

# Supplementary material: An Elastic Basis for Spectral Shape Correspondence

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## ACM Reference Format:

Florine Hartwig, Josua Sassen, Omri Azencot, Martin Rumpf, and Mirela Ben-Chen. 2023. Supplementary material: An Elastic Basis for Spectral Shape Correspondence. In *Special Interest Group on Computer Graphics and Interactive Techniques Conference Conference Proceedings (SIGGRAPH '23 Conference Proceedings)*, August 6–10, 2023, Los Angeles, CA, USA. ACM, New York, NY, USA, 4 pages. <https://doi.org/10.1145/3588432.3591518>

## 1 TECHNICAL COMPUTATIONS

### 1.1 Proofs of Lemmata

LEMMA 1. Let  $F \in \mathbb{R}^{n,m}$  with  $n, m > 0$  be a linear operator between two finite-dimensional Hilbert Spaces and  $\|\cdot\|$  the corresponding Hilbert–Schmidt norm then

a) for all injective  $\Phi_k \in \mathbb{R}^{n,k}$ ,  $k > 0$

$$\|F\|^2 = \|\Phi_k \Phi_k^\dagger F\|^2 + \|(I - \Phi_k \Phi_k^\dagger) F\|^2$$

b) and for all injective  $\Phi_k \in \mathbb{R}^{m,k}$ ,  $k > 0$

$$\|F\|^2 = \|F \Phi_k \Phi_k^\dagger\|^2 + \|F (I - \Phi_k \Phi_k^\dagger)\|^2.$$

PROOF. Considering an injective  $\Phi_k \in \mathbb{R}^{n,k}$ , we define  $P := \Phi_k \Phi_k^\dagger \in \mathbb{R}^{n,n}$ . We use  $P^2 = P$  and  $P^* = P$ . This holds because  $\Phi_k^\dagger$  is an orthogonal projection with respect to the scalar product. For an explicit calculation, see Lemma 3. We have

$$\|PF\|^2 = \text{tr}((PF)^* PF) = \text{tr}(F^* PPF) = \text{tr}(F^* PF)$$

and similar

$$\|(I - P)F\|^2 = \text{tr}(F^* (I - P) (I - P) F) = \text{tr}(F^* (I - P) F).$$

Using the additivity of the trace, we arrive at the statement a). Statement b) follows similarly using the invariance under cyclic permutations of the trace.  $\square$

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SIGGRAPH '23 Conference Proceedings, August 6–10, 2023, Los Angeles, CA, USA  
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ACM ISBN 979-8-4007-0159-7/23/08.  
<https://doi.org/10.1145/3588432.3591518>

Statement b) is an orthogonal splitting of the source space of the operator  $F$ . For this to hold, it is important to consider the Hilbert–Schmidt norm. A weighted Frobenius norm would only reflect the correct scalar product on the target space.

LEMMA 2. Let  $X \in \mathbb{R}^{m,k}$ ,  $Y \in \mathbb{R}^{n,k}$  be linear operators between finite-dimensional Hilbert spaces with scalar products  $G_1 \in \mathbb{R}^{k,k}$  and  $G_2 \in \mathbb{R}^{n,n}$ .

- a) if  $G_2$  is diagonal the minimization  $\arg\min_{\Pi \in \Pi} \|\Pi^T X - Y\|^2$  is row separable,  
b) if  $A \in \mathbb{R}^{n,n}$  is a positive definite diagonal matrix and  $G_2$  is diagonal the minimization  $\arg\min_{\Pi \in \Pi} \|X^T \Pi A - Y^T\|^2$  is column separable.

To obtain Lemma 4.2 in the main text, we set  $X = \Phi_{2,k} C_{12}^*$ ,  $Y = \Phi_{1,k}$ ,  $G_1 = M_{1,k}$ , and  $G_2 = M_1$  and apply statement a). Similarly, we set  $X^T = C_{12} \Phi_{2,k}^\dagger M_2^{-1}$ ,  $Y^T = \Phi_{1,k}^\dagger$ ,  $G_1 = M_{1,k}$ , and  $G_2 = A = M_1$  and apply statement b) to obtain the corresponding statement in Section 4.3 in the main text on the dual perspective.

