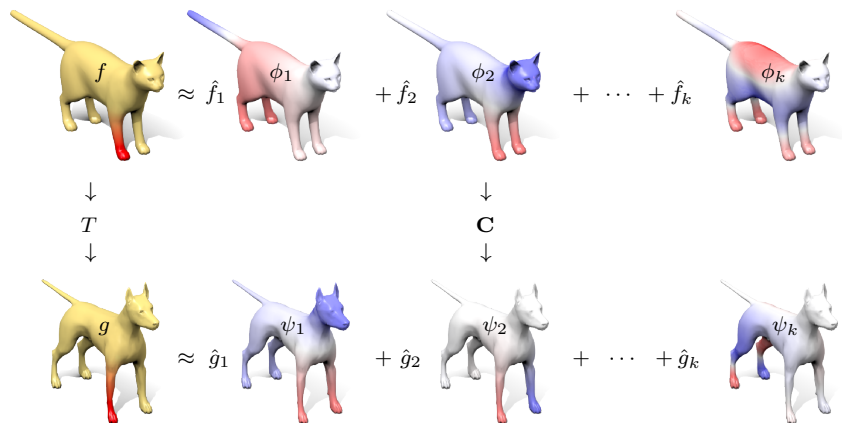
Functional correspondence in spectral domain



Functional correspondence in spectral domain



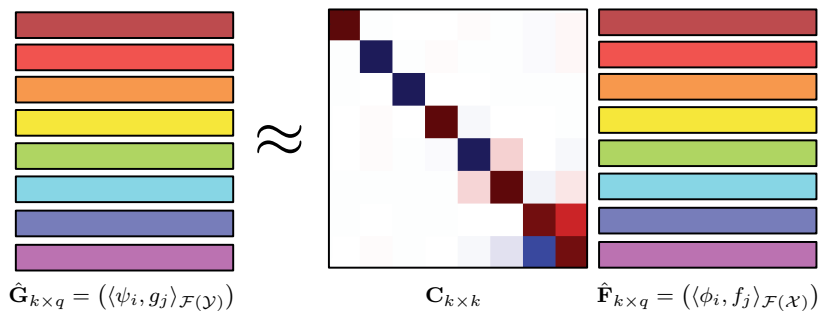
Functional correspondence in spectral domain



Functional correspondence boils down to a **linear equation** w.r.t. C

$$g = Tf \quad \iff \quad \hat{g} = C\hat{f}$$

Functional correspondence in spectral domain



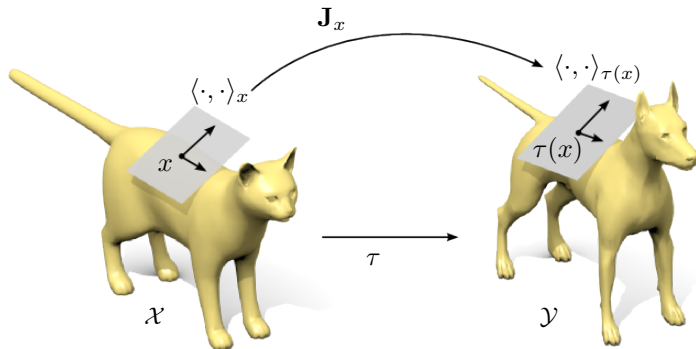
where $\hat{\mathbf{F}}$, $\hat{\mathbf{G}}$ are Fourier coefficients of corresponding 'probe' functions

$$g_i \approx T f_i \quad i = 1, \dots, q \geq k$$

Functional maps: Summary

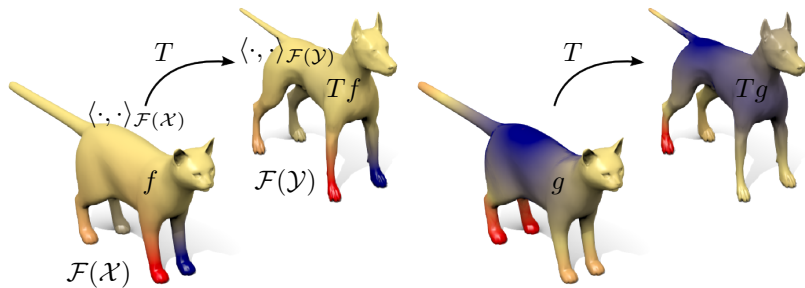
- Intrinsic by construction
- Operate with Fourier coefficients, forget about specific discretization (can apply to meshes, point clouds, etc.)
- Advantage: replace intractable combinatorial problems with tractable linear algebra problems
- Disadvantage: hard to guarantee properties of maps (e.g. bijectivity)
- Some properties can be guaranteed: e.g. **area preservation** = **orthogonality** of the functional map ($\mathbf{C}^\top \mathbf{C} = \mathbf{I}$)

