Spherical Multitaper Analysis via Spatio-Spectrally Concentrated Slepian Functions: Theory and Applications

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Localization in a nutshell

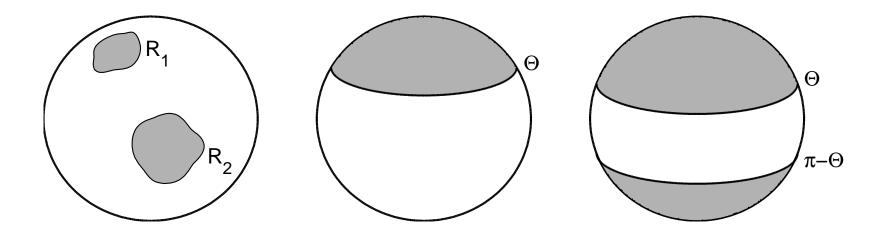
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We can use these "Slepian" functions as **windows**, for spectral analysis, or we can use them as a **(sparse) basis** to represent geophysical observables—on a sphere.



In the 60s Slepian et al. solved the problem of concentrating a bandlimited signal

$$g(t) = \frac{1}{2\pi} \int_{-W}^{+W} G(\omega) e^{i\omega t} d\omega, \qquad |W| < \infty, \tag{1}$$

into a time interval $|t| \leq T$. The "Slepian functions" optimize the concentration

$$\lambda = \frac{\int_{-T}^{+T} g^2(t) dt}{\int_{-\infty}^{+\infty} g^2(t) dt}, \qquad 0 < \lambda < 1.$$
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(2)

They are eigenfunctions of a Fredholm integral equation,

$$\int_{-T}^{T} \left[\frac{\sin W(t - t')}{\pi(t - t')} \right] g(t') dt' = \lambda g(t).$$
(3)

Similarly, two-dimensional Slepian functions are bandlimited Fourier expansions

$$g(\mathbf{x}) = \frac{1}{(2\pi)^2} \int_{\mathcal{K}} G(\mathbf{k}) \, e^{i\mathbf{k}\cdot\mathbf{x}} \, d\mathbf{k}, \qquad |\mathcal{K}| < \infty, \tag{4}$$

that concentrate into a finite **spatial** region $\mathcal{R} \in \mathbb{R}^2$ of area A by maximizing

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These are also **eigenfunctions** of a Fredholm integral equation,

$$\int_{\mathcal{R}} \left[\frac{1}{(2\pi)^2} \int_{\mathcal{K}} e^{i\mathbf{k} \cdot (\mathbf{x} - \mathbf{x}')} d\mathbf{k} \right] g(\mathbf{x}') d\mathbf{x}' = \lambda g(\mathbf{x}).$$
(6)

On a **sphere**, Slepian functions are bandlimited *spherical-harmonic* expansions

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(7)

that are concentrated within a region $R \in \Omega$ by optimizing the energy ratio