PROOF. We first relate the Hilbert–Schmidt norm  $\|F\|^2$  of a general operator  $F$  between finite-dimensional Hilbert spaces with scalar products  $G$  and  $\tilde{G}$ , respectively, to the usual Frobenius norm  $\|\cdot\|_2$ . This reads as

$$\begin{aligned} \|F\|^2 &:= \text{tr}(G^{-1} F^T \tilde{G} F) \\ &= \text{tr}(\sqrt{G^{-1}} F^T \sqrt{\tilde{G}} \sqrt{\tilde{G}} F \sqrt{G^{-1}}) \\ &= \left\| \sqrt{\tilde{G}} F \sqrt{G^{-1}} \right\|_2^2, \end{aligned}$$

where  $\sqrt{B}$  denotes the square root of positive-definite matrices  $B$ . Applying this to the minimization in a), we can rewrite it as

$$\arg\min_{\Pi \in \Pi} \left\| \sqrt{G_2} \left( \Pi^T X \sqrt{G_1^{-1}} - Y \sqrt{G_1^{-1}} \right) \right\|_2^2$$

As  $\Pi \in \Pi$ , we have that each column of  $\Pi \in \{0, 1\}^{m,n}$  has exactly one non-zero entry. Hence,  $\Pi^T X$  is a row permutation of  $X$ . As  $G_2$  is diagonal by assumption, the factor  $\sqrt{G_2}$  is weighting the matrices row-wise and can be omitted. The minimization can then be solved

row-wise by setting  $\Pi(i, j) = 1$  if and only if

$$i = \operatorname{argmin}_{r \in \{1, \dots, m\}} \left| \sqrt{G_1^{-1}} \left( X^T e_r - Y^T e_j \right) \right|_2^2$$

for all  $j = 1, \dots, n$ , which is the same as

$$i = \operatorname{argmin}_{r \in \{1, \dots, m\}} \left| G_1^{-1} \left( X^T e_r - Y^T e_j \right) \right|_{G_1}^2.$$

For statement b), we rewrite the minimization as

$$\operatorname{argmin}_{\Pi \in \Pi} \left\| \sqrt{G_1} \left( X^T \Pi A - Y^T \right) \sqrt{G_2^{-1}} \right\|_2^2$$

Now,  $X\Pi$  is a permutation of the columns of  $X$ . As  $\sqrt{G_2}$  and  $A^{-1}$  are diagonal and multiplication from the right is weighting the columns, we can solve the minimization by setting  $\Pi(i, j) = 1$  if and only if

$$i = \operatorname{argmin}_{r \in \{1, \dots, m\}} \left| \sqrt{G_1} \left( X^T e_r - Y^T A^{-1} e_j \right) \right|_2^2$$

for all  $j = 1, \dots, n$ .  $\square$

**LEMMA 3 (ORTHOGONAL PROJECTION).** *The operator  $\Phi_k \Phi_k^\dagger \in \mathbb{R}^{n,n}$  is self-adjoint for an injective  $\Phi_k \in \mathbb{R}^{n,k}$  with  $n > k > 0$ , i.e. it holds  $(\Phi_k \Phi_k^\dagger)^* = \Phi_k \Phi_k^\dagger$ .*

**PROOF.** Let us recall the definition  $\Phi_k^\dagger = G_k^{-1} \Phi_k^T G$  with  $G_k = \Phi_k^T G \Phi_k$ , where  $G \in \mathbb{R}^{n,n}$  represents the scalar product of the Hilbert space. We have

$$\left( \Phi_k \Phi_k^\dagger \right)^* = G^{-1} (\Phi_k^\dagger)^T \Phi_k^T G = G_k^{-1} G_k \Phi_k G_k^{-1} \Phi_k^T G = \Phi_k \Phi_k^\dagger$$

$\square$

## 1.2 Computation of the adjoint

Computation of the adjoint (Formula (7))

$$\begin{aligned} C_{12}^* &= M_{2,k}^{-1} \Phi_{2,k}^T P_{12}^T (\Phi_{1,k}^\dagger)^T M_{1,k} \\ &= \left( M_{2,k}^{-1} \Phi_{2,k}^T M_2 \right) M_2^{-1} P_{12}^T (\Phi_{1,k}^\dagger)^T M_{1,k} \\ &= \left( M_{2,k}^{-1} \Phi_{2,k}^T M_2 \right) \left( M_2^{-1} P_{12}^T M_1 \right) \Phi_{1,k} M_{1,k}^{-1} M_{1,k} = \Phi_{2,k}^\dagger P_{12}^* \Phi_{1,k}, \end{aligned}$$

where we used  $(\Phi_{1,k}^\dagger)^T = M_1 \Phi_{1,k} M_{1,k}^{-1}$ .

## 2 ADDITIONAL VISUALIZATION

### 2.1 Additional qualitative results

In Figure 1 we give additional qualitative results for the remaining methods in Figure 5 of the main paper, see Section 5.1 for details. In Figure 2 we show a colormap representation for the experiment described in Section 5.2 of the main document. Moreover, we show the results for a shape pair with median error of our method in Figure 3. In this example the extrinsic features of the shapes vary strongly.

