

Steerable Kernels for Arbitrarily-Sampled Spaces

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Introduction

- Steerable kernels are efficient, $O(k)$, $k \ll N$.
 - Perform expensive inner products once.
 - Inexpensive weighted sum of inner products span family of desired kernels.
- Many basis functions determined *ad hoc*.
 - (e.g. Specific for oriented edges)
 - Correspondence matrix formalizes discrete mapping.
 - Optimal real-valued basis functions.
 - Tessellation of data space (square, hexagonal, ...)
 - Mapping class (rotation, scaling, ...)

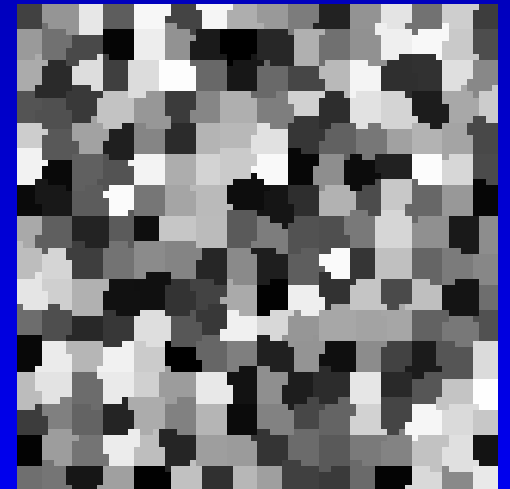
Steerable functions

$$F_{\alpha}(\mathbf{x}) = \sum_{k=1}^N b_k(\alpha) G_k(\mathbf{x})$$

- Steerable function $F_{\alpha}(x)$
- Data vector x
- Deformation parameter α
- Basis functions $G_k(x)$
- Interpolation functions $b_k(\alpha)$