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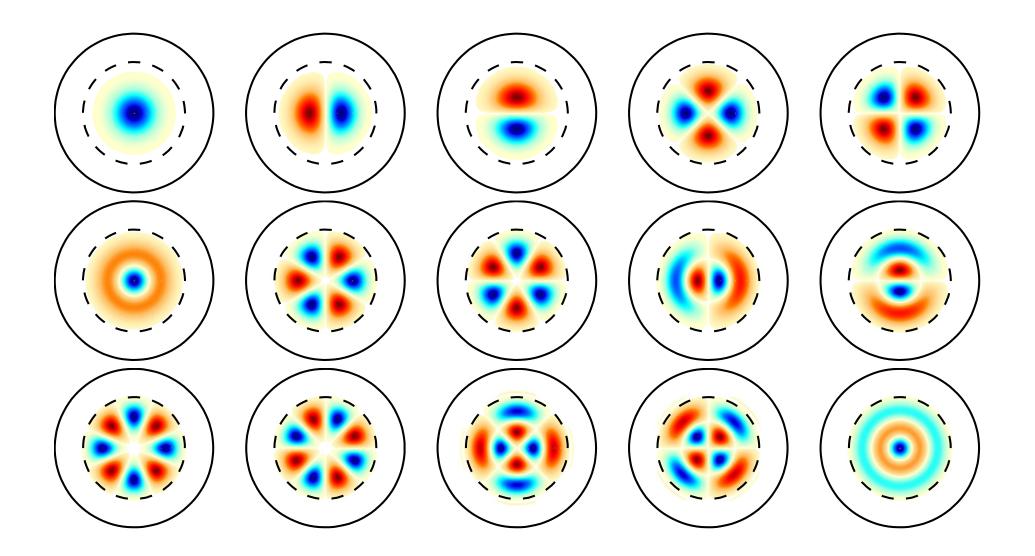
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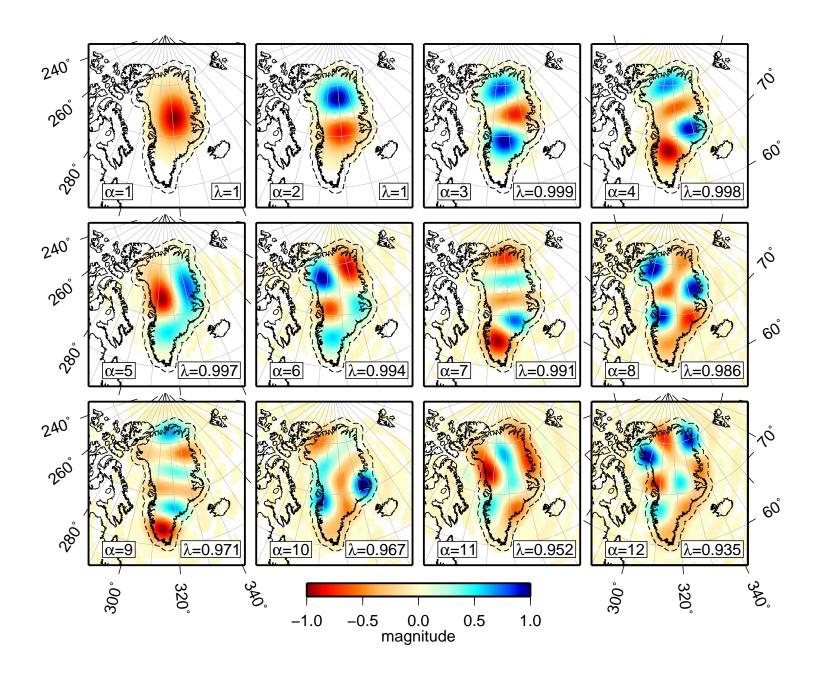
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They are **eigenfunctions** of a Fredholm equation, with P_l a Legendre function,

$$\int_{R} \left[\sum_{l=0}^{L} \left(\frac{2l+1}{4\pi} \right) P_{l}(\hat{\mathbf{r}} \cdot \hat{\mathbf{r}}') \right] g(\hat{\mathbf{r}}') \, d\Omega' = \lambda g(\hat{\mathbf{r}}). \tag{9}$$

Some examples of Slepian functions — 1





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Thus, the Slepian functions are **bases** for **bandlimited** geophysical processes **anywhere** (not just on the domain for which they were constructed, though, there, they will be a **sparse** basis). Their trace is a space-bandwidth joint "Shannon" *area*.

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Remember that the *trace* of an operator is the *sum* of all of its eigenvalues, N.

In the *spectral* domain, the Slepian functions are eigenfunctions of equations that have *spacelimited spectral delta functions* as kernels. On the **sphere**, we solve for the spherical harmonic expansion coefficients of the functions as

$$\sum_{l'=0}^{L} \sum_{m'=-l'}^{l'} \left[\int_{R} Y_{lm} Y_{l'm'} \, d\Omega \right] g_{l'm'} = \lambda g_{lm}, \qquad 0 < \lambda < 1.$$
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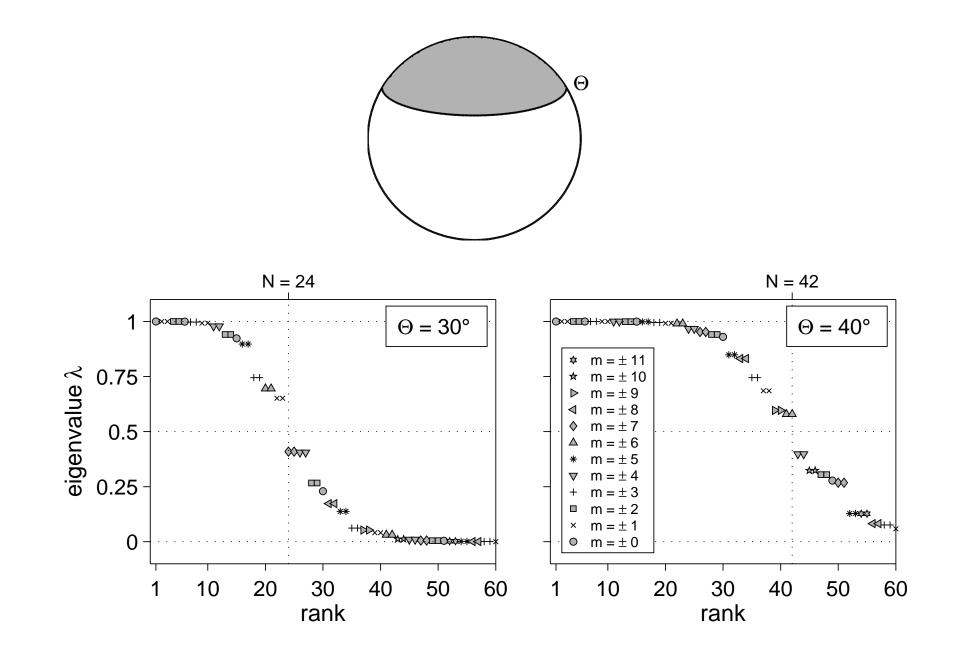
We define the **spatiospectral localization kernel**, with eigenvalues λ , as

