# **Shape-from-operators**: recovering shapes from intrinsic differential operators

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**TUM**, Germany 26 November 2014



Slide idea: Crane

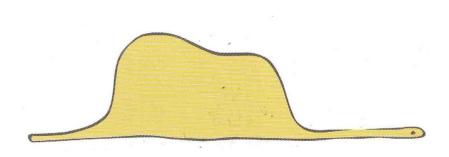


Slide idea: Crane

#### The challenges of non-rigidity

"Once upon a time, a child imagined a fierce boa constrictor that swallowed an elephant, giving rise to a most peculiar shape.

[... However] nobody saw in his drawing more than a hat.



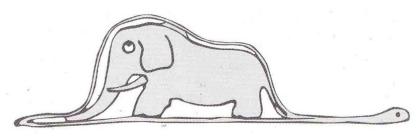
#### The challenges of non-rigidity

"Once upon a time, a child imagined a fierce boa constrictor that swallowed an elephant, giving rise to a most peculiar shape.

[... However] nobody saw in his drawing more than a hat.

Then the child proceeded to draw an explanatory drawing showing the elephant inside the snake's expansible stomach..."

de Saint-Exupéry, The Little Prince



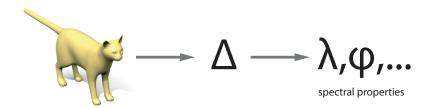


# Can one hear the shape of a drum?

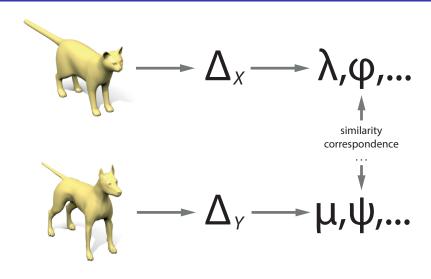


Kac, 66 6/33

## Spectral shape analysis



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#### Shape-from-Operator inverse problem



**Shape-from-Operator:** find a shape whose Laplacian (or another intrinsic operator) satisfies some properties

#### Shape-from-Operator inverse problem

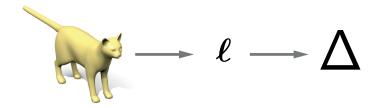


**Shape-from-Operator:** find a shape whose Laplacian (or another intrinsic operator) satisfies some properties

- Embedding from angles<sup>1</sup>, curvature<sup>2</sup>, discrete fundamental forms<sup>3</sup>,...
- Edge length from Laplacian<sup>4,5</sup>
- Closest commuting operators<sup>6</sup>
- Laplacian colormaps<sup>7</sup>

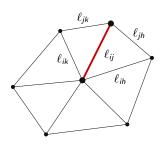
 $<sup>^{1}</sup>$ Sheffer, de Sturler 2001;  $^{2}$ Ben-Chen et al. 2008;  $^{3}$ Wang et al. 2012;  $^{4}$ Zeng et al. 2012;  $^{5}$ de Goes et al. 2014;  $^{6}$ B, Glashoff, Loring 2013;  $^{7}$ Eynard, Kovnatsky, B 2014

#### From shapes to operators

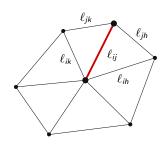


- Laplacian is **intrinsic** = expressible in terms of discrete metric (edge lengths)
- Embedding induces a discrete metric (and thus also a Laplacian)
- Not unique: many embeddings give rise to the same metric (isometries)

• Triangular mesh (X, E, F)



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- Discrete metric  $\ell = (\ell_{ij}, \ (i,j) \in E)$



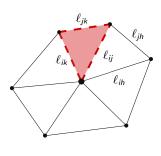
- Triangular mesh (X, E, F)
- Discrete metric  $\ell = (\ell_{ij}, \ (i,j) \in E)$
- Intrinsic edge weights

$$w_{ij} = \frac{-\ell_{ij}^2 + \ell_{jk}^2 + \ell_{ik}^2}{8A_{ijk}} + \frac{-\ell_{ij}^2 + \ell_{jh}^2 + \ell_{ih}^2}{8A_{ijh}}$$

where by Heron's formula

$$A_{ijk} = \sqrt{s(s - \ell_{ij})(s - \ell_{jk})(s - \ell_{ik})},$$

and s is the semi-perimeter



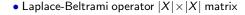
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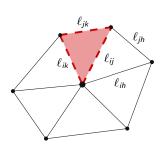
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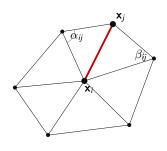
$$L = D - W$$

where 
$$\mathbf{D} = \operatorname{diag}(\sum_{i \neq j} w_{ij})$$



- Triangular mesh (X, E, F)
- $\bullet$  Embedding of the mesh in  $\mathbb{R}^3$  specifies the coordinates  $\boldsymbol{X}=\{\boldsymbol{x}_1,\dots,\boldsymbol{x}_n\}$
- Embedding X induces the metric