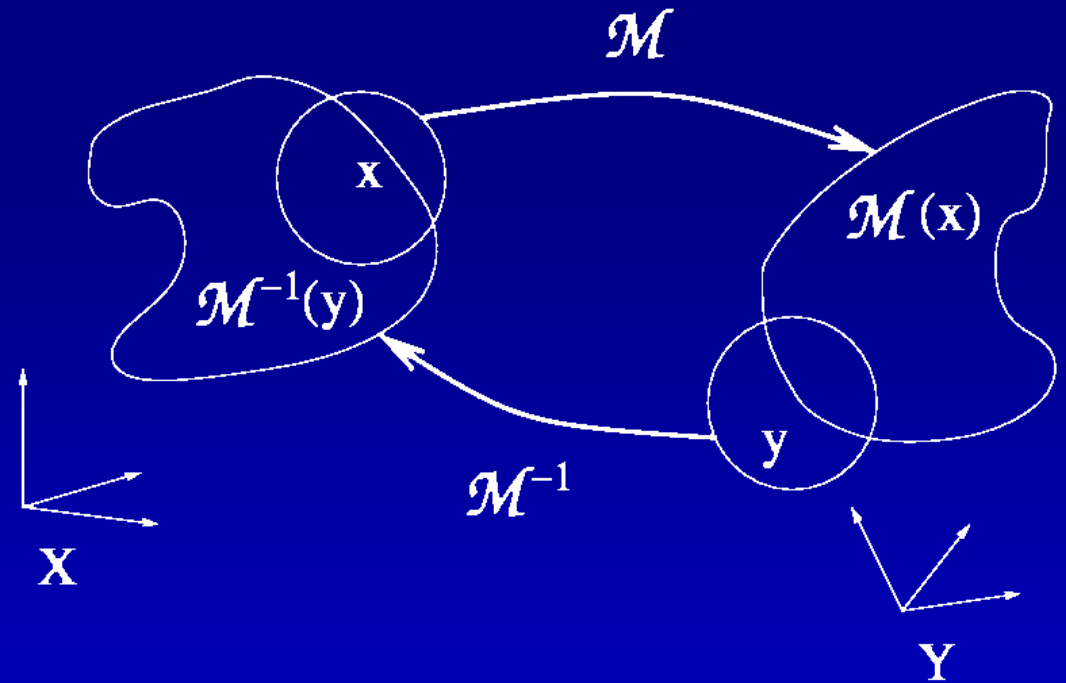
Steerable functions for rotation

- Rotation about center of aperture.
 - Only pixels within inscribed circle can fully rotate.
- Square pixels in windows of 32x32 pixels.
- Hexagonal pixels in windows of $\sim 32 \times 32$ pixels.
- Irregularly-spaced Voronoi cells in 16x16 grid.
 - Voronoi cells' positions perturbed randomly by 0% to 20% of diameter.
 - Can useful basis images still be computed?



Correspondence matrix \mathbf{H}

Encode mapping between each pixel in first view $\langle I_0 \rangle_i$ and each pixel in second view $\langle I_\alpha \rangle_j$.