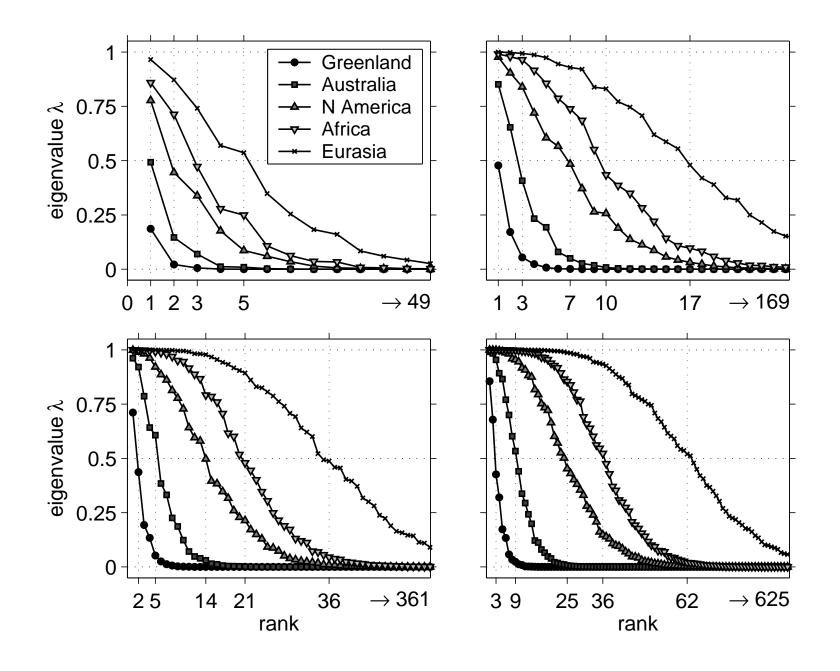
$$D_{lm,l'm'} = \int_{R} Y_{lm} Y_{l'm'} \, d\Omega, \qquad \text{tr}\{\mathsf{D}\} = (L+1)^2 \frac{A}{4\pi}.$$
 (14)

Many of the eigenvalues are very, very small. Thus, D may be hard to calculate and even harder to invert.

And remember that the spatial region R can be completely arbitrary.

Eigenvalue behavior — 1





A "lucky accident": the "magic of commutation" 12/42

Diagonalization of the operator D, with elements

$$D_{lm,l'm'} = \int_R Y_{lm} Y_{l'm'} \, d\Omega,$$

is often hard and sometimes impossible.

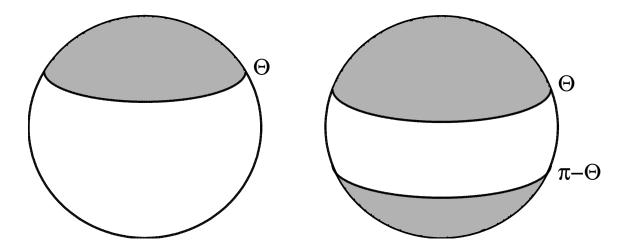
(15)

Diagonalization of the operator D, with elements

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But if R is **axisymmetric**, i.e. a **single polar cap** or a **double polar cap**, we can find the Slepian functions as the solutions to a **different** eigenvalue problem involving a *very simple* kernel with *very well-behaved* eigenvalues.



Spherical harmonics Y_{lm} form an **orthonormal** basis on Ω :

$$\int_{\Omega} Y_{lm} Y_{l'm'} \, d\Omega = \delta_{ll'} \delta_{mm'}. \tag{17}$$

The spherical harmonics Y_{lm} are **not orthogonal** on R:

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The eigenfunctions of D are called **Slepian functions**, $g(\hat{\mathbf{r}})$. They form a **bandlimited localized basis**, *doubly* orthogonal: on R (to λ) and *also* on Ω (to 1).

