

SAMPLING SCHEMES FOR 2-D SIGNALS WITH FINITE RATE OF INNOVATION USING KERNELS THAT REPRODUCE POLYNOMIALS

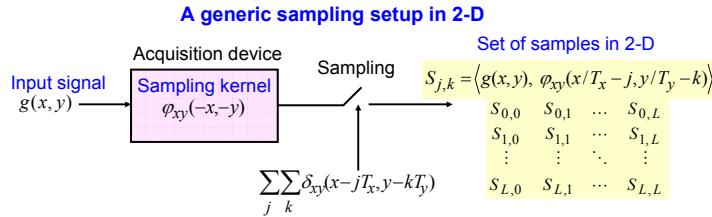
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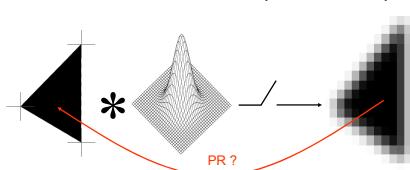
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1. INTRODUCTION AND MOTIVATION



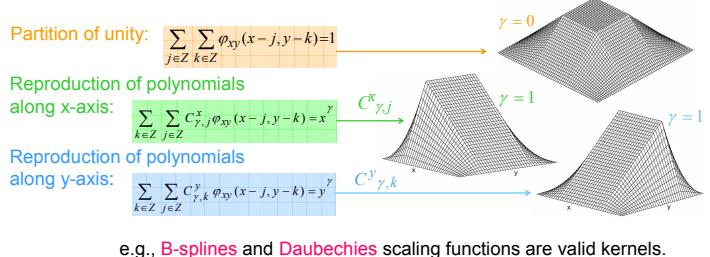
Motivation
Image resolution enhancement and super-resolution photogrammetry.



What signals? Non-bandlimited signals but with a finite number of degrees of freedom (rate of innovation), and thus known as signals with Finite Rate of Innovation (FRI) [Vetterli et al] [1,2]. E.g., Streams of Diracs, non-uniform splines, and piecewise polynomials. **Present Focus:** Sets of 2-D Diracs, bilevel and planar polygons.

Reconstruction algorithms? Polynomial reproduction, Complex-moments, Annihilating filter, and Directional derivatives.

Sampling kernel and properties? Any kernel $\phi(x,y)$ that is of compact support and can reproduce polynomials of degrees $\gamma=0,1,2\dots,\Gamma-1$ such that



e.g., B-splines and Daubechies scaling functions are valid kernels.

2. SAMPLING OF SIGNALS WITH FINITE RATE OF INNOVATION (FRI) IN 2-D

• SETS OF 2-D DIRACS (LOCAL RECONSTRUCTION)

Consider $g(x,y) = \sum_j \sum_k a_{j,k} \delta_{xy}(x-x_j, y-y_k)$ and $\phi_{xy}(x,y)$ with support $L_x \times L_y$ such that there is at most one Dirac $a_{p,q} \delta_{xy}(x-x_p, y-y_q)$ in an area of size $L_x T_x \times L_y T_y$.

From the **partition of unity** the **amplitude** is determined as:

$$a_{p,q} = \sum_{j=1}^{L_x} \sum_{k=1}^{L_y} S_{j,k}$$

And using polynomial reproduction properties along **x-axis** and **y-axis**, the **coordinate position** (x_p, y_p) is determined as:

$$x_p = \frac{\sum_{j=1}^{L_x} \sum_{k=1}^{L_y} C_{1,j}^x S_{j,k}}{a_{p,q}}$$

$$y_q = \frac{\sum_{j=1}^{L_x} \sum_{k=1}^{L_y} C_{1,k}^y S_{j,k}}{a_{p,q}}$$

• BILEVEL POLYGONS & DIRACS using COMPLEX-MOMENTS (GLOBAL APPROACH)

Moments are used to characterize unspecified objects. [Shohat et al 1943, Elad et al. 2004]. For a convex, bilevel polygon $g(x,y)$ with N corner points, and an analytic function $h(z)=z^n$, $z=x+\sqrt{-1}y$ in closure Ω , the complex-moments of the polygon follow [Milanfar et al.] [3]:

$$\begin{aligned} \sum_{i=1}^N z_i^n &= \iint_{\Omega} g(x,y) h^n(z) dx dy \\ &= \iint_{\Omega} g(x,y) (z^n)^* dx dy \\ &= n(n-1) \iint_{\Omega} g(x,y) z^{(n-2)} dx dy \\ &= n(n-1) z_{n-2}^n \quad (\text{simple complex-moment}) \\ &= z_n^w \quad (\text{weighted complex-moment}) \end{aligned}$$

The z_i can be retrieved from z_n^w using annihilating filter $A(z)$ (Prony's like method) such that $A[i]*z_n^w = 0$.

where ρ_i are complex weights and z_i are corner point coordinates of the bilevel polygon $g(x,y)$, $n = 2, 3, \dots, 2N+1$.

Theorem [Milanfar et al.] [3]: For a given non-degenerate, simply connected, and convex polygon in the complex Cartesian plane, all its N corner points are uniquely determined by its weighted complex-moments z_n^w up to order $2N-1$.

A sampling perspective to above theorem follows...

