

# ON THE APPROXIMATION POWER OF SPLINES: ORTHOGONAL VERSUS HEXAGONAL LATTICES

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## Extended abstract

Recently, we have proposed a novel family of bivariate, non-separable splines [1]. These splines, called “hex-splines” have been designed to deal with hexagonally sampled data. Incorporating the shape of the Voronoi cell of a hexagonal lattice, they preserve the twelve-fold symmetry of the hexagon tiling cell. Similar to B-splines, we can use them to provide a link between the discrete and the continuous domain, which is required for many fundamental operations such as interpolation and resampling [2]. The question we answer in this paper is “How well do the hex-splines approximate a given function in the continuous domain?” and more specifically “How do they compare to separable B-splines deployed on a lattice with the same sampling density?”

A general signal space, spanned by shifted versions of a function  $\varphi(\mathbf{x})$  (such as a spline) on a lattice described by a matrix  $\mathbf{R} = [\mathbf{r}_1 \mathbf{r}_2]$ , contains all signals

$$s(\mathbf{x}) = \sum_{\mathbf{k}} c(\mathbf{k}) \varphi(\mathbf{x} - \mathbf{R}\mathbf{k}); \quad c(\mathbf{k}) \in l_2(\mathbb{Z}^2). \quad (1)$$

In general, the coefficients  $c(\mathbf{k})$  are determined as

$$c(\mathbf{k}) = \int g(\mathbf{x}) \tilde{\varphi}(\mathbf{x} - \mathbf{R}\mathbf{k}) d\mathbf{x}, \quad (2)$$

where  $g$  is the original function and  $\tilde{\varphi}$  is the prefilter. The optimal choice, i.e., corresponding to an orthogonal projection into the function space, is the dual filter  $\tilde{\varphi}_d = \hat{\varphi} / \hat{\alpha}_\varphi$ . Here  $\hat{\alpha}_\varphi$  is the Fourier transform of the sampled autocorrelation function of  $\varphi$ . Another common choice is the interpolation prefilter, which selects  $c(\mathbf{k})$  such that  $s(\mathbf{R}\mathbf{k}) = g(\mathbf{R}\mathbf{k})$ .

Separable B-splines are a perfect fit to be used as basis functions on conventional rectangular lattices. For hexagonal lattices, one can use a “slanted” version of the B-splines. Their support correspond to a rhomboid. Recently, we proposed the use of hex-splines, which are inspired on the Voronoi cell indicator function and exhibit a hexagonal support. Higher order hex-splines are constructed by

successive two-dimensional convolutions. First, we want to compare hex-splines versus slanted B-splines on the same hexagonal lattice. Second, we also compare hex-splines on a hexagonal lattice against separable B-splines on a square lattice with the same sampling density. These comparisons are done from an approximation theory point of view.

Approximation theory provides us with a convenient way to quantify the approximation error by integration with an error kernel  $E(\boldsymbol{\omega})$  in the Fourier domain: [3]

$$\|s(\mathbf{x}) - g(\mathbf{x})\|^2 = \frac{1}{4\pi^2} \int |\hat{g}(\boldsymbol{\omega})|^2 E(\boldsymbol{\omega}) d\boldsymbol{\omega}. \quad (3)$$

This error kernel is composed out of two parts:

$$\begin{aligned} E(\boldsymbol{\omega}) &= E_{\min} + E_{\text{res}} \\ &= 1 - \frac{|\hat{\varphi}(\boldsymbol{\omega})|^2}{\hat{\alpha}_\varphi(\boldsymbol{\omega})} + E_{\text{res}}. \end{aligned}$$

Most important, in the case of using the optimal prefilter (orthogonal projection), this kernel reduces to  $E_{\min}$ .

The asymptotic behavior tells us how well the approximation converges to the original  $g$  when the sampling lattice is made finer by a scaling factor  $h$ . In this case, the argument of the error kernel under the integral of Eq. (3) is scaled accordingly as  $E(h\boldsymbol{\omega})$ . So by analyzing  $E(\boldsymbol{\omega})$  around  $\mathbf{0}$  we obtain

$$\begin{aligned} \|s(\mathbf{x}) - g(\mathbf{x})\|^2 &\propto E(h\boldsymbol{\omega}) \\ &\propto h^{2L} O(\|\boldsymbol{\omega}\|^{2L}), \end{aligned} \quad (4)$$

when  $L$  is the order of approximation. The constants in front of  $\omega_1^k \omega_2^{2L-k}$  allow us to compare the behavior of  $E(\boldsymbol{\omega})$  between different signal spaces when the order of approximation is the same.

After a brief introduction on hex-splines and approximation theory, this paper concentrates on orthogonal projection, i.e., using the optimal prefilter  $\tilde{\varphi}_d$ . As mentioned before, this is the best possible way to approach a function in the spline space. First, we compute the asymptotic constants for the hex-splines; i.e., the accurate asymptotic behavior of

(4) when the sampling grid gets denser. Second, we compare the asymptotic constants against the ones we obtain for slanted B-splines (with the same order of approximation  $L$  as the hex-splines) on the same lattice and B-splines (again the same order) on a square lattice with the same sampling density. As result, we find that the asymptotic constants for the hex-splines are smaller than those for the B-splines in both cases. Therefore, hex-splines are asymptotically a better representation than splines on hexagonal lattices. Second, hex-splines on a hexagonal lattice allow to obtain a better approximation for a given function than B-splines on an orthogonal lattice with the same sampling density.

## References

- [1] Dimitri Van De Ville, Rik Van de Walle, Wilfried Philips, and Ignace Lemahieu, “Image resampling between orthogonal and hexagonal lattices,” in *Proceedings of the IEEE International Conference on Image Processing*, Rochester (NY), USA, Sept. 2002, vol. 3, pp. 389–392, IEEE.
- [2] M. Unser, “Sampling—50 years after Shannon,” *Proceedings of the IEEE*, vol. 88, no. 4, pp. 569–587, Apr. 2000.
- [3] Thierry Blu and Michael Unser, “Quantitative Fourier analysis of approximation techniques: Part I—interpolators and projectors,” *IEEE Transactions on Signal Processing*, vol. 47, no. 10, pp. 2783–2795, Oct. 1999.