The Shannon number, or sum of the eigenvalues, the space-bandwidth product,

$$N = (L+1)^2 \frac{A}{4\pi},$$

is the **effective dimension** of the space for which the bandlimited *g* are a **basis**.

The expansion of a bandlimited process on the sphere in *either* spherical harmonics or in Slepian functions is equal and *exact*:

$$s(\hat{\mathbf{r}}) = \sum_{l=0}^{L} \sum_{m=-l}^{l} s_{lm} Y_{lm}(\hat{\mathbf{r}}) = \sum_{\alpha=1}^{(L+1)^2} s_{\alpha} g_{\alpha}(\hat{\mathbf{r}}).$$
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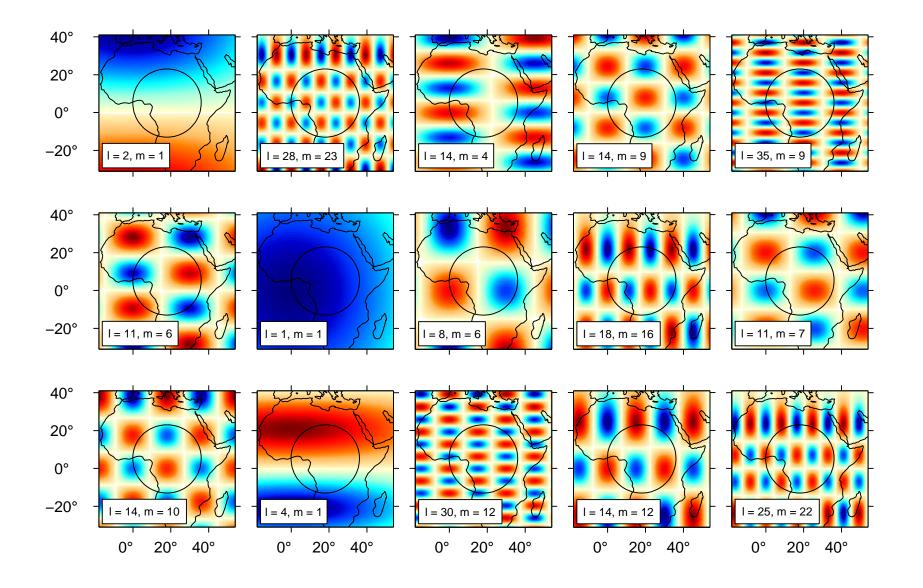
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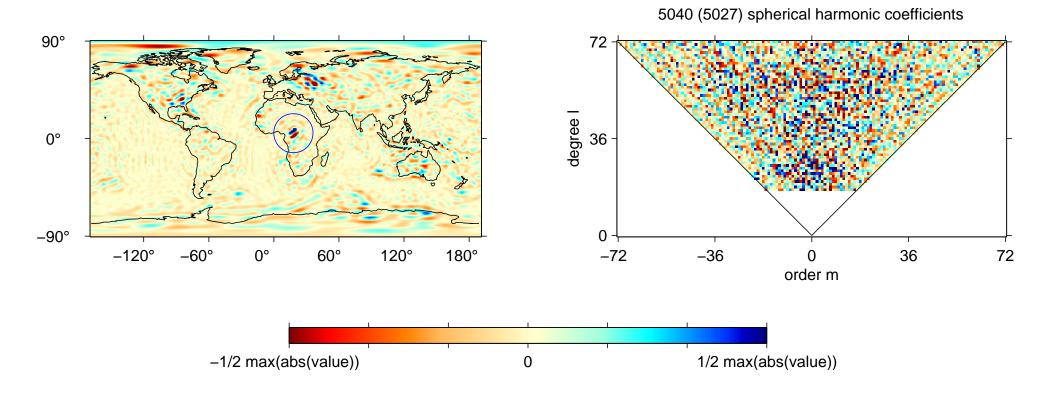
But if the signal is **regional** in nature, an expansion into Slepian functions up until the Shannon number will be **approximate but sparse**:

$$s(\hat{\mathbf{r}}) \approx \sum_{\alpha=1}^{N} s_{\alpha} g_{\alpha}(\hat{\mathbf{r}}), \quad \hat{\mathbf{r}} \in R.$$
 (20)

The *mean squared reconstruction error* in the noiseless case is determined by the neglected eigenvalues, which are **tiny** beyond the Shannon number.

