Functional maps + products = \heartsuit

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Faceshift (acquired by Apple in 2015)

Pointwise correspondence



Point-wise map $t: \mathcal{X} \to \mathcal{Y}$

Functional correspondence



Functional map $T: L^2(\mathcal{X}) \to L^2(\mathcal{Y})$

Ovsjanikov, Ben-Chen, Solomon, Butscher, Guibas 2012

Functional maps in spectral domain



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Functional maps in spectral domain



where A, B are Fourier coefficients of corresponding 'probe' functions $g_i\approx Tf_i \qquad i=1,\ldots,q\geq k$

Ovsjanikov, Ben-Chen, Solomon, Butscher, Guibas 2012

Laplacian eigenbasis



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



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- Is Laplacian eigenbasis the best way to represent functional maps?

Theorem Functional map $T: L^2(\mathcal{X}) \to L^2(\mathcal{Y})$ is a pointwise map iff $T(f \cdot h) = (Tf) \cdot (Th)$

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Kishor, Manhas 1993; Shtern, Kimmel 2013 (product-based pointwise descriptors)

$$f \approx \sum_{i=0}^{k} a_i \phi_i$$

$$f \approx \sum_{i=0}^{k} a_i \phi_i + \sum_{j=1}^{r} \tilde{a}_j \prod_{l=1}^{r_j} \phi_{i_{jl}} \qquad i_{jl} \in \{1, \dots, k\}$$

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- Product basis is linearly dependent
- Orthogonality is lost
- Higher orders r become unstable
- Finding optimal approximation that minimizes the number of products used is NP-hard

Example: 1D product basis





Approximation of a step function (black) using standard (blue) and product (red) bases of order r = 2



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Approximation of the shape 3D coordinates in standard and product bases



Reconstruction of the shape 3D coordinates using standard and product bases of different order (k = 20)





$$f \approx \sum_{i=0}^{k} a_i \phi_i + \sum_{i,j=1}^{k} \tilde{a}_{ij} \phi_i \cdot \phi_j$$
$$Tf \approx T\left(\sum_{i=0}^{k} a_i \phi_i + \sum_{i,j=1}^{k} \tilde{a}_{ij} \phi_i \cdot \phi_j\right)$$

$$Tf \approx \sum_{i=0}^{k} a_i T\phi_i + \sum_{i,j=1}^{k} \tilde{a}_{ij} T(\phi_i \cdot \phi_j)$$

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$$\tilde{\mathbf{C}}(\mathbf{C}) = \begin{bmatrix} \mathbf{C} & \phi_{00} \mathbf{c}_0^\top \otimes \mathbf{C}_{01} + \phi_{00} \begin{bmatrix} \mathbf{0} \\ \mathbf{C}_{11} \end{bmatrix} \otimes \mathbf{c}_0^\top \\ \mathbf{C}_{11} \otimes \mathbf{C}_{11} \end{bmatrix}$$

is matrix of size $(k^2+k+1)\times (k^2+k+1)$ expressed in terms of ${\bf C},$ and

$$\mathbf{C}_{11} = \begin{bmatrix} c_{11} & \dots & c_{1k} \\ \vdots & \vdots & \vdots \\ c_{k1} & \dots & c_{kk} \end{bmatrix} \quad \mathbf{C}_{01} = \begin{bmatrix} c_{01} & \dots & c_{0k} \\ \vdots & \vdots & \vdots \\ c_{k1} & \dots & c_{kk} \end{bmatrix} \quad \mathbf{c}_{0}^{\top} = \begin{bmatrix} c_{00} & \dots & c_{0k} \end{bmatrix}$$

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 $\Rightarrow (k^2+k+1)^2$ coefficients, but only $(k+1)^2$ degrees of freedom!

