# Robust Slepian Functions on the Sphere

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Abstract—Slepian functions on the sphere maximally concentrate energy inside a region for a given bandlimit.

Index Terms-key, word, keyword list

#### I. Introduction

#### A. Background

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#### B. Contributions

- We use a weighting function,  $h(\hat{x})$  to generalise the standard Slepian formulation, section II-B.
- The Slepian probem is known to yield dual-orthogonality; when it is generalised through the weighting function then this property has a generalization and further there is a three-fold orthogonality, section II-D. Further, the derivation for orthogonality is succinct and general using the notion of isomorphism.
- Regularization for the case of measurement error or noise yields robustness to noise enhancement that would occur for eigenfunctions with small eigenvalues, section III-B.

# C. Paper Organization

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### II. PROBLEM FORMULATION

#### A. Notation

The natural Hilbert space on the sphere is denoted  $L^2(\mathbb{S}^2)$  with inner product  $\langle f,g\rangle\triangleq\int_{\mathbb{S}^2}f(\widehat{\boldsymbol{x}})\,\overline{g(\widehat{\boldsymbol{x}})}\,ds(\widehat{\boldsymbol{x}})$ . The spherical harmonic transform (SHT) is given by

$$(f)_{\ell}^{m} = \langle f, Y_{\ell}^{m} \rangle = \int_{\mathbb{S}^{2}} f(\widehat{x}) \, \overline{Y_{\ell}^{m}(\widehat{x})} \, ds(\widehat{x})$$

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for degree  $\ell \in \{0, 1, \ldots\}$  and order m where  $|m| \leq \ell$ .

The subspace of band-limited functions on the sphere of maximum degree L is denoted  $\mathcal{H}_L \in L^2(\mathbb{S}^2)$  and is N-dimensional where

$$N \triangleq (L+1)^2$$

If a signal  $f(\widehat{x})$  is band-limited to L then  $\langle f, Y_{\ell}^m \rangle = 0$  for  $\ell > L$  and it has the spectral (spherical harmonic) representation given by the vector

$$\mathbf{f} = \left[ (f)_0^0, (f)_1^{-1}, (f)_1^0, (f)_1^1, \dots, (f)_L^L \right] \in \mathbb{C}^N.$$
 (1)

This vector can be indexed with  $n=0,1,2,\ldots,N-1$ , where  $n=\ell(\ell+1)+m$ . Generally when we say f is band-limited then the maximum degree L is understand.

The spatial and spectral representations are related through isomorphism [1]

$$\langle f, g \rangle = \langle \mathbf{f}, \mathbf{g} \rangle_{\mathbb{C}^N}, \quad \forall f, g$$
 (2)

where the spectral-domain inner product is  $\langle \mathbf{f}, \mathbf{g} \rangle_{\mathbb{C}^N} = \mathbf{g}^H \mathbf{f}$ . This isomorphism greatly simplifies our demonstrations of different types of orthogonality.

## B. Weighted spatial-domain concentration problem

Let  $h(\widehat{x})$  be a real, non-negative weighting function bounded by unity on the unit sphere  $\mathbb{S}^2$ :

$$0 < h(\widehat{\boldsymbol{x}}) < 1, \quad \forall \widehat{\boldsymbol{x}} \in \mathbb{S}^2.$$
 (3)

Then we seek the band-limited signal  $f(\hat{x}) \in \mathcal{H}_L$  that maximizes the following weighted spatial-domain concentration

$$\lambda_0 = \max_{f \in \mathcal{H}_L} \left\{ \frac{\int_{\mathbb{S}^2} h(\widehat{x}) \left| f(\widehat{x}) \right|^2 ds(\widehat{x})}{\int_{\mathbb{S}^2} \left| f(\widehat{x}) \right|^2 ds(\widehat{x})} \right\}. \tag{4}$$

1) Concentration interpretation: By specifying the signal to be band-limited then we are limiting the spread of energy in the spectral domain. By judicious choose of the weighting function,  $h(\widehat{x})$  in (4), we attempt to limit the spatial-domain distribution of energy to some localized portion of the sphere. The uncertainty principle says we cannot simultaneously concentrate energy in the spatial and spectral domains. A concentration problem is when we attempt to concentrate as much as possible in both domains.

## C. Spectral-domain reformulation of concentration problem

The weighted concentration problem (4) can be written in the spectral domain as the Rayleigh quotient

$$\lambda_0 = \max_{\mathbf{f} \in \mathbb{C}^H} \frac{\mathbf{f}^H \mathbf{H} \mathbf{f}}{\mathbf{f}^H \mathbf{f}},\tag{5}$$

where the spectral-domain Hermitian matrix H has elements

$$H_{\ell,p}^{m,q} \triangleq \int_{\mathbb{S}^2} h(\widehat{\boldsymbol{x}}) Y_p^q(\widehat{\boldsymbol{x}}) \overline{Y_\ell^m(\widehat{\boldsymbol{x}})} ds(\widehat{\boldsymbol{x}}), \tag{6}$$

and f is the spectral representation of  $f(\hat{x})$ . Note that the rows and columns of matrix H are indexed consistent with f in (1).