$$\begin{aligned} \langle \mathbf{H} \rangle_{ij} &\triangleq \mathbb{P} \left([\mathbf{X}]_i \overset{M}{\iff} [\mathbf{Y}]_j \right) \subset [0, 1] \\ &= \frac{\mathcal{A} \left(\mathcal{M}([\mathbf{X}]_i) \cap [\mathbf{Y}]_j \right) + \mathcal{A} \left([\mathbf{X}]_i \cap \mathcal{M}^{-1}([\mathbf{Y}]_j) \right)}{\mathcal{A}([\mathbf{X}]_i) + \mathcal{A}([\mathbf{Y}]_j)} \end{aligned}$$

Decomposing the correspondence matrix

$$\mathbf{H}_\alpha \stackrel{\text{SVD}}{=} \mathbf{U}_0 \mathbf{\Sigma}_0 \mathbf{V}_\alpha^T$$

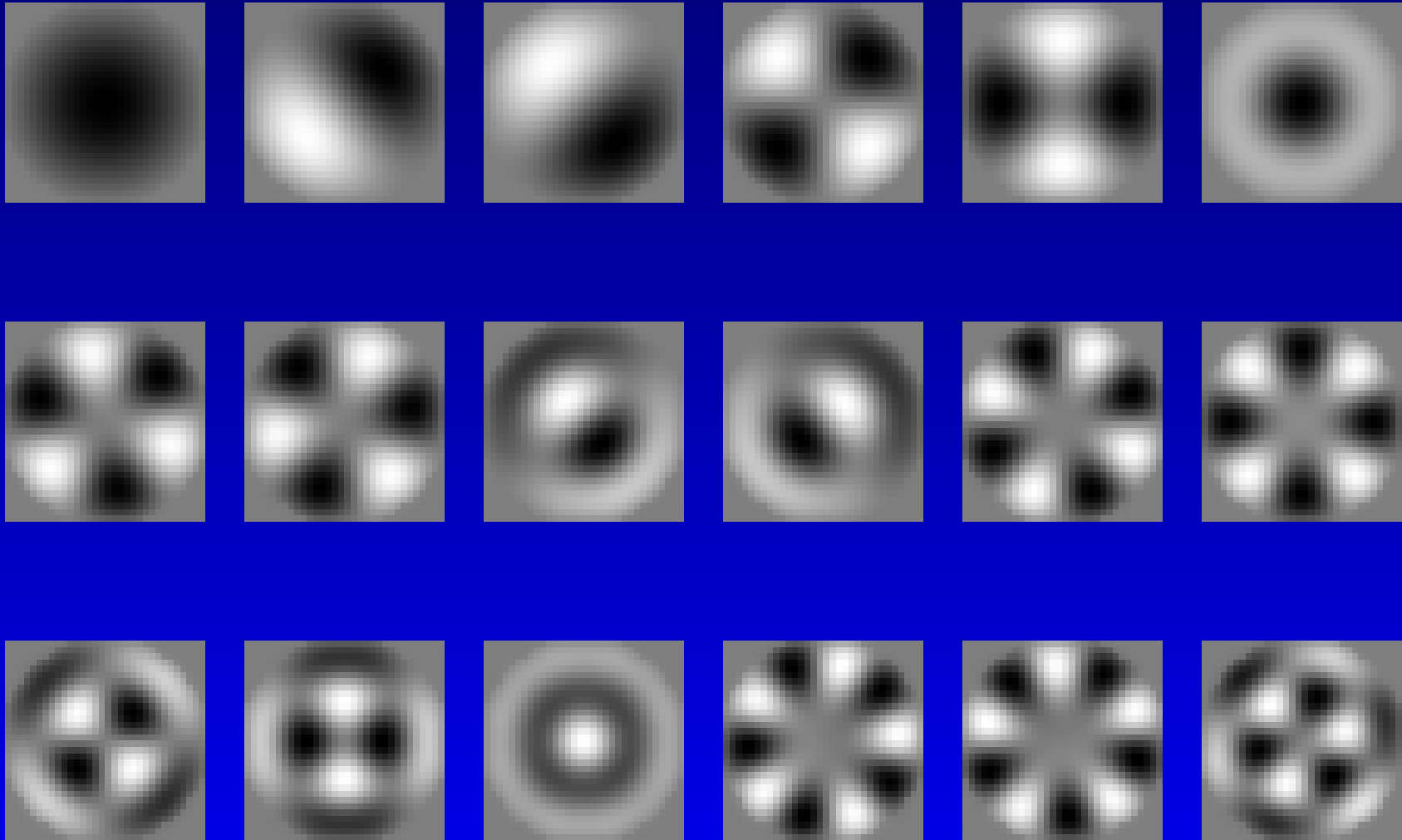
$$\mathcal{I}_0 = \mathbf{\Phi}_\alpha \mathcal{I}_\alpha$$