Standard vs product bases



Function approximation and transfer error using standard and product bases

Example of correspondence on FAUST dataset



Source

Standard basis

Product basis

Correspondence (shown with matching colors) and correspondence error on SCAPE shapes using standard and product bases

Nongeng, Melzi, Rodolà, Castellani, B, Ovsjanikov 2018; data: Bogo et al. 2014

Correspondence quality



Quality of functional maps computed with standard and product bases on FAUST (left) and TOSCA (right) shapes

Future directions

• Instead of improving a given functional map, finding pointwise functional maps by solving the non-linear problem

$$\min_{\mathbf{C} \in \mathbb{R}^{k \times k}} \| \mathbf{B} - \tilde{\mathbf{C}}(\mathbf{C}) \mathbf{A} \|_{\mathrm{F}}^2$$

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• Instead of improving a given functional map, finding pointwise functional maps by solving the non-linear problem

$$\min_{\mathbf{C} \in \mathbb{R}^{k \times k}} \| \mathbf{B} - \tilde{\mathbf{C}}(\mathbf{C}) \mathbf{A} \|_{\mathrm{F}}^2$$

• More general definition of products (potentially combined with learning)

Issues with functional maps

- Finding correspondence boils down to solving a linear problem
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Correspondence in the product space



Functional map $T_{\mu} : \mathcal{F}(\mathcal{M}) \to \mathcal{F}(\mathcal{N})$ associated with a density $\mu \in L^1(\mathcal{M} \times \mathcal{N})$ on the product manifold $(\mathcal{M} \times \mathcal{N}, g_{\mathcal{M}} \oplus g_{\mathcal{N}})$

$$T_{\mu}(g)(x) = \int_{\mathcal{N}} g(y)\mu(x,y)\mathrm{d}y$$

Laplacian eigenbasis on product manifold

Theorem Let $\mathcal{M} \times \mathcal{N}$ be a product manifold and let $\Delta_{\mathcal{M} \times \mathcal{N}} \xi = \gamma \xi$ Then, there exist ϕ, ψ and α, β s.t. $\Delta_{\mathcal{M}} \phi = \alpha \phi$ and $\Delta_{\mathcal{N}} \psi = \beta \psi$ and $\gamma = \alpha + \beta$ $\xi = \phi \wedge \psi$

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Berger, Gauduchon, Mazet 1971

Representation equivalence

Theorem Let $c_{ij} = \langle \phi_i, T_\mu(\psi_j) \rangle_{\mathcal{M}}$ be the representation of T_μ in orthogonal bases $\{\phi_i\}_{i \geq 1}$, $\{\psi_i\}_{i \geq 1}$ and let $p_{ij} = \langle \phi_i \wedge \psi_j, \mu \rangle_{\mathcal{M} \times \mathcal{N}}$ such that $\mu = \sum_{ij} (\phi_i \wedge \psi_j) p_{ij}$. Then $c_{ij} = p_{ij}$ for all i, j.

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Rodolà, Lähner, BB, Solomon 2018

Representation efficiency

$$\mu = \sum_{\ell=0}^{k} p_{\ell} \xi_{\ell}$$



 $\phi_0 \wedge \psi_0$



 $\phi_1 \wedge \psi_0$



 $\phi_0 \wedge \psi_1$

Separable basis



 $\phi_2 \wedge \psi_0$



 $\phi_3 \wedge \psi_0$

Representation efficiency

$$\mu = \sum_{\ell=0}^{k} p_{\ell} \xi_{\ell}$$



Eigenfunctions of the Hamiltonian operator $H=\Delta_{\mathcal{M}\times\mathcal{N}}+V,$ where V is the localization potential

Rodolà, Lähner, BB, Solomon 2018; Choukroun et al. 2017; Melzi, Rodolà, Castellani, B 2017



Delta function transfer using functional map on the product space computed in **separable basis**. Groundtruth correspondence shown in black.



Delta function transfer using functional map on the product space computed in **localized basis (90% area)**. Groundtruth correspondence shown in black.



Delta function transfer using functional map on the product space computed in **localized basis (25% area)**. Groundtruth correspondence shown in black.



Delta function transfer using functional map on the product space computed in **localized basis (5% area)**. Groundtruth correspondence shown in black.



Delta function transfer using functional map on the product space computed in **localized basis (1% area)**. Groundtruth correspondence shown in black.



Quality of correspondence on product manifold using different basis localization



Delta function transfer using functional map on the product space computed in **separable basis**.



Delta function transfer using functional map on the product space computed in **localized basis (15% area)**.



Delta function transfer using functional map on the product space computed in **localized basis (10% area)**.



Quality of correspondence on product manifold using different basis localization

Two completely different uses of products yield novel and interesting representations of functional maps and shed new light on old problems

Functional maps + products = \heartsuit



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With the generous support of







Thank you!