# Functional maps + products $=\bigcirc$ 

## Michael Bronstein

USI Lugano / Imperial College / Intel

SIAM AM, Portland, 13 July 2018


2005

\$100 2010

ㅁ.
\$20
2014

\$17
2017

Faceshift (acquired by Apple in 2015)

## Pointwise correspondence



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

## Functional correspondence



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

## Functional maps in spectral domain



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

## Functional maps in spectral domain

$\square$



$\mathbf{C}_{k \times k}$

$\mathbf{A}_{k \times q}=\left(\left\langle\phi_{i}, f_{j}\right\rangle_{\mathcal{X}}\right)$
where $\mathbf{A}, \mathbf{B}$ are Fourier coefficients of corresponding 'probe' functions

$$
g_{i} \approx T f_{i} \quad i=1, \ldots, q \geq k
$$

## 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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- Resulting map is not pointwise! Recovering a pointwise map from functional map is a hard problem!
- Is Laplacian eigenbasis the best way to represent functional maps?


## Pointwise maps $=$ product preserving maps

Theorem Functional map $T: L^{2}(\mathcal{X}) \rightarrow L^{2}(\mathcal{Y})$ is a pointwise map iff

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T(f \cdot h)=(T f) \cdot(T h)
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for all $f, h \in L^{2}(\mathcal{X})$

Kishor, Manhas 1993

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Solution: represent $T$ in product bases

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

# Functional maps in product bases 

## Product basis

$$
f \approx \sum_{i=0}^{k} a_{i} \phi_{i}
$$

Nongeng, Melzi, Rodolà, Castellani, B, Ovsjanikov 2018

## Product basis

$$
f \approx \sum_{i=0}^{k} a_{i} \phi_{i}+\sum_{j=1}^{r} \tilde{a}_{j} \prod_{l=1}^{r_{j}} \phi_{i_{j l}} \quad i_{j l} \in\{1, \ldots, k\}
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- Adds higher frequency information: for trigonometric bases 2nd order products max double the frequency since

$$
\left.\left.\cos (n x) \cdot \cos (m x)=\frac{1}{2}[\cos ((n+m) x))+\cos ((n-m) x)\right)\right]
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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



## 1D function approximation: standard vs product basis



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

## 1D function approximation: standard vs product basis



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

## 1D function approximation: standard vs product basis



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

## 3D function approximation: standard vs product basis

Standard basis


Original


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

## 3D function approximation: standard vs product basis



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

## Functional maps in product bases



Nongeng, Melzi, Rodolà, Castellani, B, Ovsjanikov 2018

## Functional maps in product bases



## Functional maps in product bases

$$
f \approx \sum_{i=0}^{k} a_{i} \phi_{i}+\sum_{i, j=1}^{k} \tilde{a}_{i j} \phi_{i} \cdot \phi_{j}
$$

## Functional maps in product bases

$$
T f \approx T\left(\sum_{i=0}^{k} a_{i} \phi_{i}+\sum_{i, j=1}^{k} \tilde{a}_{i j} \phi_{i} \cdot \phi_{j}\right)
$$

## Functional maps in product bases

$$
T f \approx \sum_{i=0}^{k} a_{i} T \phi_{i}+\sum_{i, j=1}^{k} \tilde{a}_{i j} T\left(\phi_{i} \cdot \phi_{j}\right)
$$

## Functional maps in product bases

$$
T f \approx \sum_{i=0}^{k} a_{i} T \phi_{i}+\sum_{i, j=1}^{k} \tilde{a}_{i j} T\left(\phi_{i}\right) \cdot T\left(\phi_{j}\right)
$$

## Functional maps in product bases

$$
T f \approx \sum_{i, j=0}^{k} a_{i} c_{i j} \psi_{j}+\sum_{i, j=1}^{k} \sum_{l, l^{\prime}=0}^{k} \tilde{a}_{i j} c_{i l} c_{i l^{\prime}} \psi_{l} \cdot \psi_{l^{\prime}}
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$$

## Structure of matrix $\tilde{\mathbf{C}}$




Nongeng, Melzi, Rodolà, Castellani, B, Ovsjanikov 2018

## Structure of matrix $\tilde{\mathbf{C}}$

$$
\tilde{\mathbf{C}}(\mathbf{C})=\left[\begin{array}{cc}
\mathbf{C} & \phi_{00} \mathbf{c}_{0}^{\top} \otimes \mathbf{C}_{01}+\phi_{00}\left[\begin{array}{c}
\mathbf{0} \\
\mathbf{C}_{11}
\end{array}\right] \otimes \mathbf{c}_{0}^{\top} \\
\mathbf{C}_{11} \otimes \mathbf{C}_{11}
\end{array}\right]
$$

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

$$
\mathbf{C}_{11}=\left[\begin{array}{ccc}
c_{11} & \ldots & c_{1 k} \\
\vdots & \vdots & \vdots \\
c_{k 1} & \ldots & c_{k k}
\end{array}\right] \quad \mathbf{C}_{01}=\left[\begin{array}{ccc}
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$$

$\Rightarrow\left(k^{2}+k+1\right)^{2}$ 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

## 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}
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- More general definition of products (potentially combined with learning)


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# Functional maps in product spaces 

## Correspondence in the product space



Source


Target


Product manifold

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

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

## 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 $\phi, \psi$ and $\alpha, \beta$ s.t. $\Delta_{\mathcal{M}} \phi=\alpha \phi$ and $\Delta_{\mathcal{N}} \psi=\beta \psi$ and

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\gamma=\alpha+\beta \quad \xi=\phi \wedge \psi
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## Representation equivalence

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

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Rodolà, Lähner, BB, Solomon 2018

## Representation efficiency

$$
\mu=\sum_{\ell=0}^{k} p_{\ell} \xi_{\ell}
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Rodolà, Lähner, BB, Solomon 2018

## Representation efficiency

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Non-separable localized basis

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

## Example: 1D correspondence



Source


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

Rodolà, Lähner, BB, Solomon 2018

## Example: 1D correspondence



Source


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

## Example: 1D correspondence



Source


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

## Example: 1D correspondence



Source


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

## Example: 1D correspondence



Source


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

## Example: 1D correspondence



Quality of correspondence on product manifold using different basis localization

## Example: 2D correspondence



Source


Target

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

## Example: 2D correspondence



Source


Target

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

## Example: 2D correspondence



Source


Target

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

## Example: 2D correspondence



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 $=\bigcirc$


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Thechnion
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With the generous support of

## Thank you!