$$\mathcal{I}_\alpha = \mathbf{\Phi}_\alpha^T \mathcal{I}_0$$

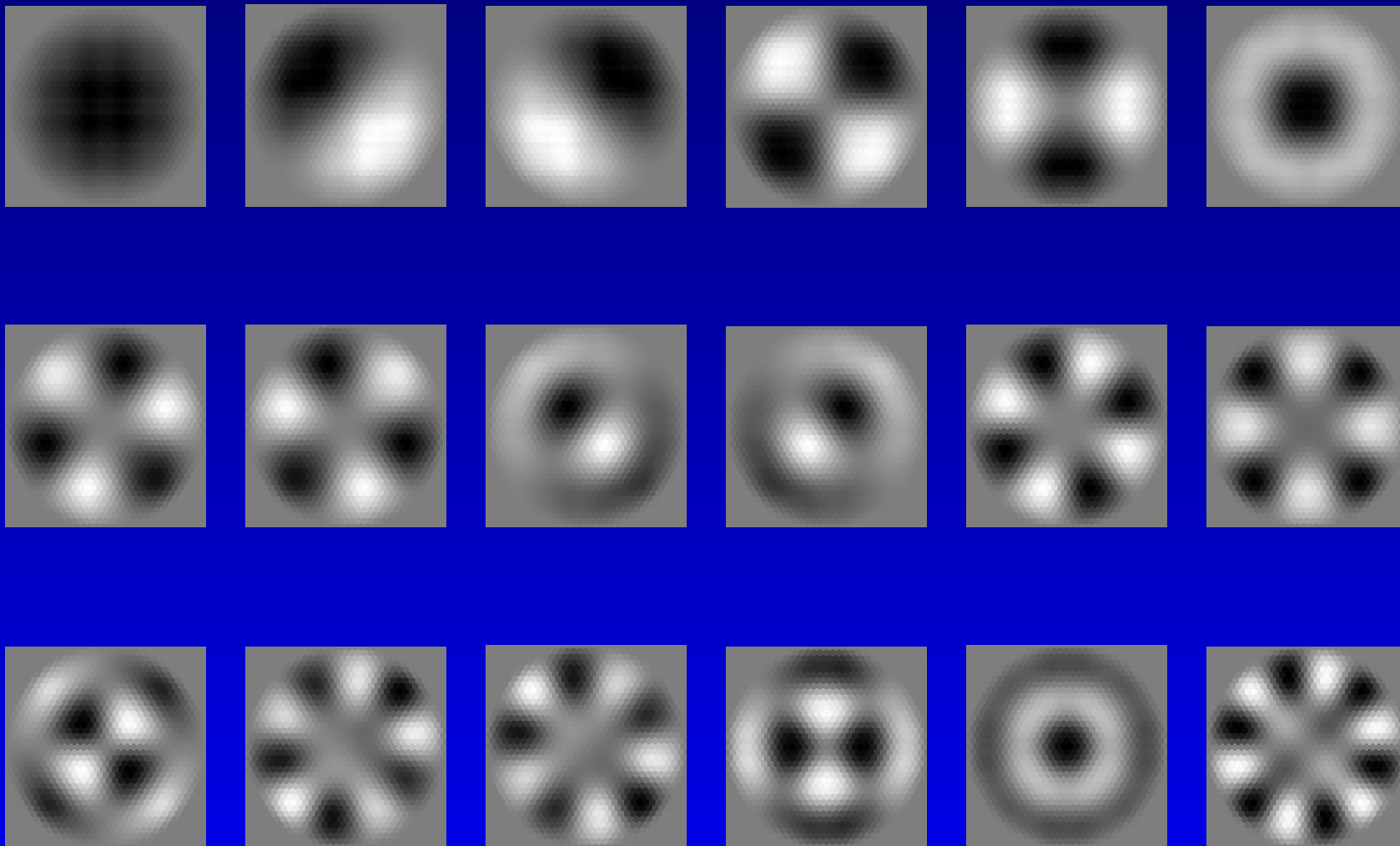
$$\mathcal{I}_0 = \mathbf{U}_0 \mathbf{V}_\alpha^T \mathcal{I}_\alpha$$