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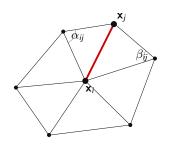


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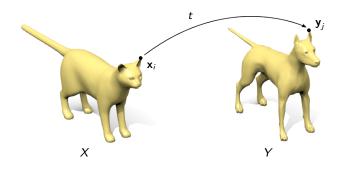
$$\boldsymbol{\ell} = (\|\mathbf{x}_i - \mathbf{x}_j\|, (i, j) \in \boldsymbol{E})$$

Cotangent weights

$$w_{ij} = egin{cases} rac{\cot(lpha_{ij}) + \cot(eta_{ij})}{2}, & (i,j) \in E \ 0, & ext{otherwise} \end{cases}$$

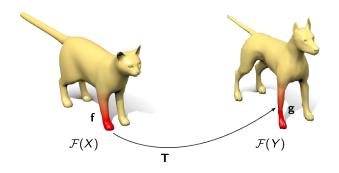


## Functional maps

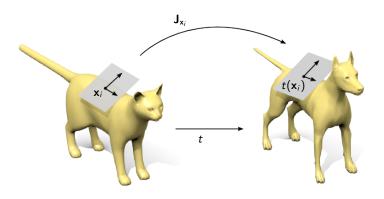


Point-wise maps  $t: X \to Y$ 

## Functional maps

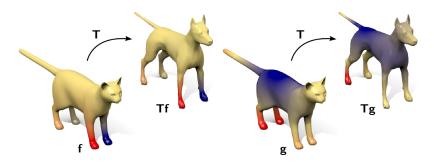


Functional maps  $T \colon \mathcal{F}(X) \to \mathcal{F}(Y)$ 



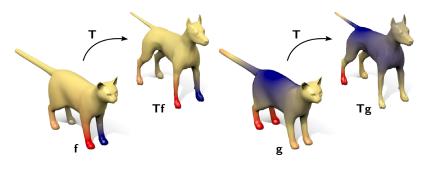
Distortions induced by a map = change of inner products of vectors

Rustamov et al., 2013



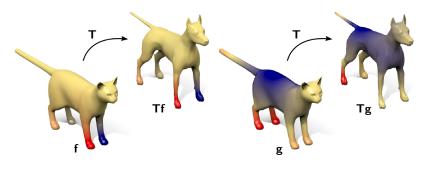
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 $\langle f,g\rangle_{\mathcal{F}(X)} \neq \langle Tf,Tg\rangle_{\mathcal{F}(Y)}$ 

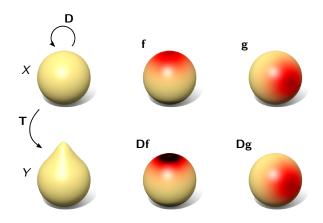
Rustamov et al., 2013 14/



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

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

$$\langle \textbf{T} \textbf{f}, \textbf{T} \textbf{g} \rangle_{\mathcal{F}(Y)} = \langle \textbf{f}, \textbf{D} \textbf{g} \rangle_{\mathcal{F}(X)}$$



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

Rustamov et al., 2013

## Shape difference discretization

• area-based,

$$\langle f, g \rangle_{L^2(X)} = \int_X f(x)g(x)d\mu(x)$$
  
 $\mathbf{D} = \mathbf{V}_{X,Y} = \mathbf{A}_X^{-1}\mathbf{T}^{\top}\mathbf{A}_Y\mathbf{T}$ 

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conformal-based,

$$\langle f, g \rangle_{H^1(X)} = \int_X \nabla f(x) \nabla g(x) d\mu(x)$$
  
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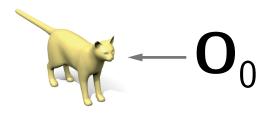
 $\bullet$  if  $\boldsymbol{V}=\boldsymbol{I},$  the map preserves the areas

