ON THE CHOICE OF WINDOW FOR SPATIAL SMOOTHING OF SPHERICAL DATA

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ABSTRACT

This paper investigates spectral filtering using isotropic spectral windows, which is a computationally efficient method of spatial smoothing on the sphere. We propose a Slepian eigenfunction window, which is obtained as a solution of the concentration problem on the sphere, as a good choice of the window function. We also unify a comprehensive set of quantitative tools, both spatial and spectral, to assess and compare the performance of different smoothing windows (i.e., smoothers). We analyze and compare the performance of the proposed window against the two best available candidates in the literature: von-Hann window and von Mises-Fisher distribution window. We establish that the latter window includes the popular Gauss window as a subcase. We show that the Slepian eigenfunction window has the smallest spatial variance (better spatial localization) and the smallest side-lobe level.

Index Terms—2-sphere; unit sphere; windows; convolution; smoothing; spherical harmonic transform.

1. INTRODUCTION

Signals are naturally defined on the 2-sphere in a variety of science and engineering disciplines including geodesy [1, 2], cosmology [3], biomedical imaging [4], and wireless channel modeling [5]. Smoothing of such signals in the spatial domain, for noise reduction or for band-limiting the signal, is often carried out as isotropic convolution [1–4, 6–8], which corresponds to the windowing of the spectral domain of the signal in the spectral domain. The choice of the window function is, therefore, crucial in the smoothing process.

1.1. Relation to Prior Work

Many window functions have been proposed for spatial smoothing of signals on the sphere, either in the spatial or the spectral domain [1–4, 7]. Ideally, the window should be (i) a rectangular shape (sharp cut-off) in the spectral domain such that complete information is preserved while band-limiting the signal and (ii) a Dirac delta in the spatial domain to avoid accumulation of the signal at each spatial position from the neighbouring signal components. However, these ideal characteristics in both the spatial and the spectral domains cannot be achieved simultaneously by virtue of the uncertainty principle. Hence, smoothed or tapered spectral windows are often used in practice. For example, the Gauss window has been used in geodesy for isotropic smoothing of gravity data observed over the sphere [1, 2]. The von Mises-Fisher distribution window has been used to perform isotropic convolution for smoothing cosmological observations [3]. Furthermore, Gauss-Weistrass window is used for diffusion-based spatial smoothing over the sphere [7] and for smoothing of cortical surfaces in biomedical imaging [4], where it has been termed as the heat kernel. Recently, spatial and spectral measures to compare the performance of different spectral windows were proposed in [8] and it was shown that the von-Hann window serves as a good choice for spectral windowing. However [8] analysed the window functions popular in geodesy by adapting the measures of performance in [9] and did not consider some other measures [4, 10] and windows [3, 7].

1.2. Contributions

In this work, we propose to use the suitably selected eigenfunction window for isotropic convolution (spectral windowing). The proposed window is obtained as a solution of Slepian concentration problem on the sphere [11–13]. We also unify the different figures of merit in the literature to measure and analyse the performance of different smoothing windows (i.e., smoothers). We analyse the von-Hann, von Mises-Fisher distribution (Gauss) and eigenfunction windows and show that the proposed eigenfunction window serves as a better choice in terms of a number of spatial and spectral performance measures.

The paper is organized as follows. The mathematical background and window functions are reviewed in Section 2. The problem statement is defined in Section 3. The measures for quantifying the performance of the smoothers are presented in Section 4. The results are discussed in Section 5.

2. PRELIMINARIES

2.1. Signals and Systems on the 2-Sphere

We consider the complex Hilbert space finite energy functions on the 2-sphere, $L^2(S^2)$, which is equipped with the following inner product

$$\langle f, h \rangle \triangleq \int_{S^2} f(\hat{x}) \overline{h(\hat{x})} d\hat{x},$$

(1)

where $\hat{x} \equiv \hat{x}(\theta, \phi) \triangleq (\sin \theta \cos \phi, \sin \theta \sin \phi, \cos \theta) \in \mathbb{R}^3$ is a unit vector which parameterizes a point on the 2-sphere, $S^2$, $d\hat{x} = \sin \theta d\theta d\phi$, and $(\cdot)^\dagger$ denotes complex conjugate. $\theta \in [0, \pi]$ denotes the co-latitude measured with respect to the positive z-axis and $\phi \in [0, 2\pi]$ denotes the longitude measured with respect to the positive $x$-axis in the $x$-$y$ plane. The inner product in (1) induces a norm $\|f\| \triangleq \langle f, f \rangle^{1/2}$, and the functions with finite induced norm are referred as signals on the 2-sphere in this paper.