$$\mathbf{U}_0^T \mathcal{I}_0 = \underbrace{\vec{w}}_{r \times 1} = \mathbf{V}_\alpha^T \mathcal{I}_\alpha$$

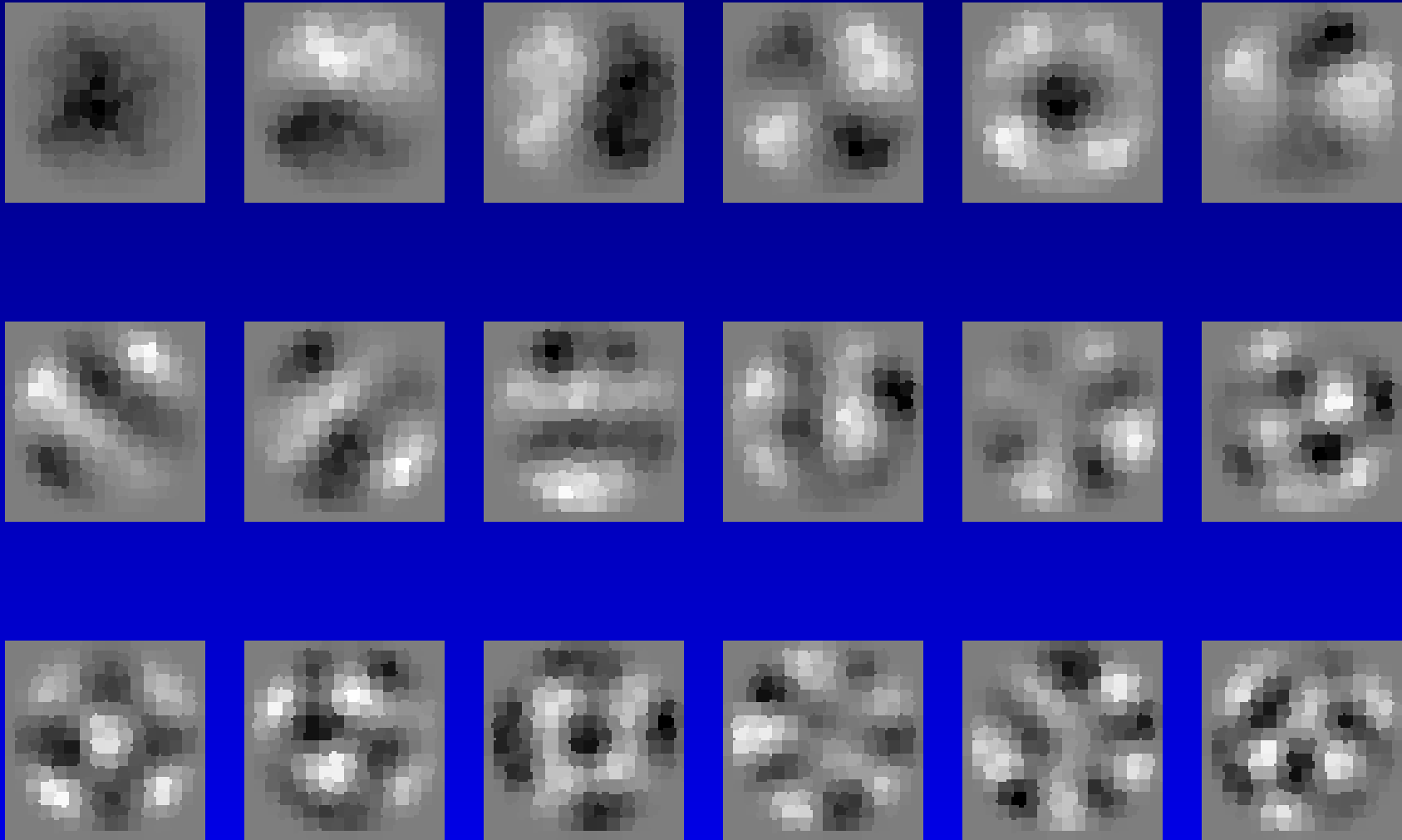
Rectilinear rotation basis fields



Hexagon rotation basis fields



Perturbed rotation basis fields



k-th order approximation

$$\begin{aligned}\mathcal{I}_0 &\approx \hat{\mathbf{U}}_0 \hat{\mathbf{V}}_\alpha^T \mathcal{I}_\alpha \\ \hat{\mathbf{U}}_0^T \mathcal{I}_0 &= \hat{\mathbf{w}}_{k \times 1} = \hat{\mathbf{V}}_\alpha^T \mathcal{I}_\alpha\end{aligned}$$

Solving \mathbf{V}_α without SVD

$$\mathbf{H}_\alpha \stackrel{\text{SVD}}{=} \mathbf{U}_0 \mathbf{\Sigma}_0 \mathbf{V}_\alpha^T$$

$$\begin{aligned} \mathbf{V}_\alpha &= \mathbf{H}_\alpha^T \mathbf{U}_0 \mathbf{\Sigma}_0^+, \\ \langle \mathbf{\Sigma}_0^+ \rangle_{ij} &= \begin{cases} 1 / \langle \mathbf{\Sigma}_0 \rangle_{ij} & , i = j, i, j \leq r \\ 0 & , \text{otherwise} . \end{cases} \end{aligned}$$

V resembles U : deformation interpolants

$$\hat{\mathbf{V}}_{\alpha} \underset{N \times k}{=} \underset{N \times k}{\hat{\mathbf{U}}_0} \underset{k \times k}{\hat{\mathbf{R}}_{\alpha}}$$

$$\hat{\mathbf{R}}_{\alpha} = \hat{\mathbf{U}}_0^T \hat{\mathbf{V}}_{\alpha} = \hat{\mathbf{U}}_0^T \mathbf{H}_{\alpha}^T \hat{\mathbf{U}}_0 \hat{\Sigma}_0^+$$