Shape difference operators



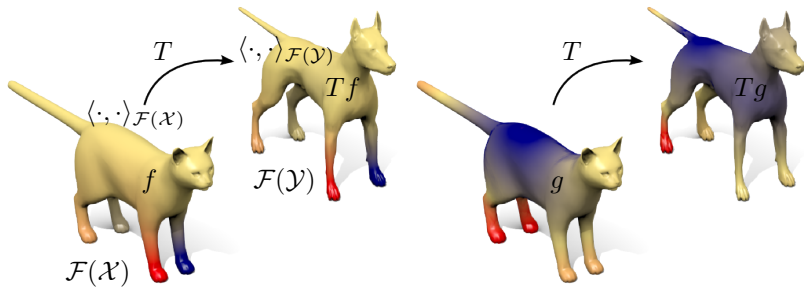
Distortions induced by a map = change of Riemannian metric
(inner product of **tangent vectors**)

Shape difference operators



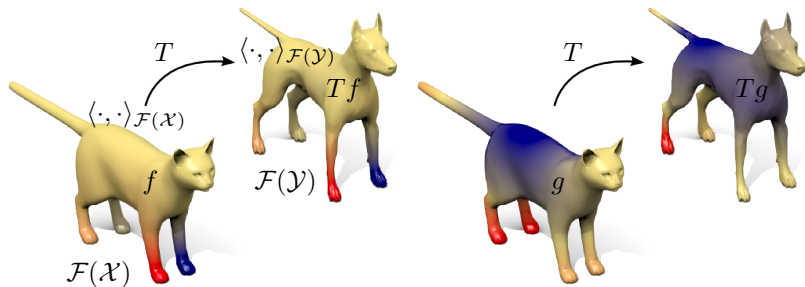
Distortions induced by a map = change of inner products of **functions**

Shape difference operators



$$\langle f, g \rangle_{\mathcal{F}(\mathcal{X})} \neq \langle Tf, Tg \rangle_{\mathcal{F}(\mathcal{Y})}$$

Shape difference operators

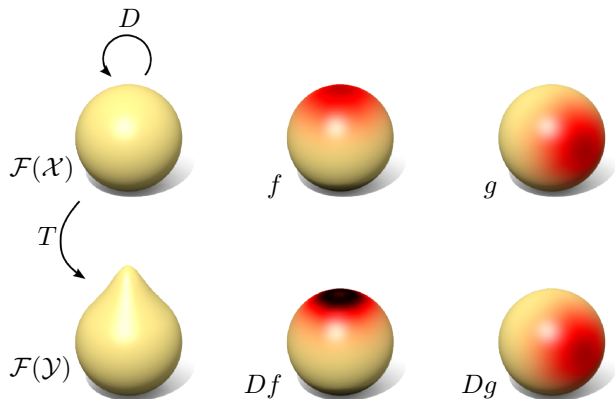


$$\langle f, g \rangle_{\mathcal{F}(X)} \neq \langle Tf, Tg \rangle_{\mathcal{F}(Y)}$$

Riesz theorem: there exists a unique self-adjoint linear operator $D: \mathcal{F}(X) \rightarrow \mathcal{F}(X)$ such that

$$\langle Tf, Tg \rangle_{\mathcal{F}(Y)} = \langle f, Dg \rangle_{\mathcal{F}(X)}$$

Shape difference operators



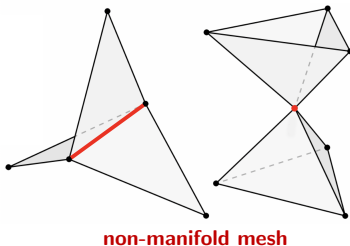
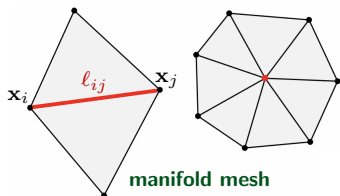
- Captures the **difference** in the geometry of the two shapes
- **Depends** on choice of inner product

Discretization

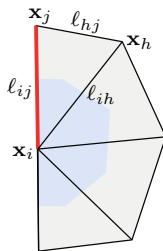
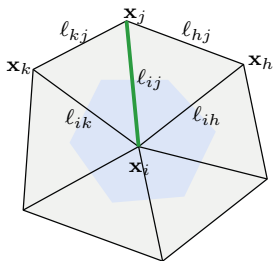
Manifold meshes

Surface discretized as an embedded **triangular mesh** $(\mathbf{X}, \mathcal{E}, \mathcal{F})$

- **Nodes:** points in 3D ($n \times 3$ matrix \mathbf{X} of coordinates)
- **Edges:** (i, j) , weighted by metric $\ell_{ij} = \|\mathbf{x}_i - \mathbf{x}_j\|$
- **Faces:** (i, j, k) s.t. $(i, j), (i, k), (k, j) \in \mathcal{E}$
- **Manifoldness assumption:**
 - Each edge is shared by **two triangles**
 - Boundary of triangles incident on each node forms a **single loop** of edges