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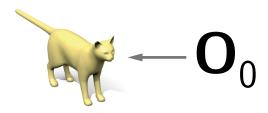
$$\mathbf{D} = \mathbf{R}_{X,Y} = \mathbf{W}_X^{\dagger} \mathbf{T}^{\top} \mathbf{W}_Y \mathbf{T}$$

- ullet if  ${f R}={f I}$ , the map preserves the angles
- ullet if  $oldsymbol{V} = oldsymbol{R} = oldsymbol{I}$ , the map is an isometry



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

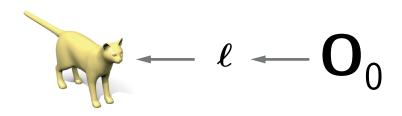
$$\min_{\boldsymbol{X}} \mathcal{E}\big(\boldsymbol{O}(\boldsymbol{\ell}(\boldsymbol{X})), \boldsymbol{O}_0\big)$$



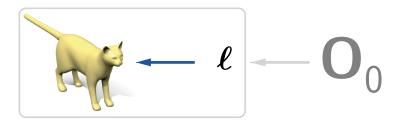
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Note: O depends on X indirectly through the discrete metric  $\ell(X)$ , very hard for optimization!



- Metric-from-Operator (MfO):  $\min_{\ell} \mathcal{E} \left( \mathbf{O}(\ell), \mathbf{O}_0 \right)$  s.t. triangle inequality
- Shape-from-Metric (SfM):  $\min_{\mathbf{X}} \sum_{ij \in E}^{n} (\|\mathbf{x}_i \mathbf{x}_j\| \ell_{ij})^2$ ,



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#### Shape-from-metric

Special setting of MDS: given a metric  $\ell$ , find its Euclidean realization by minimizing the stress

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

where

$$v_{ij} = egin{cases} 1 & ext{if } ij \in E, \ 0 & ext{otherwise} \end{cases}$$

Leeuw et al., 1977

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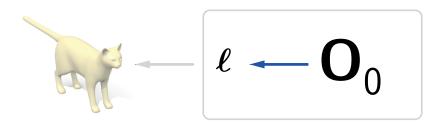
SMACOF algorithm: fixed point iteration of the form

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

where

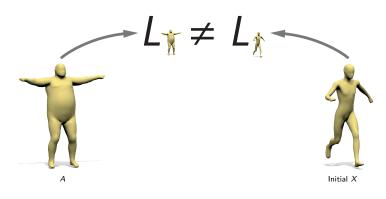
$$\mathbf{Z} = \begin{cases} -\mathbf{v}_{ij} & \text{if } i \neq j, \\ \sum_{i \neq j} \mathbf{v}_{ij} & \text{if } i = j \end{cases} \qquad \mathbf{B}(\mathbf{X}) = \begin{cases} -\frac{\mathbf{v}_{ij}\ell_{ij}}{\|\mathbf{x}_i - \mathbf{x}_j\|} & \text{if } i \neq j \text{ and } \mathbf{x}_i \neq \mathbf{x}_j,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

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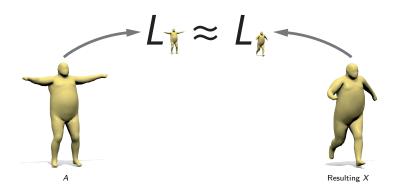
#### Shape-from-Laplacian



Given a reference Laplacian operator  $\mathbf{L}_A$ , and a corresponding initial shape X, deform X by minimizing

$$\min_{\boldsymbol{X}}\|\boldsymbol{L}(\boldsymbol{\ell}(\boldsymbol{X}))-\boldsymbol{L}_{\!A}\|$$

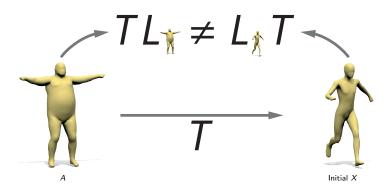
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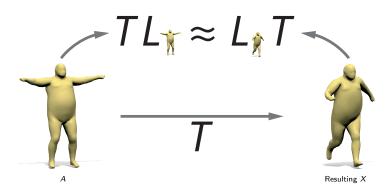
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Given a reference Laplacian operator  $\mathbf{L}_A$ , and a initial shape X related by functional correspondence T, deform X by minimizing