The Rayleigh quotient (5) is solved by finding the eigenvector,  $\mathbf{v}_0$ , corresponding to the largest eigenvalue,  $\lambda_0$ , of  $\mathbf{H}$ . All eigenvalues of  $\mathbf{H}$  are real and non-negative,  $\lambda_0 \geq \lambda_1 \geq \lambda_2 \geq \cdots \geq 0$ , and the corresponding eigenvectors,  $\mathbf{v}_0, \mathbf{v}_1, \mathbf{v}_2, \ldots$ , can be chosen as orthonormal, that is,

$$\mathbf{H}\mathbf{v}_n = \lambda_n \mathbf{v}_n, \quad \text{with} \quad \langle \mathbf{v}_n, \mathbf{v}_m \rangle_{\mathbb{C}^N} = \delta_{m,n}.$$

If the components of the dominant spectral-domain eigenvector,  $\mathbf{v}_0$ , are  $(v_0)_{\ell}^m$  then the spatial-domain eigenfunction is

$$v_{0}(\widehat{\boldsymbol{x}}) = \sum_{\ell,m} (v_{0})_{\ell}^{m} Y_{\ell}^{m}(\widehat{\boldsymbol{x}})$$

$$= \arg \max_{f \in \mathcal{H}_{L}} \left\{ \frac{\int_{\mathbb{S}^{2}} h(\widehat{\boldsymbol{x}}) \left| f(\widehat{\boldsymbol{x}}) \right|^{2} ds(\widehat{\boldsymbol{x}})}{\int_{\mathbb{S}^{2}} \left| f(\widehat{\boldsymbol{x}}) \right|^{2} ds(\widehat{\boldsymbol{x}})} \right\}. \tag{7}$$

and this is the optimal spatial function that attains (4).

# D. Three-fold spatial-domain orthogonality of eigenfunctions

Firstly, the N eigenvectors  $\mathbf{v}_n$ ,  $n = 0, 1, 2, \dots, N-1$ , of  $\mathbf{H}$  are orthonormal in  $\mathbb{C}^N$ . Then by isomorphism, (2), we have *orthonormality* of the N eigenfunctions  $v_n(\widehat{x})$  in  $\mathcal{H}_L$ :

$$\langle v_n, v_m \rangle = \langle \mathbf{v}_n, \mathbf{v}_m \rangle_{\mathbb{C}^N} = \delta_{m,n}.$$

Secondly, since we have eigenvectors then

$$\langle \mathbf{H} \mathbf{v}_n, \mathbf{v}_m \rangle_{\mathbb{C}^N} = \mathbf{v}_m^H \mathbf{H} \mathbf{v}_n = \mathbf{v}_m^H \lambda_n \mathbf{v}_n$$
  
=  $\lambda_n \langle \mathbf{v}_n, \mathbf{v}_m \rangle_{\mathbb{C}^N} = \lambda_n \delta_{n.m.}$ 

So, by isomorphism, spatially this is the same as

$$\int_{\mathbb{S}^2} h(\widehat{\boldsymbol{x}}) \, v_n(\widehat{\boldsymbol{x}}) \, \overline{v_m(\widehat{\boldsymbol{x}})} \, ds(\widehat{\boldsymbol{x}}) = \lambda_n \, \delta_{n,m}.$$

This is a spatial-domain orthogonality of the  $v_n(\widehat{x})$  with respect to a weighted spatial-domain inner product

$$\langle f, g \rangle_h \triangleq \int_{\mathbb{S}^2} h(\widehat{\boldsymbol{x}}) \, f(\widehat{\boldsymbol{x}}) \, \overline{g(\widehat{\boldsymbol{x}})} \, ds(\widehat{\boldsymbol{x}}).$$

Under this weighted inner product  $\|v_n\|_h^2 = \lambda_n < 1$ . However, it is clear that, whenever  $\lambda_n > 0$ ,  $\{v_n(\widehat{x})/\sqrt{\lambda_n}\}$  are *orthonormal* in the weighted inner product space.