Laplacian discretization



Cotangent Laplacian $\Delta = \mathbf{A}^{-1}\mathbf{W}$ expressed in terms of discrete metric $l_{ij} = \|\mathbf{x}_i - \mathbf{x}_j\|$ where

$$w_{ij} = \begin{cases} \frac{-l_{ij}^2 + l_{jk}^2 + l_{ki}^2}{8A_{ijk}} + \frac{-l_{ij}^2 + l_{jh}^2 + l_{hi}^2}{8A_{ijh}} & \text{if } e_{ij} \in \mathcal{E}_i \\ \frac{-l_{ij}^2 + l_{jh}^2 + l_{hi}^2}{8A_{ijh}} & \text{if } e_{ij} \in \mathcal{E}_b \\ -\sum_{k \neq i} w_{ik} & \text{if } i = j \end{cases} \quad \mathbf{A} = \begin{pmatrix} a_1 & & \\ & \ddots & \\ & & a_n \end{pmatrix}$$

where A_{ijk} is area of triangle ijk and $a_i = \frac{1}{3} \sum_{ijk:ij,ik \in \mathcal{E}} A_{ijk}$

Shape difference discretization

- **area-based**

$$\langle f, g \rangle_{L^2(\mathcal{X})} = \int_{\mathcal{X}} f(x)g(x)dx$$

$$\mathbf{D} = \mathbf{V}_{\mathcal{X}, \mathcal{Y}} = \mathbf{A}_{\mathcal{X}}^{-1} \mathbf{T}^{\top} \mathbf{A}_{\mathcal{Y}} \mathbf{T}$$

Shape difference discretization

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- **conformal-based**

$$\langle f, g \rangle_{H^1(\mathcal{X})} = \int_{\mathcal{X}} \langle \nabla f(x), \nabla g(x) \rangle_x dx$$

$$\mathbf{D} = \mathbf{R}_{\mathcal{X},\mathcal{Y}} = \mathbf{W}_{\mathcal{X}}^{\dagger} \mathbf{T}^{\top} \mathbf{W}_{\mathcal{Y}} \mathbf{T}$$

Shape difference discretization

- **area**-based

$$\langle f, g \rangle_{L^2(\mathcal{X})} = \int_{\mathcal{X}} f(x)g(x)dx$$

$$\mathbf{D} = \mathbf{V}_{\mathcal{X},\mathcal{Y}} = \mathbf{A}_{\mathcal{X}}^{-1} \mathbf{T}^{\top} \mathbf{A}_{\mathcal{Y}} \mathbf{T}$$

- $\mathbf{V} = \mathbf{I}$: **area-preserving** maps

- **conformal**-based

$$\langle f, g \rangle_{H^1(\mathcal{X})} = \int_{\mathcal{X}} \langle \nabla f(x), \nabla g(x) \rangle_x dx$$

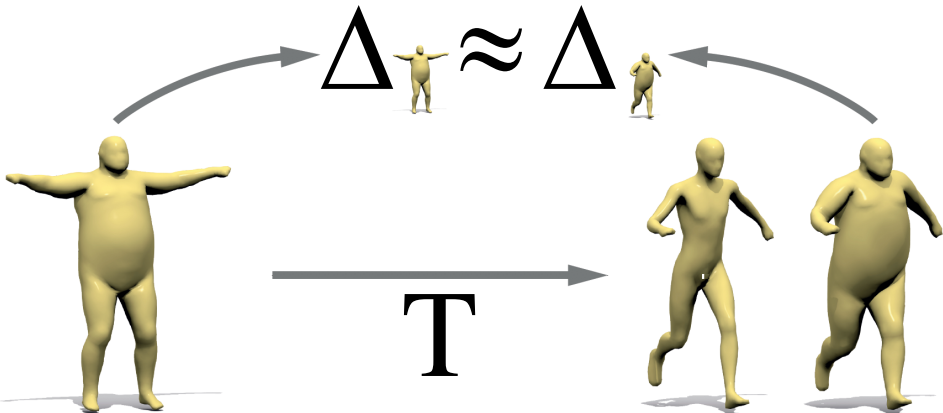
$$\mathbf{D} = \mathbf{R}_{\mathcal{X},\mathcal{Y}} = \mathbf{W}_{\mathcal{X}}^{\dagger} \mathbf{T}^{\top} \mathbf{W}_{\mathcal{Y}} \mathbf{T}$$

- $\mathbf{R} = \mathbf{I}$: **angle-preserving** (conformal) map

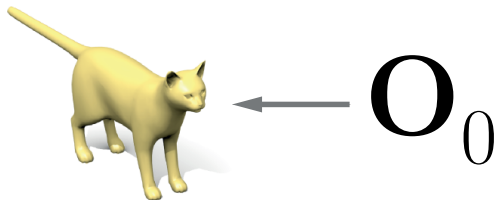
- $\mathbf{V} = \mathbf{R} = \mathbf{I}$: **isometric** map

Shape-from-Operator: Recovering Shapes from Intrinsic Operators

Davide Boscaini, Davide Eynard, Drosos Kourounis, and Michael M. Bronstein
Università della Svizzera Italiana (USI), Lugano, Switzerland