**Table 1: Runtime report (in sec).**

model (number vertices)	LB basis	Ours
Cat Lion (ca. 6k)	2.82/66.74	33.59/87.96
Homer (ca. 6.5k)	1.53/24.06	22.96/33.29
Head (ca. 15k)	2.31/42.29	38.28/131.88

### 2.2 Qualitative results for different values of $k$

We show qualitative results for the iterative process initialized by a ground-truth vertex map as described in 5.4.2 in the main paper in Figure 4.

### 2.3 Runtime analysis

We report runtime values in Table 1 for the experiments of the main document shown in Figures 5 and 6. We distinguish between the computation of the basis functions (first value) and the iterative method (second value).

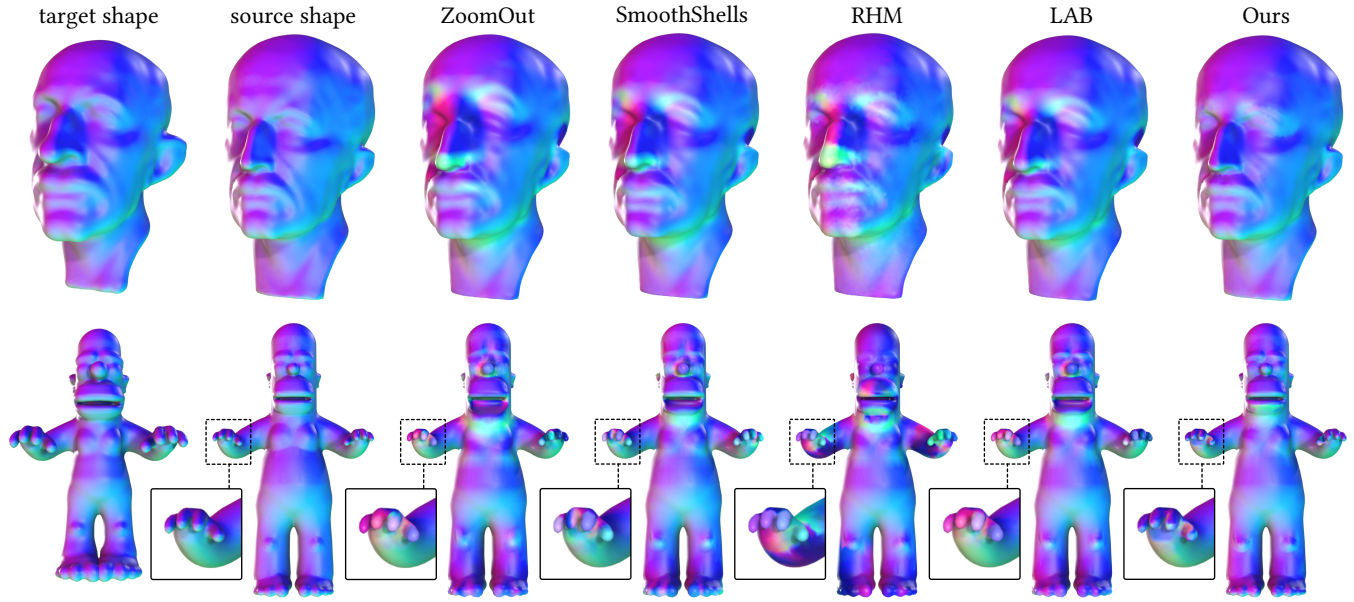


Figure 1: Additional qualitative results for Figure 5 of the main paper. See Section 5.1 in main paper for details and Figure 5 of the main paper for a quantitative evaluation of these results.