Finally, there is a third sense in which the  $v_n(\widehat{x})$  are orthogonal. Implicitly define a third inner product through

$$\langle f, g \rangle = \langle f, g \rangle_h + \langle f, g \rangle_{1-h}.$$

Then, given (3), we have  $0 \le 1 - h(\hat{x}) \le 1$  and  $v_n(\hat{x})$  are *orthogonal* with the complementary weighted inner product

$$\langle f, g \rangle_{1-h} \triangleq \int_{\mathbb{S}^2} (1 - h(\widehat{\boldsymbol{x}})) f(\widehat{\boldsymbol{x}}) \overline{g(\widehat{\boldsymbol{x}})} ds(\widehat{\boldsymbol{x}}),$$

and can be normalized as in  $\{v_n(\widehat{x})/\sqrt{1-\lambda_n}\}\ (\lambda_n < 1)$ . Further, in this case, the Hermitian matrix is  $\mathbf{H}^c \triangleq \mathbf{I} - \mathbf{H}$ .

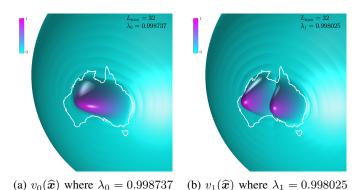


Fig. 1: Magnitudes of the two dominant eigenfunctions for Australia including Tasmania region for band-limit L=32.

In summary, the eigenfunctions satisfy the three-fold *spatial-domain orthogonality* (with spectral counterparts):

$$\langle v_n, v_m \rangle = \delta_{m,n} \qquad (= \langle \mathbf{v}_n, \mathbf{v}_m \rangle_{\mathbb{C}^N})$$

$$\langle v_n, v_m \rangle_h = \lambda_n \, \delta_{m,n} \qquad (= \langle \mathbf{H} \mathbf{v}_n, \mathbf{v}_m \rangle_{\mathbb{C}^N})$$

$$\langle v_n, v_m \rangle_{1-h} = (1 - \lambda_n) \, \delta_{m,n} \qquad (= \langle \mathbf{H}^c \mathbf{v}_n, \mathbf{v}_m \rangle_{\mathbb{C}^N})$$

which implies the energy concentrations are  $\|v_n\|^2=1$ ,  $\|v_n\|_h^2=\lambda_n$ , and  $\|v_n\|_{1-h}^2=1-\lambda_n$ .

# E. Slepian spatial-domain concentration

Defining a region  $R \in \mathbb{S}^2$ , then selecting the real, non-negative weighting function, as  $h(\widehat{x}) = \chi_R(\widehat{x})$ , where

$$\chi_R(\widehat{\boldsymbol{x}}) \triangleq \begin{cases} 1 & \widehat{\boldsymbol{x}} \in R \\ 0 & \text{otherwise} \end{cases}$$
(8)

in (4), and  $\langle f,g\rangle_R=\int_R f(\widehat{\boldsymbol{x}})\,\overline{g(\widehat{\boldsymbol{x}})}\,ds(\widehat{\boldsymbol{x}})$ , is the inner product, we recover the Slepian concentration problem on the sphere

$$\lambda_0 = \max_{f \in \mathcal{H}_L} \frac{\int_R |f(\widehat{\boldsymbol{x}})|^2 ds(\widehat{\boldsymbol{x}})}{\int_{\mathbb{R}^2} |f(\widehat{\boldsymbol{x}})|^2 ds(\widehat{\boldsymbol{x}})},\tag{9}$$

whose spectral-domain Hermitian matrix H has elements

$$H_{\ell,p}^{m,q} \triangleq \int_{\mathcal{P}} Y_p^q(\widehat{x}) \, \overline{Y_\ell^m(\widehat{x})} \, ds(\widehat{x}). \tag{10}$$

We have the three-fold spatial-domain orthogonality of Slepian eigenfunctions: on the whole sphere, within region R and within region  $\mathbb{S}^2 \setminus R$ .

For illustration, on the Earth, normalized with unit radius, let the region  $R \in \mathbb{S}^2$  be the Australian continent including Tasmania, and let band-limit L=32. Therefore N=1089 and the resulting spectral-domain Hermitian matrix is  $\mathbf{H}$  is  $1089 \times 1089$ . The two dominant eigenfunctions  $v_0(\widehat{\boldsymbol{x}})$  and  $v_1(\widehat{\boldsymbol{x}})$  are shown in Fig. 1.

#### III. TOWARDS ROBUST MODELLING IN A REGION

# A. Expansion modelling without noise

Using the orthonormal sequence  $\{v_n(\widehat{x})\}$  in  $\mathcal{H}_L$ , any bandlimited function  $f(\widehat{x})$  has expansion, valid for  $\widehat{x} \in \mathbb{S}^2$ ,

$$f(\widehat{\boldsymbol{x}}) = \sum_{n=0}^{N-1} (f)_n v_n(\widehat{\boldsymbol{x}}) = \sum_{n=0}^{N-1} \underbrace{\sqrt{\lambda_n} (f)_n}_{\triangleq (f)_{him}} \frac{v_n(\widehat{\boldsymbol{x}})}{\sqrt{\lambda_n}}, \quad (11)$$