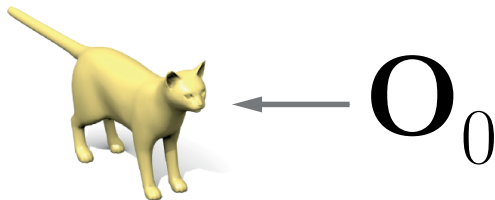
Shape-from-Operator problems



Generic **Shape-from-Operator (SfO)** problem: given some intrinsic operator \mathbf{O}_0 , find an embedding \mathbf{X} by minimizing some cost function

$$\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} E(\mathbf{O}(\ell(\mathbf{X})), \mathbf{O}_0)$$

Shape-from-Operator problems



Generic **Shape-from-Operator (SfO)** problem: given some intrinsic operator O_0 , find an embedding X by minimizing some cost function

$$\min_{X \in \mathbb{R}^{n \times 3}} E(O(\ell(X)), O_0)$$

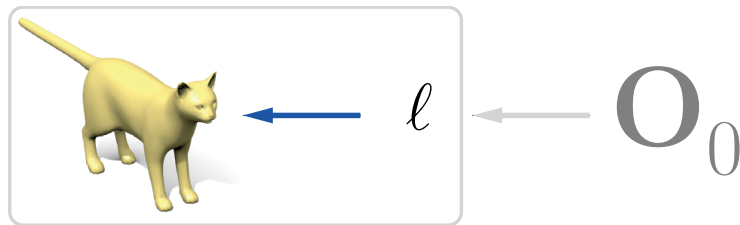
O depends on X indirectly through the discrete metric $\ell(X)$, very hard for optimization!

Shape-from-Operator problems



- **Metric-from-Operator (MfO):** $\min_{\ell} E(\mathbf{O}(\ell), \mathbf{O}_0)$ s.t. Δ inequality
- **Shape-from-Metric (SfM):** $\min_{\mathbf{x} \in \mathbb{R}^{n \times 3}} \sum_{(i,j) \in \mathcal{E}} (\|\mathbf{x}_i - \mathbf{x}_j\| - \ell_{ij})^2,$

Shape-from-Operator problems



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Shape-from-metric

Special setting of **MDS**: given a metric ℓ , find its Euclidean realization by minimizing the stress

$$\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} \sum_{i,j=1}^n v_{ij} (\|\mathbf{x}_i - \mathbf{x}_j\| - \ell_{ij})^2,$$

where

$$v_{ij} = \begin{cases} 1 & \text{if } (i, j) \in \mathcal{E}, \\ 0 & \text{otherwise} \end{cases}$$

Shape-from-metric

Special setting of **MDS**: given a metric ℓ , find its Euclidean realization by minimizing the stress

$$\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} \sum_{i,j=1}^n v_{ij} (\|\mathbf{x}_i - \mathbf{x}_j\| - \ell_{ij})^2,$$

where

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SMACOF algorithm: fixed point iteration of the form

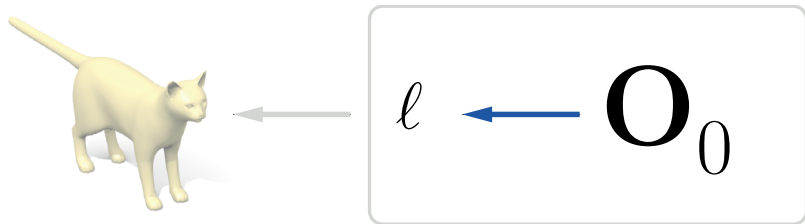
$$\mathbf{X} \leftarrow \mathbf{Z}^\dagger \mathbf{B}(\mathbf{X}) \mathbf{X}$$

where

$$\mathbf{Z} = \begin{cases} -v_{ij} & \text{if } i \neq j, \\ \sum_{i \neq j} v_{ij} & \text{if } i = j \end{cases} \quad \mathbf{B}(\mathbf{X}) = \begin{cases} -\frac{v_{ij} \ell_{ij}}{\|\mathbf{x}_i - \mathbf{x}_j\|} & \text{if } i \neq j \text{ and } \mathbf{x}_i \neq \mathbf{x}_j, \\ 0 & \text{if } i \neq j \text{ and } \mathbf{x}_i = \mathbf{x}_j, \\ \sum_{i \neq j} b_{ij} & \text{if } i = j \end{cases}$$