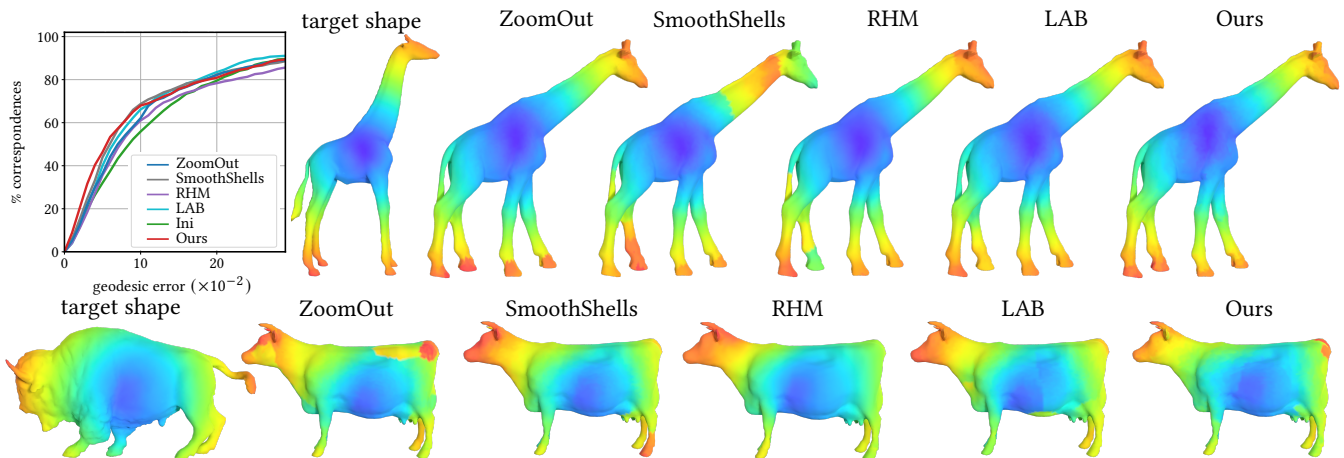


Figure 2: Colormap representation of the results of Figure 8 in the main paper.

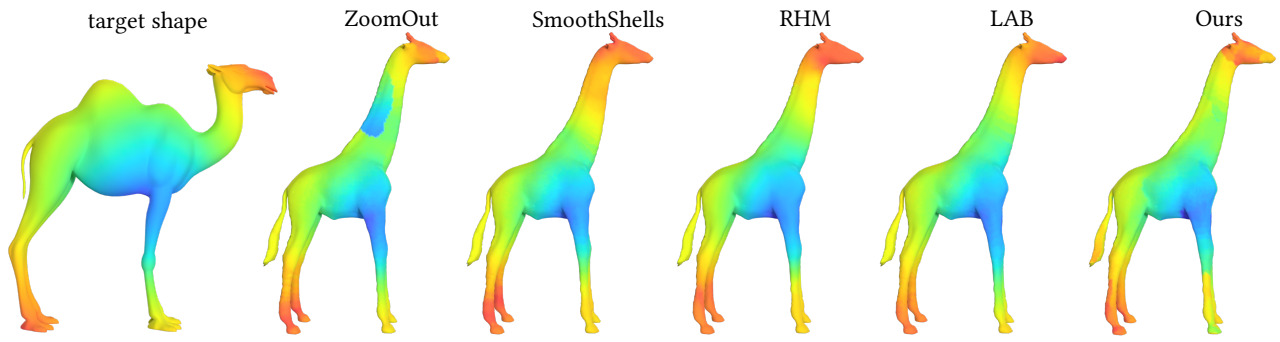


Figure 3: Correspondence of Shrec20 with median error of our method, see Section 5.4.2 for details.

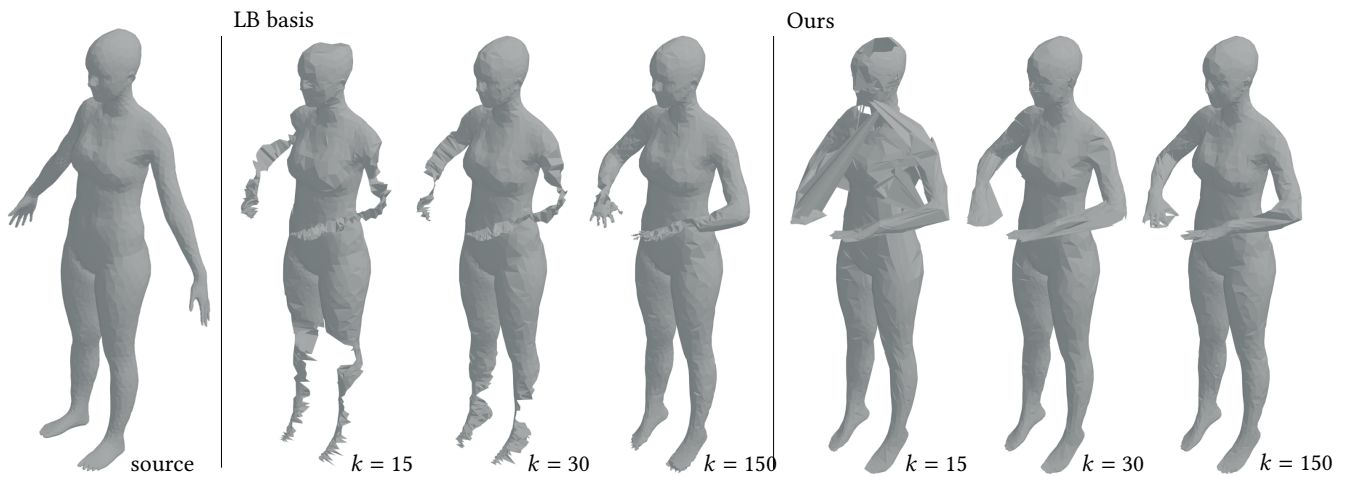


Figure 4: Qualitative visualization of results of one correspondence for different values of  $k$  for the experiment in Figure 10 of the main paper. We visualize the computed correspondence by showing the image of the resulting vertex map.