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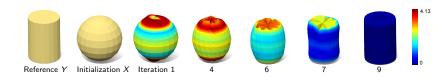
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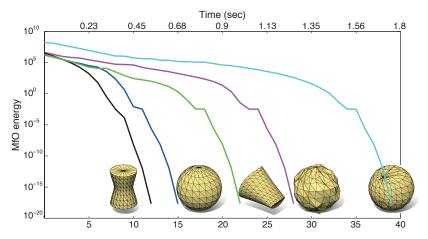
#### Shape-from-Laplacian convergence



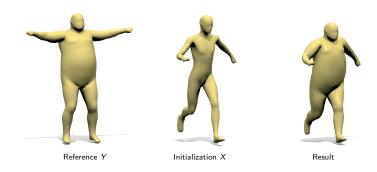
Convergence of our method in the shape-from-Laplacian optimization problem.

Colors show vertex-wise MfO energy contribution

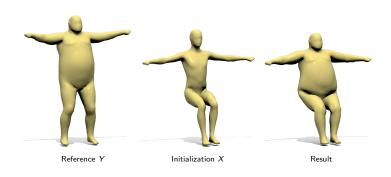
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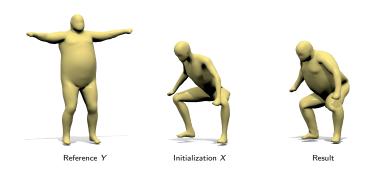
Convergence of our method in the shape-from-Laplacian optimization problem using different initializations.



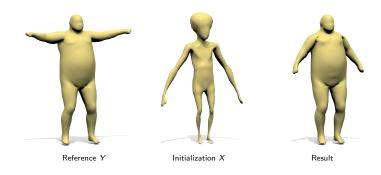
"Modify X such that  $\mathbf{L}_X$  becomes as similar as possible to reference Laplacian  $\mathbf{L}_Y$ "



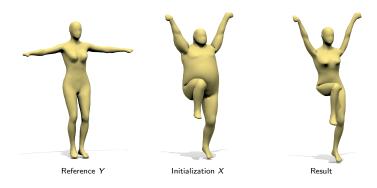
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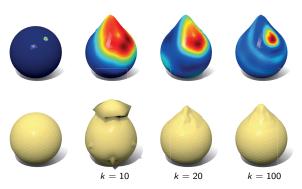
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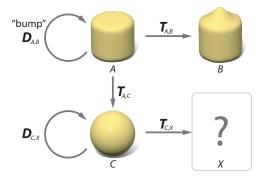
#### Sensitivity to map quality

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



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

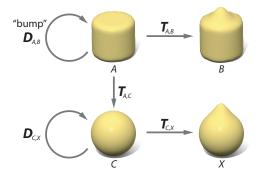
#### Shape-from-difference operator



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

$$\min_{\boldsymbol{\mathsf{X}}} \quad \|\boldsymbol{\mathsf{D}}_{\boldsymbol{\mathsf{C}},\boldsymbol{\mathsf{X}}}(\boldsymbol{\ell}(\boldsymbol{\mathsf{X}}))\boldsymbol{\mathsf{T}}_{\boldsymbol{\mathsf{A}},\boldsymbol{\mathsf{C}}} - \boldsymbol{\mathsf{T}}_{\boldsymbol{\mathsf{A}},\boldsymbol{\mathsf{C}}}\boldsymbol{\mathsf{D}}_{\boldsymbol{\mathsf{A}},\boldsymbol{\mathsf{B}}}\|$$

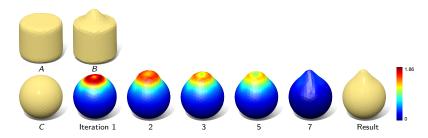
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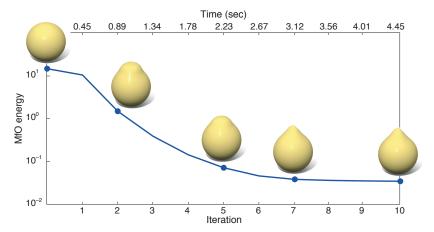
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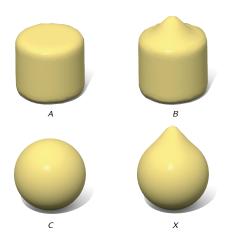
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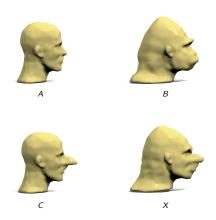
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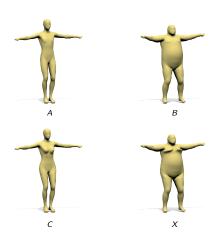
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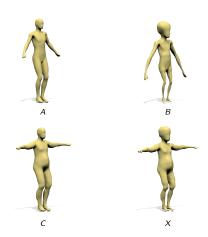
"Find X such that the difference operator between C, X is as similar as possible to the given difference operator between A, B"