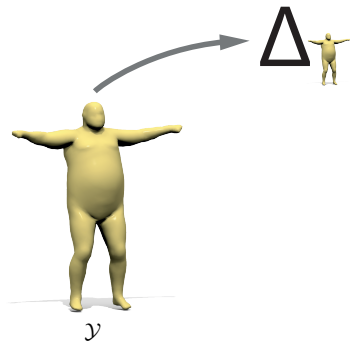
Leeuw et al., 1977

From operators to shapes



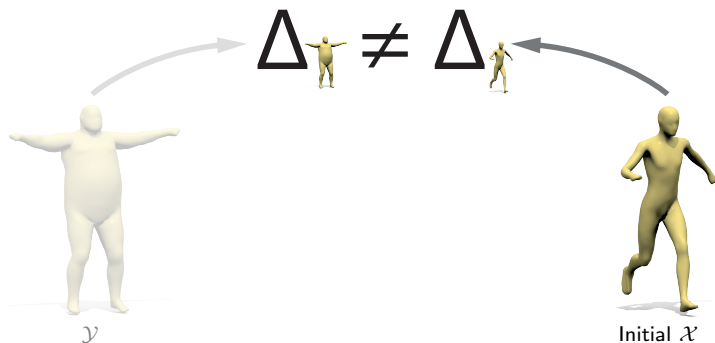
- **Metric-from-Operator (MfO):** $\min_{\ell} E(\mathbf{O}(\ell), \mathbf{O}_0)$ s.t. Δ inequality
- **Shape-from-Metric (SfM):** $\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} \sum_{(i,j) \in \mathcal{E}} (\|\mathbf{x}_i - \mathbf{x}_j\| - \ell_{ij})^2,$

Shape-from-Laplacian



Given a reference Laplacian operator Δ_γ

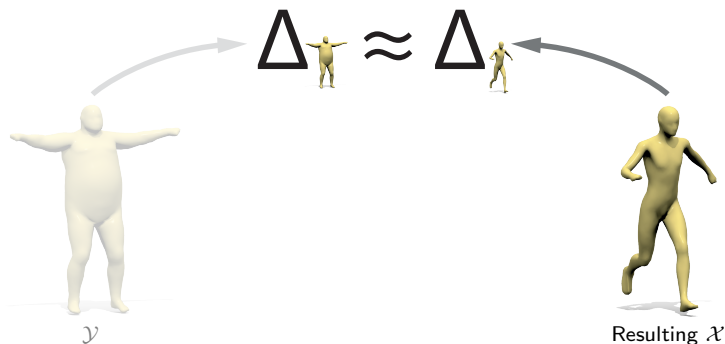
Shape-from-Laplacian



Given a reference Laplacian operator $\Delta_{\mathcal{Y}}$, and a corresponding initial shape \mathcal{X} , deform \mathcal{X} by minimizing

$$\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} \|\mathbf{A}^{-1}(\ell(\mathbf{X}))\mathbf{W}(\ell(\mathbf{X})) - \Delta_{\mathcal{Y}}\|$$

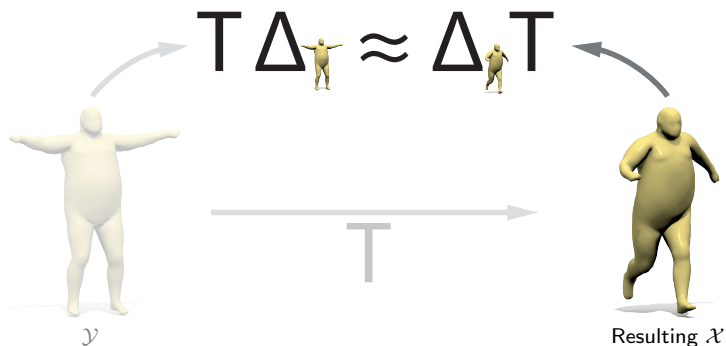
Shape-from-Laplacian



Given a reference Laplacian operator Δ_{γ} , and a corresponding initial shape \mathcal{X} , deform \mathcal{X} by minimizing

$$\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} \|\mathbf{A}^{-1}(\ell(\mathbf{X}))\mathbf{W}(\ell(\mathbf{X})) - \Delta_{\gamma}\|$$

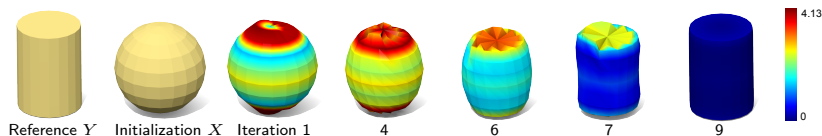
Shape-from-Laplacian



Given a reference Laplacian operator $\Delta_{\mathcal{Y}}$, and a initial shape \mathcal{X} related by functional correspondence \mathbf{T} , deform \mathcal{X} by minimizing

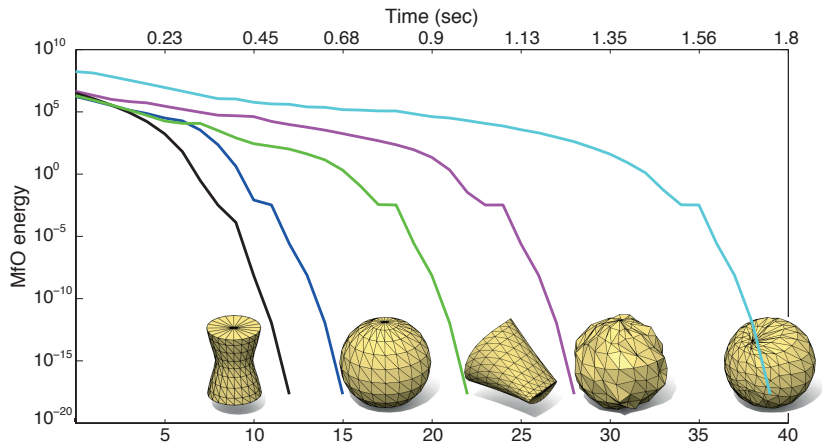
$$\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} \|\mathbf{T} \mathbf{A}^{-1}(\ell(\mathbf{X})) \mathbf{W}(\ell(\mathbf{X})) - \Delta_{\mathcal{Y}} \mathbf{T}\|$$