As the kernel $\phi_{xy}(x,y)$ can reproduce polynomials up to degree $2N-1$ ($\gamma = 0, 1, \dots, 2N-1$), all $2N$ moments r_n^w are determined from weighted sums of the samples $S_{j,k}$. For instance,

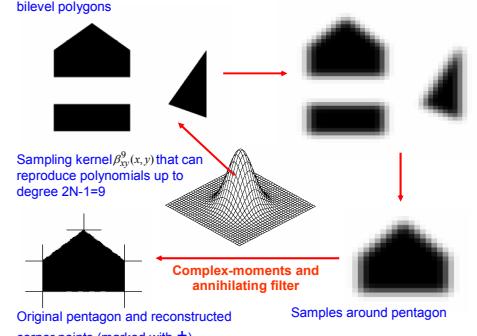
$$r_3^w = 3 \cdot (3-1) \cdot \sum_{j,k} \left(C_{1,j}^x + \sqrt{-1} C_{1,k}^y \right)^{(3-2)} S_{j,k} = \sum_{i=1}^N \rho_i z_i^3$$

Now for both bilevel polygon and set of Diracs, using **complex-moments** and **annihilating filter method**, it is straightforward to see that



Original image with three bilevel polygons

Sampled version of the image

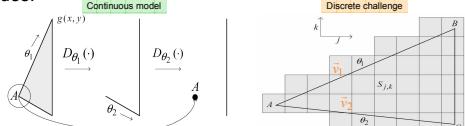


Original pentagon and reconstructed corner points (marked with +)

Samples around pentagon

• PLANAR POLYGONS based on DIRECTIONAL DERIVATIVES (LOCAL APPROACH)

Intuitively, for a planar polygon, two successive directional derivatives along two adjacent sides of the polygon results into a 2-D Dirac at the corner point formed by the respective sides.



Lattice theory [4] and [Convey and Sloan]

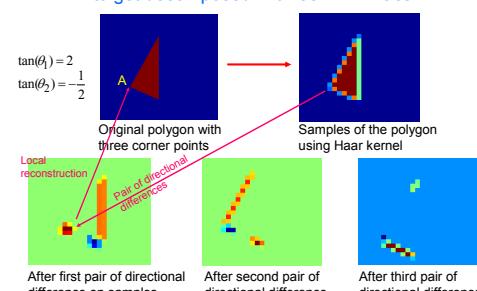
Link directional derivatives \rightarrow discrete differences. Subsampling over integer lattices.

Local directional kernels in the framework of 2-D Dirac sampling (local reconstruction).

$$V_A = \begin{bmatrix} \vec{v}_1 \\ \vec{v}_2 \end{bmatrix} = \begin{bmatrix} v_{1,1} & v_{1,2} \\ v_{2,1} & v_{2,2} \end{bmatrix}$$

$$\theta_1 = \tan^{-1}\left(\frac{v_{1,2}}{v_{1,1}}\right) \quad \theta_2 = \tan^{-1}\left(\frac{v_{2,2}}{v_{2,1}}\right)$$

A planar triangle needs three pairs of directional differences to get decomposed in three 2-D Diracs.



$$D_{\theta_1} [D_{\theta_2} [S_{j,k}]] = \frac{1}{|\det(V_A)|} \left(\frac{\partial}{\partial \theta_2} \left(\frac{\partial}{\partial \theta_1} (g(x,y)) \right) \zeta_{\theta_1 \theta_2}(x,y) \right) \text{ where } \zeta_{\theta_1 \theta_2}(x,y) = \frac{\beta_{\theta_1}^0(x,y) * \phi_{xy}(x,y)}{|\sin(\theta_2 - \theta_1)|}$$

At each corner point \rightarrow Independent modified (directional) kernel $\zeta_{\theta_1 \theta_2}(x,y)$.

$$\phi_{xy}(x,y) * \zeta_{\theta_1 \theta_2}(x,y) \text{ support: } L_x \times L_y$$

The directional kernel $\zeta_{\theta_1 \theta_2}(x,y)$ can reproduce polynomials of degrees 0 and 1 in both x and y directions.

Assume only one corner point in support of directional kernel. Using local reconstruction scheme, the amplitude and the position of Dirac at any corner point (e.g. at point A) follows:

$$a_{p,q} = \frac{\sum_{j,k} D_{\theta_1} [D_{\theta_2} [S_{j,k}]]}{|\det(V_A)|}$$

$$x_p = \frac{\sum_{j,k} C_{1,j}^x D_{\theta_1} [D_{\theta_2} [S_{j,k}]]}{a_{p,q} |\det(V_A)|}$$

$$y_q = \frac{\sum_{j,k} C_{1,k}^y D_{\theta_1} [D_{\theta_2} [S_{j,k}]]}{a_{p,q} |\det(V_A)|}$$

3. CONCLUSION AND ONGOING WORK

Conclusion: Local and global sampling choices for the classes of 2-D FRI signals with varying degrees of complexity.

Current investigations: Sampling of more general shapes such as circles, ellipses, and polygons containing polygonal holes inside.

Future plans: Extension of our 2-D sampling results in higher dimensions and effect of noise. Integration of sampling results with wavelet footprints for image resolution enhancement and super-resolution photogrammetry.

4. KEY REFERENCES

1. M Vetterli, P Marziliano, and T Blu, "Sampling signals with finite rate of innovation," IEEE Trans. on Signal Processing, 50(6): 1417-1428, June 2002.
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3. P Milanfar, G Verghese, W Karl, and A Willsky, "Reconstructing polygons from moments with connections to array processing," IEEE Trans. on Signal Processing, 43(2): 432-443, February 1995.
4. V Velisavljevic, B Beferull-Lozano, M Vetterli, and P L Dragotti, "Discrete multi-directional wavelet bases," Proc. IEEE ICIP, Barcelona, Spain, September 2003.