Spherical harmonics [14] defined for integer degree $\ell \geq 0$ and integer order $m \in [-\ell, \ell]$ form archetype complete orthonormal set of basis functions for $L^2(S^2)$ and are expressed as follows

$$Y^m_\ell(\theta, \phi) \triangleq \sqrt{\frac{2\ell + 1}{4\pi} \frac{(\ell - m)!}{(\ell + m)!}} P^m_\ell(\cos \theta)e^{im\phi} \equiv Y^m_\ell(\hat{x}),$$

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that rotates an azimuthally symmetric function $h$ about an axis followed by $(\hat{h} \cdot \hat{g}) = h(\arccos \hat{x} \cdot \hat{y})$.}

which has unit energy normalization

\[ h(\theta; \theta_{\text{FWHM}}) = b e^{-b(1 - \cos \theta_c)}, \quad b = \frac{\ln 2}{1 - \cos \theta_{\text{FWHM}}} \]  

(8)

where the parameter $\theta_{\text{FWHM}} > 0$ indicates the degree of smoothness.

\textbf{von Hann Window (Spatial):} The von Hann window, commonly known as Hann window or raised cosine window in Euclidean domain [9], is defined in terms of the width of the main lobe $\theta_c$ as follows:

\[ h(\theta; \theta_c) = \begin{cases} \frac{1}{2} \left(1 + \cos (\theta/\theta_c) \right) & 0 \leq \theta \leq \theta_c \\ 0 \quad & \text{otherwise.} \end{cases} \]  

(9)

\textbf{von Mises-Fisher Distribution Window (Spatial and Spectral):} The von Mises-Fisher distribution window (or von Mises-Fisher window) has been used in literature for isotropic smoothing in cosmology [3] and is defined for the concentration parameter $\kappa \geq 0$ with known spectral domain representation [19]

\[ h(\theta; \kappa) \triangleq \kappa \exp(\kappa \cos \theta), \quad H_\ell(\kappa) = \frac{I_{\ell + 1/2}(\kappa)}{I_{1/2}(\kappa)}, \]  

(10)

where $I_{\ell + 1/2}(\kappa)$ denotes half-integer-order modified Bessel function of the first kind. We observe that, in fact, the von Mises-Fisher distribution window and the Gaussian window are identical when $\kappa = b$.

\textbf{Gauss-Weierstrass Window (Spectral):} This window has the following spectral characterization for $\kappa \geq 0$

\[ H_\ell(\kappa) = e^{-\ell(\ell+1)/(12\kappa)}, \]  

(11)

which has been used for spherical diffusion [7] and also for spectral smoothing of cortical surfaces [4]. We note that the Gauss-Weierstrass window is an asymptotic form of the von Mises-Fisher distribution window for large values of the parameter $\kappa$ [3].

2.4. Slepian Eigenfunction Window

With two parameters, the band-limit $L$ and angular size $\theta_c$, one can solve a Slepian concentration problem on the sphere, which finds the band-limited function that maximizes the proportion of energy that falls within the polar cap region $\theta \in [0, \theta_c]$, see [11-13]. This function is the eigenfunction with the largest eigenvalue of an associated integral equation. Generally the concentration is high (close to unity) unless both $L$ and $\theta_c$ are simultaneously too small. The specific “Slepian eigenfunction window” we refer to in this section is one where both $L$ and $\theta_c$ are simultaneous as small as possible in the following sense:

\[ \theta_c = \frac{2\pi}{L+1}, \]  

(12)

for which it can be established that one eigenfunction is at least 99% concentrated [12, 20]. Therefore, using (12) we have a one parameter “eigenfunction window” family parametrized (without loss of generality) by $L$. The eigenfunction window $h \in H^0$ with band-limit $L$ and energy concentration within cap angle $\theta_c$, given in (12), can be computed in spectral domain as a solution of following algebraic eigenvalue problem [12, 20]

\[ D h = \lambda h, \]  