Shape-from-Laplacian convergence



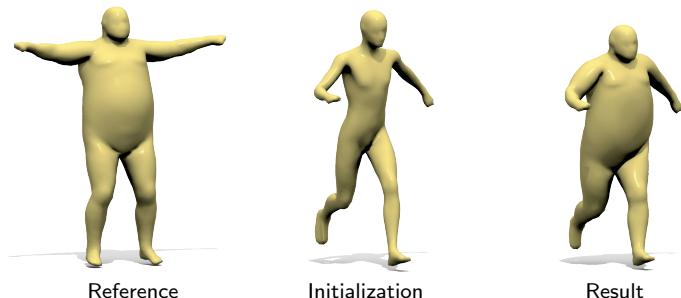
Convergence of our method in the shape-from-Laplacian optimization problem.
Colors show vertex-wise MfO energy contribution

Shape-from-Laplacian convergence



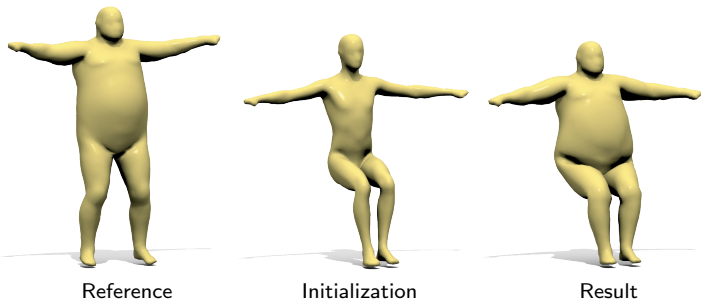
Convergence of our method in the shape-from-Laplacian optimization problem using different initializations.

Style transfer by shape-from-Laplacian



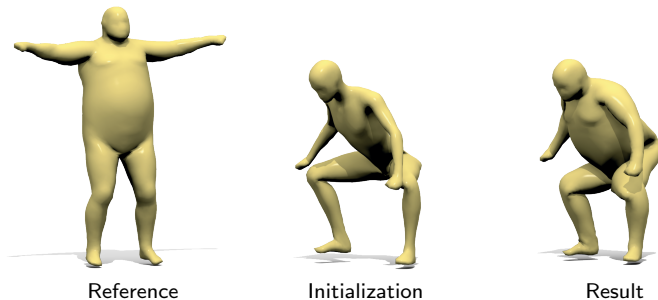
“Modify \mathcal{X} such that $\Delta_{\mathcal{X}}$ becomes as similar as possible to reference Laplacian $\Delta_{\mathcal{Y}}$ ”

Style transfer by shape-from-Laplacian



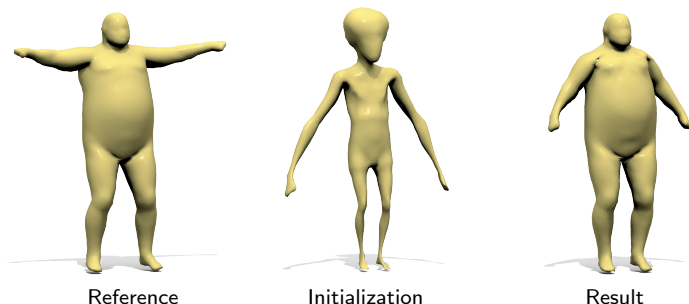
“Modify \mathcal{X} such that $\Delta_{\mathcal{X}}$ becomes as similar as possible to reference Laplacian $\Delta_{\mathcal{Y}}$ ”

Style transfer by shape-from-Laplacian



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Style transfer by shape-from-Laplacian



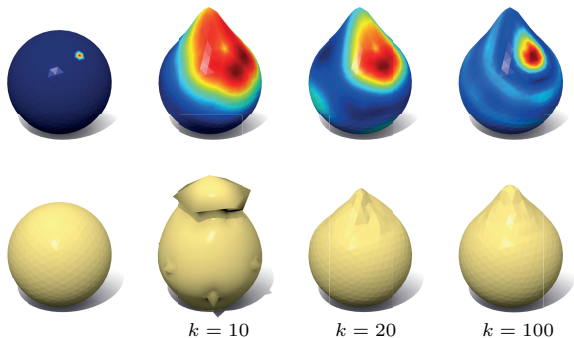
“Modify \mathcal{X} such that $\Delta_{\mathcal{X}}$ becomes as similar as possible to reference Laplacian $\Delta_{\mathcal{Y}}$ ”