where

$$(f)_n \triangleq \langle f, v_n \rangle = \int_{\mathbb{S}^2} f(\widehat{\boldsymbol{y}}) \, \overline{v_n(\widehat{\boldsymbol{y}})} \, ds(\widehat{\boldsymbol{y}}),$$

and

$$(f)_{h;n} \triangleq \left\langle f, \frac{v_n}{\sqrt{\lambda_n}} \right\rangle_h = \int_{\mathbb{S}^2} h(\widehat{\boldsymbol{y}}) \, f(\widehat{\boldsymbol{y}}) \, \frac{\overline{v_n(\widehat{\boldsymbol{y}})}}{\sqrt{\lambda_n}} \, ds(\widehat{\boldsymbol{y}}).$$

Therefore, if band-limited  $f(\widehat{x})$  is determined from local information implicit in weighting  $h(\widehat{y})$  we can determine the coefficients of interest using

$$(f)_n = \frac{1}{\lambda_n} \langle f, v_n \rangle_h = \frac{1}{\lambda_n} \int_{\mathbb{S}^2} h(\widehat{\boldsymbol{y}}) f(\widehat{\boldsymbol{y}}) \, \overline{v_n(\widehat{\boldsymbol{y}})} \, ds(\widehat{\boldsymbol{y}}). \tag{12}$$

For example with  $h(\widehat{x}) = \chi_R(\widehat{x})$  then the local information is just the information within region R.

### B. Expansion modelling with uncertainty in measurement

The energy associated with the nth eigenfunction with respect to the weighted localized inner product is  $|(g)_{h;n}|^2$ . However, this implies the energy on the sphere is  $|(g)_{h;n}|^2/\lambda_n$ . Therefore, we may see a significant growth in the energy on the sphere or significant noise enhancement whenever  $\lambda_n$  is small. We can follow the approach taken in [2] to ameliorate such noise enhancement that can occur in (14).

Suppose f is band-limited,  $f \in \mathcal{H}_L$ , but can only be observed in some localized portion of the sphere through a weighting function h and is also subject to noise

$$g(\widehat{\boldsymbol{x}}) = h(\widehat{\boldsymbol{x}}) f(\widehat{\boldsymbol{x}}) + z(\widehat{\boldsymbol{x}}),$$

The regularization objective is

$$\min_{\tilde{f} \in \mathcal{H}_L} \|\tilde{f}\|^2 \quad \text{subject to} \quad \|\tilde{f} - g\|_h^2 \le \epsilon^2$$

Write

$$\widetilde{f}(\widehat{\boldsymbol{x}}) = \sum_{n=0}^{N-1} a_n v_n(\widehat{\boldsymbol{x}}) \quad \text{and} \quad g(\widehat{\boldsymbol{x}}) = \sum_{n=0}^{N-1} b_n v_n(\widehat{\boldsymbol{x}})$$

Then

$$\|\tilde{f}\|^2 = \sum_{n=0}^{N-1} |a_n|^2$$

and

$$\|\tilde{f} - g\|_h^2 = \sum_{n=0}^{N-1} \lambda_n |a_n - b_n|^2$$

#### IV. IDEAS

- Random matrix theory: The square root of the Hermitian is the weighted local projection. Then we can use random matrix theory to characterize the eigenvalues.
- Operator Reformulation: Integral equation and kernel version of theory.
- Regularized Hermitian: Can the Hermitian be regularized directly to deliver more sensible eigenvectors? (The original idea)
- Spatial-limited problem: Do the converse Slepian problem.
- Uncertain L: Need spectral weighting to deal with uncertain L; simultaneous with spatial weighting? Make robust to imprecise band-limited value of L.
- Franks framework reformulation: Cast everything in Franks framework.
- Errors-in-variables: Uncertainty in spatial measurement (domain) and values (range).
- **RKHS:** Real positive definite weighting is 75% there. For weighted inner product is it amenable to kernel trick?
- **Beampattern Deconvolution:** Weighting is a single shot convolution. Does this show how to deconvolve imperfect measurement beamshapes?
- **Sampled space:** Pseudo-DH samples and working with information preserving spatial samples.

# V. CONCLUSIONS

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## REFERENCES

- [1] R. A. Kennedy and P. Sadeghi, *Hilbert Space Methods in Signal Processing*. Cambridge, UK: Cambridge University Press, Mar. 2013.
- [2] Y. Alem, Z. Khalid, and R. A. Kennedy, "Band-limited extrapolation on the sphere for signal reconstruction in the presence of noise," in *Proc. IEEE Int. Conf. Acoustics, Speech and Signal Processing, ICASSP'2014*, Florence, Italy, May 2014, pp. 4141–4145.

# APPENDIX A OBSCURE THING

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# APPENDIX B UNCLEAR THING

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