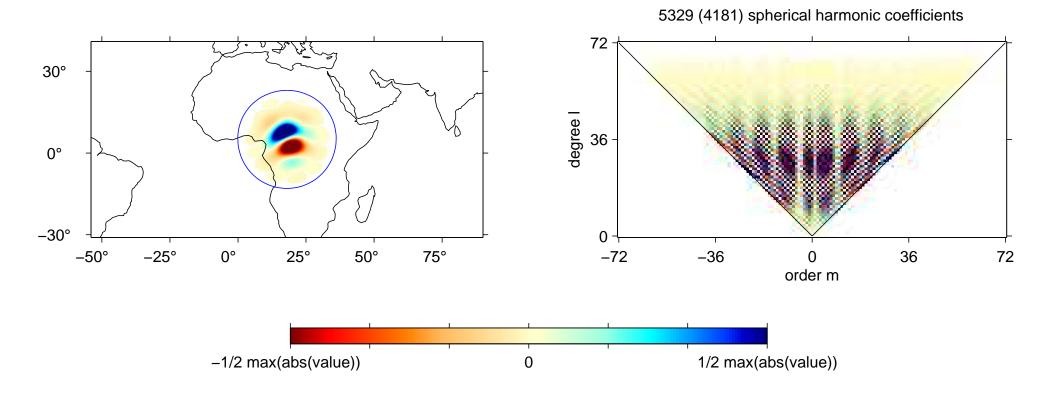


Basis I: spherical harmonics Y_{lm}



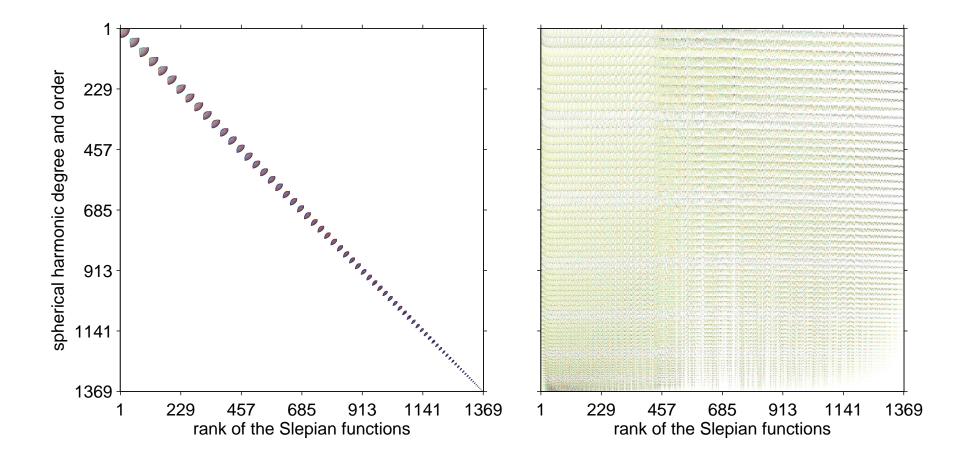
A global basis, good for global problems.

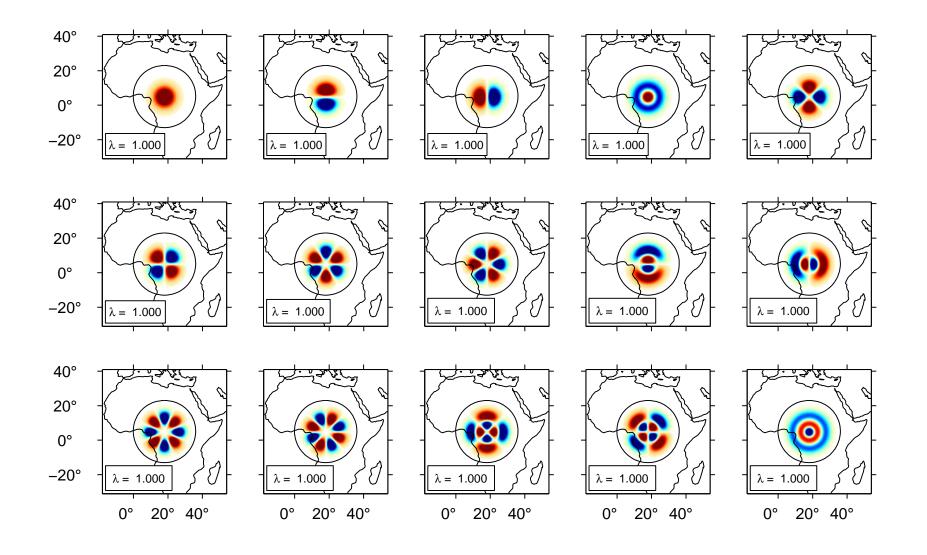
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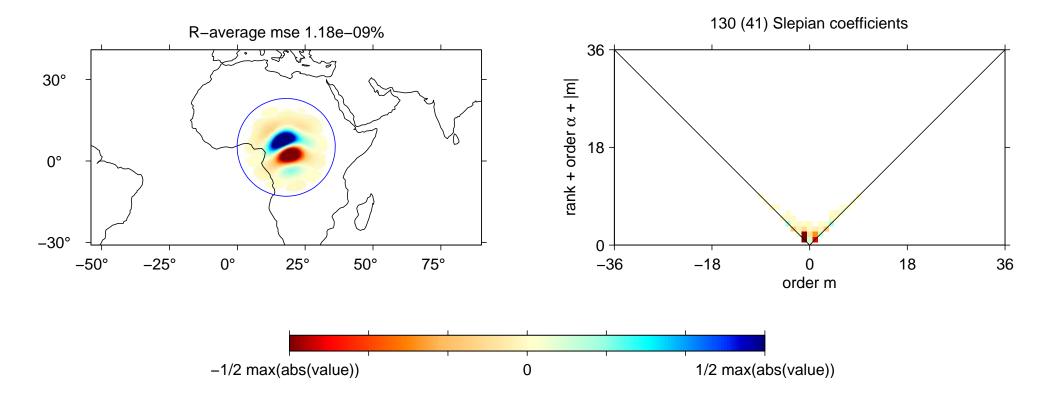