(13)

where $h = [h^0_m, h^1_m, \ldots, h^L_m]$ and $D$ is the $(L+1) \times (L+1)$ real and symmetric matrix with entries given by

\[ D_{\ell,\ell'} = 2\pi \int_0^{\pi} Y^\ell_m(0,0) Y^\ell_{m'}(0,0) \sin \theta d\theta, \]  

(14)
which can be computed analytically [12], however, requires the computation of Wigner-3j symbols [21]. Alternatively, the eigenvalue problem in (13) can be solved as an eigen decomposition of the matrix $G$ of size $(L + 1) \times (L + 1)$ that commutes with $D$, that is, $DG = GD$. It has been shown that $G$ is a tridiagonal matrix with entries of the form [13]

$$G_{\ell,\ell+1} = G_{\ell+1,\ell} = \frac{(\ell + 1)(\ell(\ell + 2) - (L)(L + 2))}{\sqrt{(2\ell + 1)(2\ell + 3)}},$$

$$G_{\ell,\ell} = -\ell(\ell + 1) \cos(\theta_c), \quad G_{\ell,\ell'} = 0. \quad \text{(15)}$$

Since $G$ is a tridiagonal matrix with simple entries given in (15), its eigen decomposition can be easily obtained and gives $L + 1$ eigenvectors of the form $h$ and the one with the smallest eigenvalue yields the spectral domain representation of the desired eigenfunction window. This procedure furnishes $(h)^2$ which are readily scaled to give the $H_l$ shown in (6) and Fig. 1. Neither the spatial form, (5), nor the spectral form, (6), of a smoothing operator are closed-form but can be easily obtained by the eigen decomposition of $G$.

3. PROBLEM STATEMENT

Isotropic convolution or spectral smoothing are used to reduce the effects of noise [2], to perform basic low-pass filtering such as forming a band-limited representation of signal and to perform spatial smoothing. For example, rectangular window (7) can be used to band-limit a signal at degree $L$. However, the choice of such a rectangular window produces ringing artifacts (due to the Gibb’s phenomenon [22]) in the spatial domain as noted in [4]. This highlights the need to find objective measure to assess performance and compare different designs for smoothers.

In summary our problem addresses the following issues:

1. to use and develop quantitative tools [8] to quantify smoother performance on the 2-sphere;
2. to implement and assess the performance of the eigenfunction window as a smoother; and
3. to compare the eigenfunction window with smoothers from the literature, [1–4, 7, 8], and Section 2.3.

4. MEASURE FOR QUANTIFYING PERFORMANCE

We categorize measures for quantifying performance into spectral and spatial measures depending on the domain under study. In the derivation of these measures, we assume that the window function under analysis unit energy normalized, that is,

$$2\pi\int_{\theta=0}^{\pi} |h(\theta)|^2 \sin \theta d\theta = 1 \iff \sum_{\ell=0}^{\infty} |h_{\ell}|^2 = \sum_{\ell=0}^{\infty} \frac{2\ell + 1}{4\pi} |H_{\ell}|^2 = 1.$$

4.1. Spectral Measures

**Spectral Variance:** Spectral variance quantifies the spread of the spectral window in the spectral domain. Mathematically, the variance of the window function $h \in \mathcal{H}^0$ in the spectral (Fourier) domain, denoted by $\sigma_F$, is defined as [10]

$$\sigma_F^2 = 4\pi \sum_{\ell=0}^{L} \frac{\ell(\ell + 1)}{2\ell + 1} |H_{\ell}|^2. \quad \text{(16)}$$

Since the windowing is applied in spectral domain, a superior window should have large spectral variance for a given band-limit $L$ of the window function.

![Fig. 2](image-url)

**Processing Loss:** Due to the smoothness imposed by the spectral window, there is information loss in the spectral domain. We quantify such processing loss for a window with band-limit $L$ as [9]

$$\beta = \left( \sum_{\ell=0}^{L} |H_{\ell}|^2 \right)^2. \quad \text{(17)}$$

The higher the value of $\beta$ for a given window, the lower the processing loss due to smoothness of the spectral window.