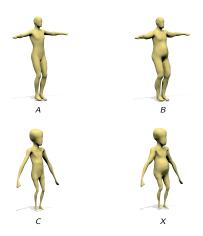
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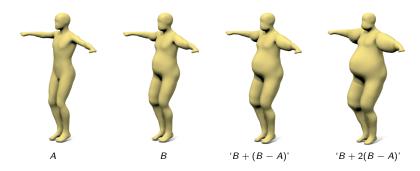


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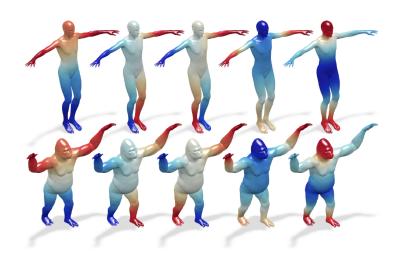


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### Shape exaggeration



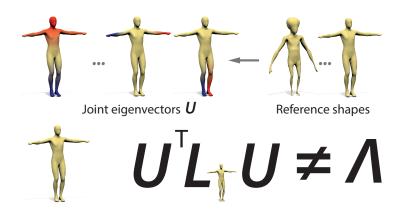
Shape exaggeration obtained by applying the difference operator between A, B to B several times.



Kovnatsky et al., 2013

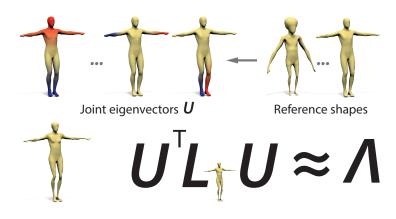


What is the shape whose Laplacian is diagonalized by the joint eigenvectors?



Given an orthonormal basis  $\mathbf{U}=(\mathbf{u}_1,\ldots,\mathbf{u}_r)$  on  $\mathcal{F}(X)$  and  $\mathbf{\Lambda}=\mathrm{diag}(\lambda_1,\ldots,\lambda_r)$ , deform shape X such that  $\mathbf{U}$  is an (approximate) eigenbasis of its Laplacian

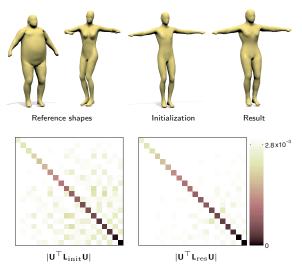
$$\min_{\mathbf{X}} \|\mathbf{W}(\ell(\mathbf{X}))\mathbf{U} - \mathbf{A}(\ell(\mathbf{X}))\mathbf{U}\mathbf{\Lambda}\| + \mu \|\ell - \ell_0\|$$



Given an orthonormal basis  $\mathbf{U}=(\mathbf{u}_1,\ldots,\mathbf{u}_r)$  on  $\mathcal{F}(X)$  and  $\mathbf{\Lambda}=\mathrm{diag}(\lambda_1,\ldots,\lambda_r)$ , deform shape X such that  $\mathbf{U}$  is an (approximate) eigenbasis of its Laplacian

$$\min_{\mathbf{X}} \|\mathbf{W}(\ell(\mathbf{X}))\mathbf{U} - \mathbf{A}(\ell(\mathbf{X}))\mathbf{U}\mathbf{\Lambda}\| + \mu \|\ell - \ell_0\|$$

#### 'Intrinsic average shape' by shape-from-eigenvectors



"Modify initial shape such that its Laplacian is diagonalized by a given basis U"

#### Conclusions

- New generic framework for shape-from-operator inverse problems
- Variety of applications in shape editing
- Other intrinsic operators
- Other shape representations
- Different solutions to shape-from-metric problem