Sensitivity to map quality

Functional map approximated as a matrix $\mathbf{T} \approx \mathbf{\Psi}_k \mathbf{C}^\top \mathbf{\Phi}_k^\top$ of rank k using the first functions in Fourier expansion (larger k = better map)

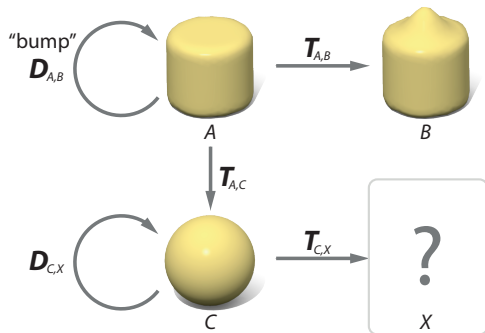
Sensitivity to map quality

Functional map approximated as a matrix $\mathbf{T} \approx \Psi_k \mathbf{C}^\top \Phi_k^\top$ of rank k using the first functions in Fourier expansion (larger k = better map)



Shape-from-Laplacian result for different quality of the map \mathbf{T}
(initial shape: sphere, reference Laplacian: bumped sphere)

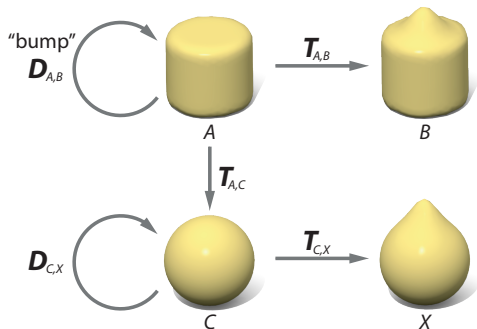
Shape-from-difference operator



Deform initial shape \mathcal{X} to make it different from \mathcal{C} same way as \mathcal{B} is different from \mathcal{A}

$$\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} \|\mathbf{D}_{\mathcal{C},\mathcal{X}}(\ell(\mathbf{X}))\mathbf{T}_{\mathcal{A},\mathcal{C}} - \mathbf{T}_{\mathcal{A},\mathcal{C}}\mathbf{D}_{\mathcal{A},\mathcal{B}}\|$$

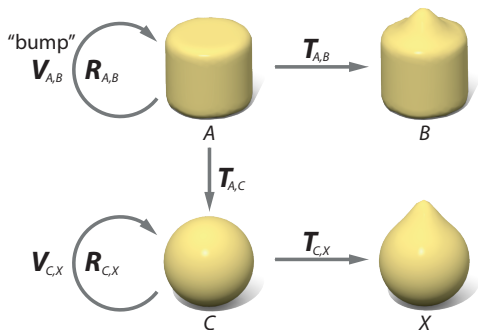
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Shape-from-difference operator

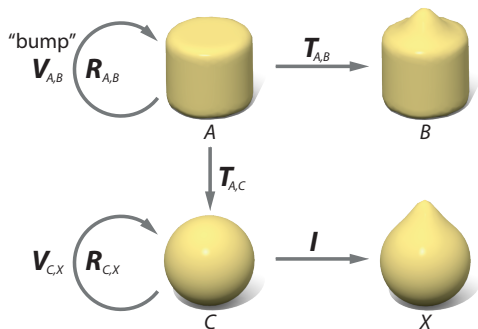


Deform initial shape \mathcal{X} to make it different from \mathcal{C} same way as \mathcal{B} is different from \mathcal{A}

$$\min_{\mathbf{X} \in \mathbb{R}^{n \times 3}} \mu \|\mathbf{A}_C^{-1} \mathbf{T}_{C,\mathcal{X}}^\top \mathbf{A}(\ell(\mathbf{X})) \mathbf{T}_{C,\mathcal{X}} \mathbf{T}_{A,C} - \mathbf{T}_{A,C} \mathbf{V}_{A,B}\| +$$

$$(1 - \mu) \|\mathbf{W}_C^\dagger \mathbf{T}_{C,\mathcal{X}}^\top \mathbf{W}(\ell(\mathbf{X})) \mathbf{T}_{C,\mathcal{X}} \mathbf{T}_{A,C} - \mathbf{T}_{A,C} \mathbf{R}_{A,B}\|$$