A global basis, **bad** for *local* problems.

An **orthogonal transform** by the eigenmatrix of D introduces welcome sparsity.

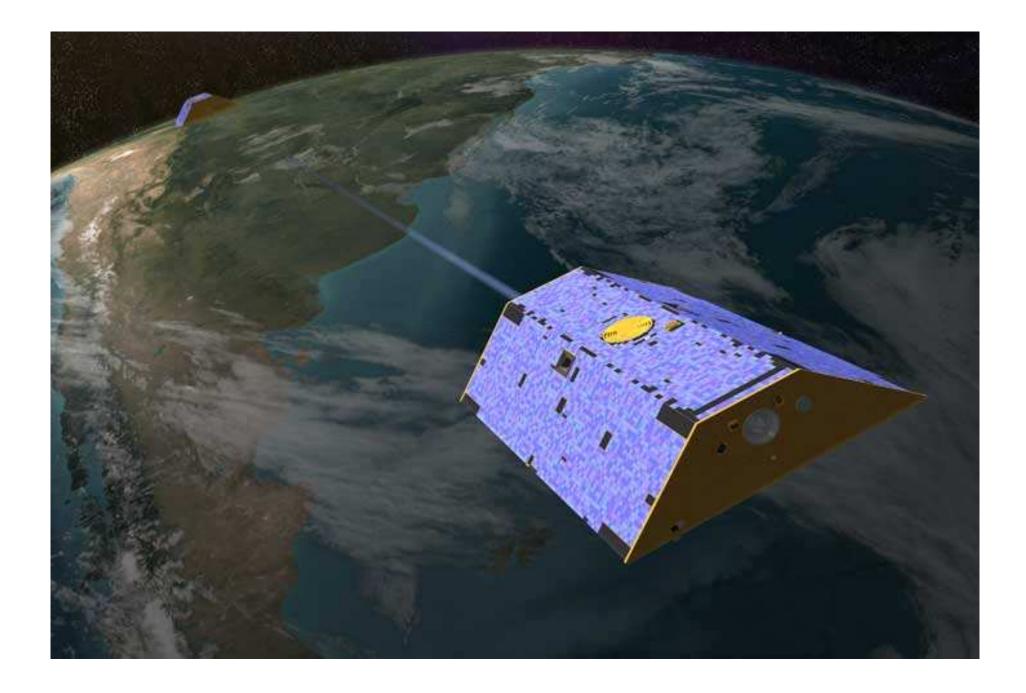




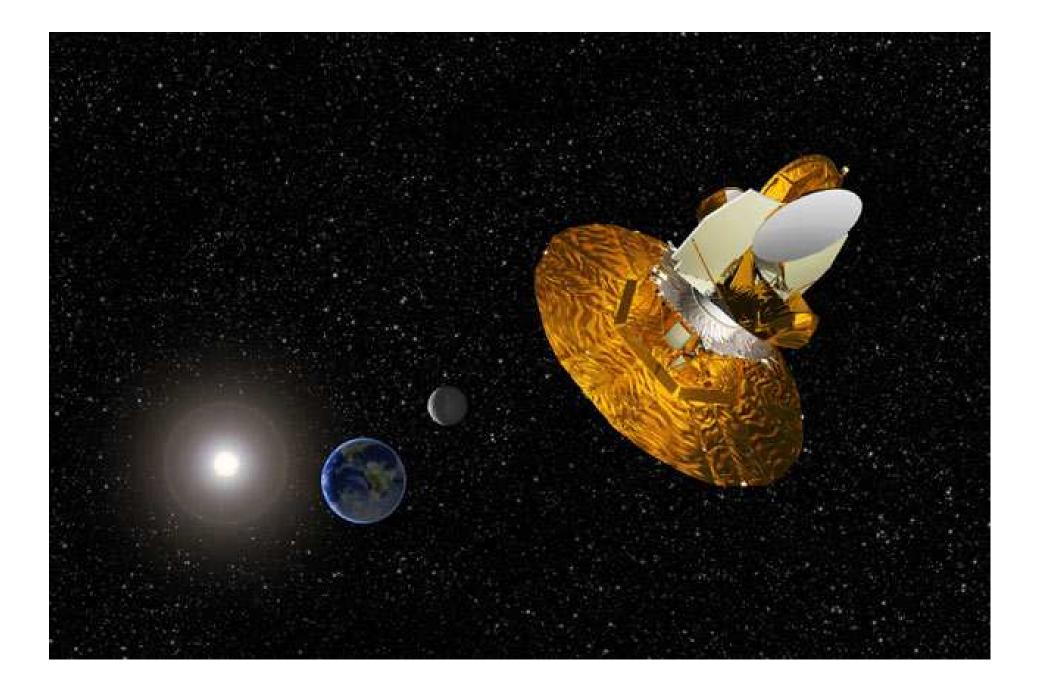


A *local* basis, **good** for *local* problems. Sparsity!

Solving problems in geophysics ...



... and cosmology



The data collected in or limited to R are signal plus noise:

We assume that $n(\mathbf{r})$ is **zero-mean** and **uncorrelated** with the signal

and consider known the **noise covariance**:

In other words: we've got **noisy** and **incomplete** data on the sphere.

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$$d(\mathbf{r}) = \begin{cases} s(\mathbf{r}) + n(\mathbf{r}) & \text{if } \mathbf{r} \in R, \\ \text{unknown/undesired} & \text{if } \mathbf{r} \in \Omega - R. \end{cases}$$

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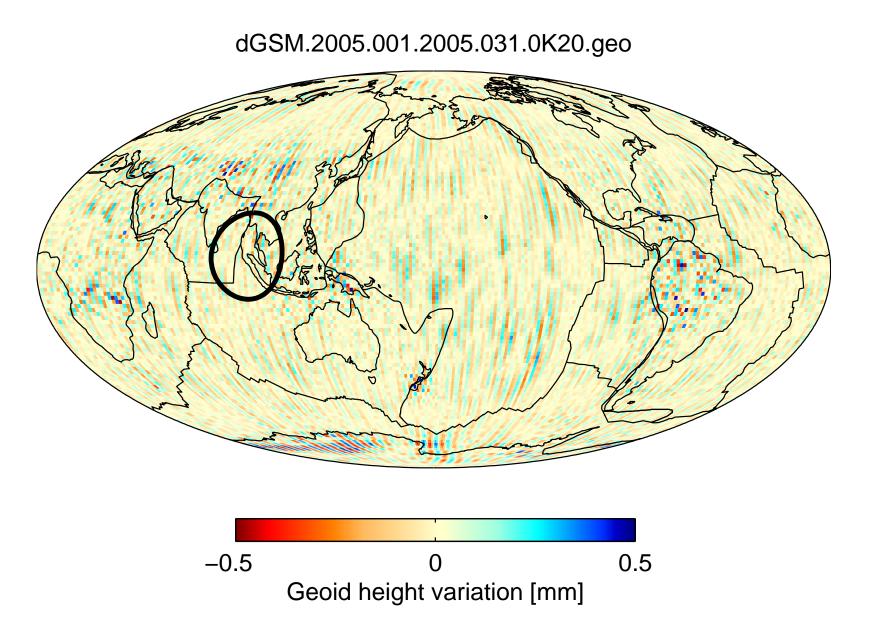
$$\langle n(\mathbf{r}) \rangle = 0$$
 and $\langle n(\mathbf{r})s(\mathbf{r'}) \rangle = 0$,