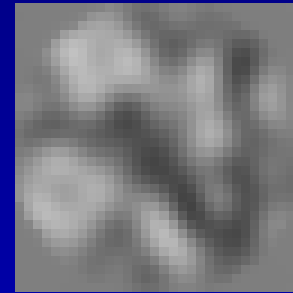
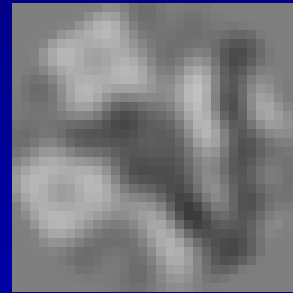
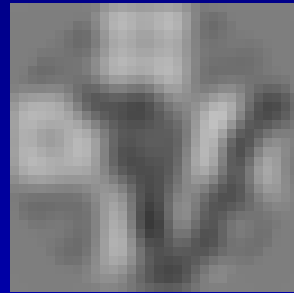
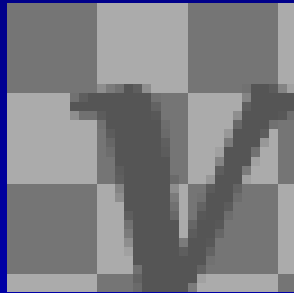
$$\hat{w}_0 = \hat{\mathbf{U}}_0^T \mathcal{I}_0, \quad \hat{\mathcal{I}}_0 = \hat{\mathbf{U}}_0 \hat{w}_0$$

$$\hat{\mathcal{I}}_{\alpha} = \hat{\mathbf{V}}_{\alpha} \hat{w}_0, \quad \tilde{\mathcal{I}}_{\alpha} = \tilde{\mathbf{V}}_{\alpha} \hat{w}_0 = \hat{\mathbf{U}}_0 \tilde{\mathbf{R}}_{\alpha} \hat{w}_0$$

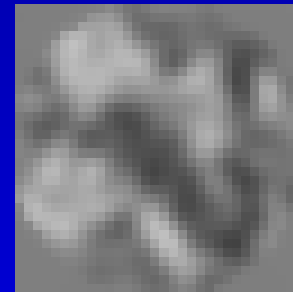
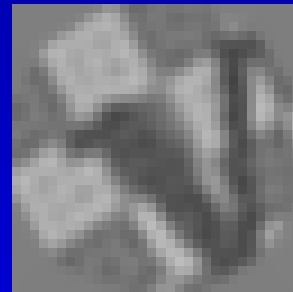
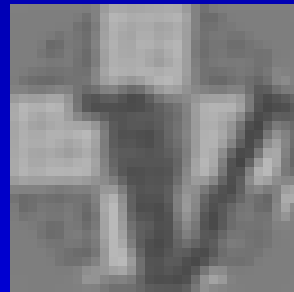
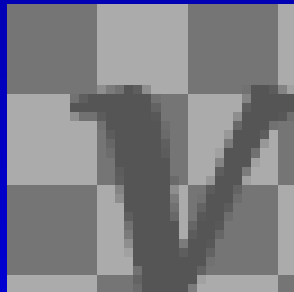
Rectilinear rotation experiment

Rotation $\alpha = 30$ degrees. 1024 pixels.

k=100



k=200



\mathcal{I}_0

$\hat{\mathcal{I}}_0$

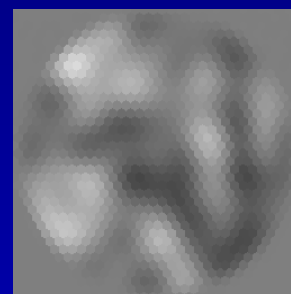
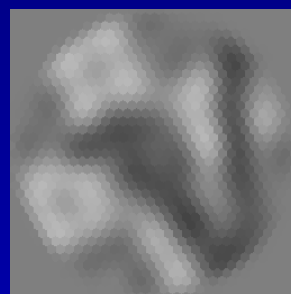
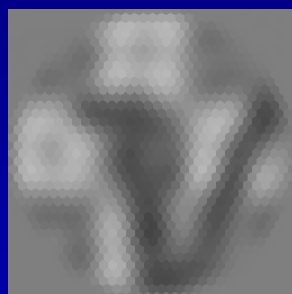
$\hat{\mathcal{I}}_\alpha$