Shape-from-difference operator

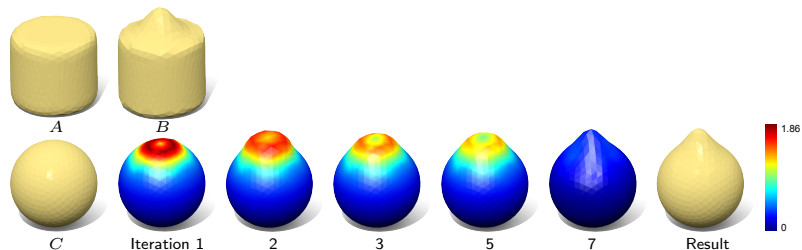


Deform initial shape \mathcal{X} to make it different from \mathcal{C} same way as \mathcal{B} is different from \mathcal{A}

$$\min_{\mathbf{X}} \quad \mu \|\mathbf{A}_C^{-1} \mathbf{A}(\ell(\mathbf{X})) \mathbf{T}_{A,C} - \mathbf{T}_{A,C} \mathbf{V}_{A,B}\| + \\ (1 - \mu) \|\mathbf{W}_C^\dagger \mathbf{W}(\ell(\mathbf{X})) \mathbf{T}_{A,C} - \mathbf{T}_{A,C} \mathbf{R}_{A,B}\|$$

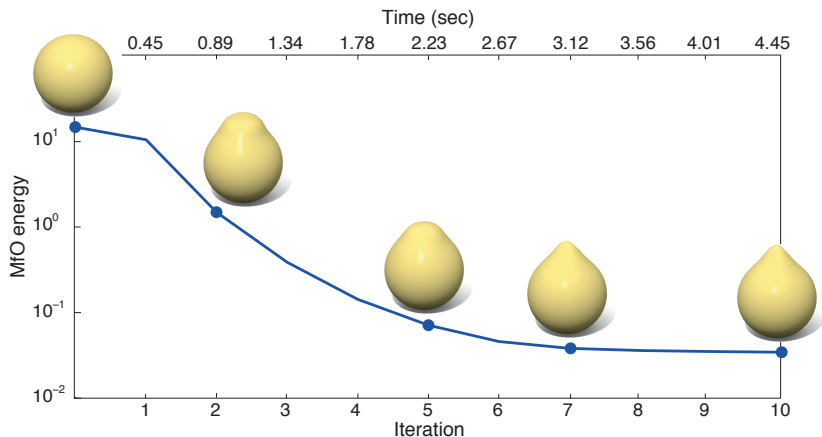
(initializing $\mathcal{X} = \mathcal{C}$ we have $\mathbf{T}_{\mathcal{C},\mathcal{X}} = \mathbf{I}$)

Shape-from-difference convergence



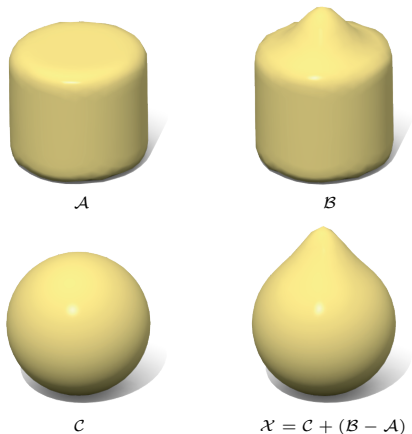
Convergence of our method in the shape-from-difference optimization problem
Colors show vertex-wise MfO energy contribution

Shape-from-difference convergence



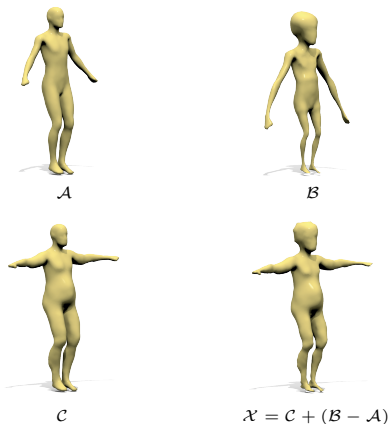
Convergence of our method in the shape-from-difference optimization problem.

Analogy synthesis by shape-from-difference



“Find \mathcal{X} such that the difference operator between \mathcal{C} , \mathcal{X} is as similar as possible to the given difference operator between \mathcal{A} , \mathcal{B} ”

Analogy synthesis by shape-from-difference



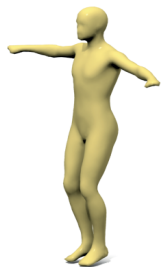
“Find X such that the difference operator between C , X is as similar as possible to the given difference operator between A , B ”

Analogy synthesis by shape-from-difference



“Find \mathcal{X} such that the difference operator between \mathcal{C} , \mathcal{X} is as similar as possible to the given difference operator between \mathcal{A} , \mathcal{B} ”

Shape exaggeration



\mathcal{A}



\mathcal{B}



' $\mathcal{B} + (\mathcal{B} - \mathcal{A})$ '



' $\mathcal{B} + 2(\mathcal{B} - \mathcal{A})$ '

Shape exaggeration obtained by applying the difference operator between \mathcal{A} , \mathcal{B} to \mathcal{B} several times

Summary

- Laplacian eigenvectors form an orthogonal basis for functions on a manifold
- Lifting to space of functions on manifolds: replace combinatorial problems with linear algebra problems
- Operate with Fourier coefficients, forget about specific discretization
- Represent shapes as operators (contains full information up to some class of transformations, e.g. isometries or conformal maps)
- Recover shapes from operators
- Interesting links to Geometric Deep Learning