4.2. Spatial Measures

Since the spectral windowing of the signal with $H_l$ is a consequence of convolution of the signal with a signal $h(\theta)$, we identify the major characteristics of the window function in the spatial domain that will allow us to compare the performance of different windows.

**Spatial Variance:** The spatial variance serves as a measure of spatial localization or peakiness of the window function around its center (usually taken as the north pole). Mathematically define spatial variance $\sigma_S^2$ as [10, 20]

$$\sigma_S^2 = 1 - \left( \pi \int_{\theta=0}^{\pi} \sin(2\theta) |h(\theta)|^2 d\theta \right)^2. \quad \text{(18)}$$

The smaller value of spatial variance indicates more localization of the window function in the spatial domain, which implies the better performance as this minimizes the accumulation of different signal components.

**Full-Width at the Half-Maximum (FWHM):** Let FWHM be denoted by $\theta_{\text{FWHM}}$ and defined as the width from the center (usually $\theta = 0$) where the function $h(\theta)$ attains the maximum value, that is, $h(\theta_{\text{FWHM}}) = \frac{1}{2} h(0)$.
It is a measure of localization of window function and is sometimes referred as 3dB width of the window function [4]. If the separation between the two same strength signal components in the spatial domain is less than twice of FWHM, the two components appear as a single component and will not be resolved in the smoothed signal.

**Main-Lobe Energy and Energy Leakage**: Main-lobe energy, denoted by \( E_m \), and energy leakage, denoted by \( E_s \), are defined as the energy in the main lobe parameterized by \( \theta_c \) and the energy outside the main lobe respectively, that is,

\[
E_m = 2\pi \int_0^{\theta_c} |h(\theta)|^2 \sin \theta \, d\theta, \quad E_s = 1 - E_m. \tag{19}
\]

The larger value of \( E_m \) indicates better performance.

**Side-lobe Level**: Side-lobe level is defined as the magnitude of the highest side-lobe. This is an important measure as it quantifies the accumulation of the largest unwanted contribution outside the mainlobe. We use \( \alpha \) to denote the magnitude of the highest side-lobe.

**Analysis Results**: For \( L = 14 \), the different windows are plotted in Fig. 2(a) and (b) in spatial domain as \( h(\theta) \) and spectral domain as \( H_\ell \) respectively. It is evident that the eigenfunction window has better spatial localization and smaller side-lobe level.

We compare the processing loss (\( \beta \)), spatial variance (\( \sigma_S \)) and side-lobe level (\( \alpha \)) for windows with band-limits in the range \( 10 \leq L \leq 22 \) in Fig. 3(a), (b) and (c) respectively. von Mises-Fisher distribution window has the higher (better) processing loss. However, the processing loss does not vary with the band-limit. The eigenfunction window has the smallest spatial variance (better spatial localization) and the smallest side-lobe level.

In order to further study the trade-off between spatial variance and side-lobe level, we plot the spatial variance versus side-lobe level in Fig. 4 for different windows, where it can be observed that eigenfunction window has better performance as compared to the von-Hann window and von Mises-Fisher distribution window.

**5. PERFORMANCE COMPARISON**

**Fair Comparison Strategy**: Here we use the measures defined in Section 4 to compare the performance of the von Mises-Fisher, von-Hann, rectangular and the eigenfunction windows (as the von Mises-Fisher distribution and Gauss window are equivalent and the Gauss-Weierstrass asymptotically equivalent). Since the windows are defined by different parameters, we devise the following strategy in order to carry out meaningful comparison: 1) for given \( L \), determine \( \theta_c \) using (12) and determine the eigenfunction window; and 2) choose parameters of von Mises-Fisher distribution window and von-Hann window such that each window function, when band-limited at \( L \), is unit energy normalized and has the spectral variance \( \sigma_F \) equal to that of eigenfunction window.

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**6. CONCLUSIONS**

In this work, we have categorised the different figures of merit to compare the performance of different spectral windows for spatial smoothing. Based on the comparison with other windows: von-Hann and von Mises-Fisher (Gauss), we have proposed the use of eigenfunction window, obtained as a solution of Slepian concentration problem on the sphere. We have shown that the eigenfunction window is more localized in spatial domain and also exhibits the smaller side-lobe level and therefore reduces the accumulation of different signal components in the spatial domain.
7. REFERENCES