$\tilde{\mathcal{I}}_\alpha$

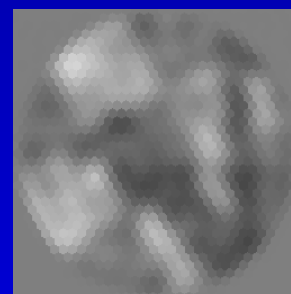
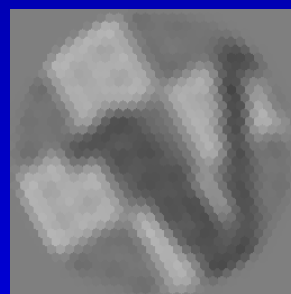
Hexagon rotation experiment

Rotation $\alpha = 30$ degrees. 1202 pixels.

k=100



k=200



\mathcal{I}_0

$\hat{\mathcal{I}}_0$

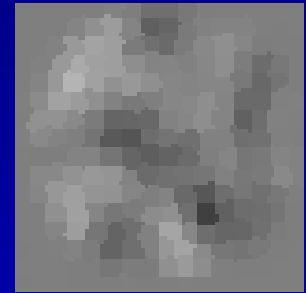
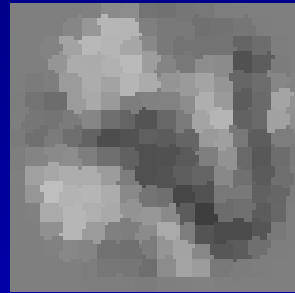
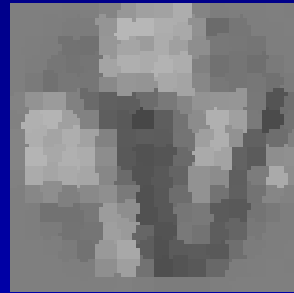
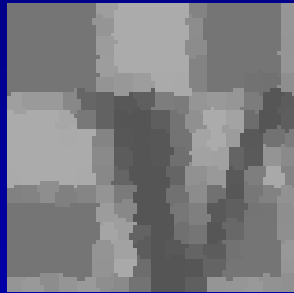
$\hat{\mathcal{I}}_\alpha$

$\tilde{\mathcal{I}}_\alpha$

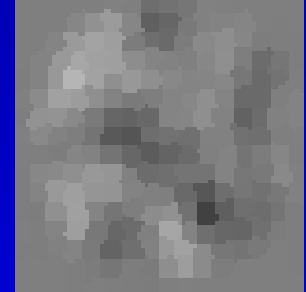
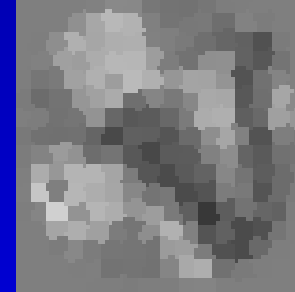
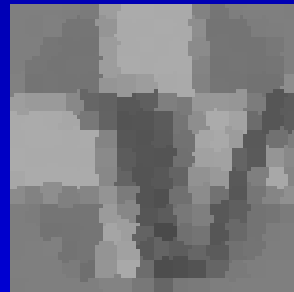
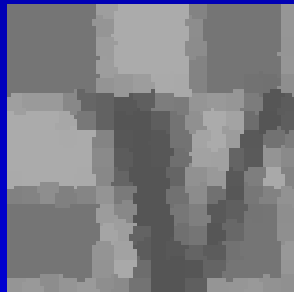
Perturbed rotation experiment

Rotation $\alpha = 30$ degrees. 256 pixels.

k=100



k=200



\mathcal{I}_0

$\hat{\mathcal{I}}_0$

$\hat{\mathcal{I}}_\alpha$

$\tilde{\mathcal{I}}_\alpha$

Future work

- Formalize Degrees of Freedom of mapping \mathbf{M} .
 - Limits information conserved by tessellation, transform.
 - $\text{DOF}(\mathbf{M}) = \text{rank}(\mathbf{H})$.
- 3-D steerable kernels.

Conclusion

- Treat pixels as Voronoi cells.
- Image deformations between sets of Voronoi cells.
 - Without projection into rectilinear space.
- Predicts optimal real valued spectral decomposition.
- Arbitrary sampling of image space.

Thanks