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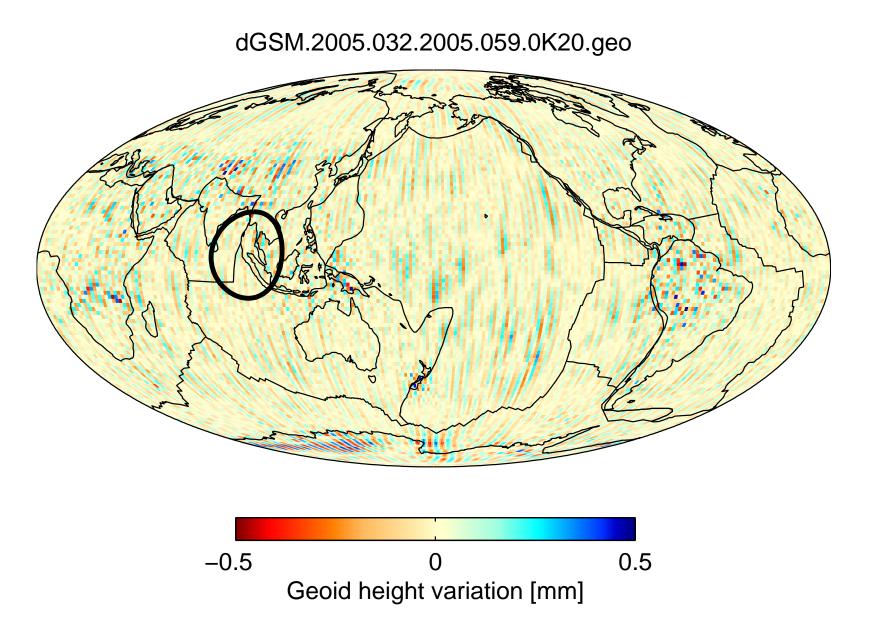
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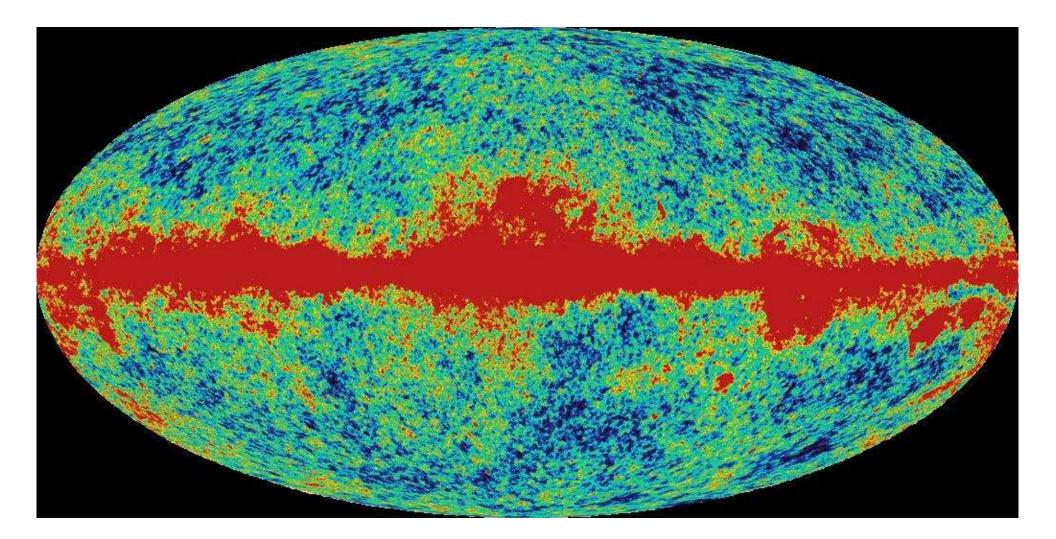
Noisy: Earth's time-variable gravity — 1



Noisy: Earth's time-variable gravity — 2



Incomplete: Cosmic Microwave Background 27/42



Common problems — 2

Consider an *unknown*, *noisily* and *incompletely observed* spherical process:

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Linear Problem:

Quadratic Problem:

Problem 2

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Given $d(\mathbf{r})$ and $\langle n(\mathbf{r})n(\mathbf{r'})\rangle$, *estimate* the signal $s(\mathbf{r})$, realizing that the estimate $\hat{s}(\mathbf{r})$ is **always bandlimited** to $0 \leq L < \infty$.

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Quadratic Problem:

Problem 2

Given $d(\mathbf{r})$ and $\langle n(\mathbf{r})n(\mathbf{r}')\rangle$, and assuming the field behaves as $\langle s_{lm}\rangle = 0$ and $\langle s_{lm}s_{l'm'}\rangle = S_l \,\delta_{ll'}\delta_{mm'},$ *estimate* the **power spectral density** S_l , for $0 \leq l < \infty$, as \hat{S}_l .

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Problem 2

Find the **power spectral density** of the signal.

Construct a **bandlimited estimate** in the spherical harmonic basis by minimizing the **misfit to the data** over R. The—*linear*—optimal solution depends on D^{-1} :

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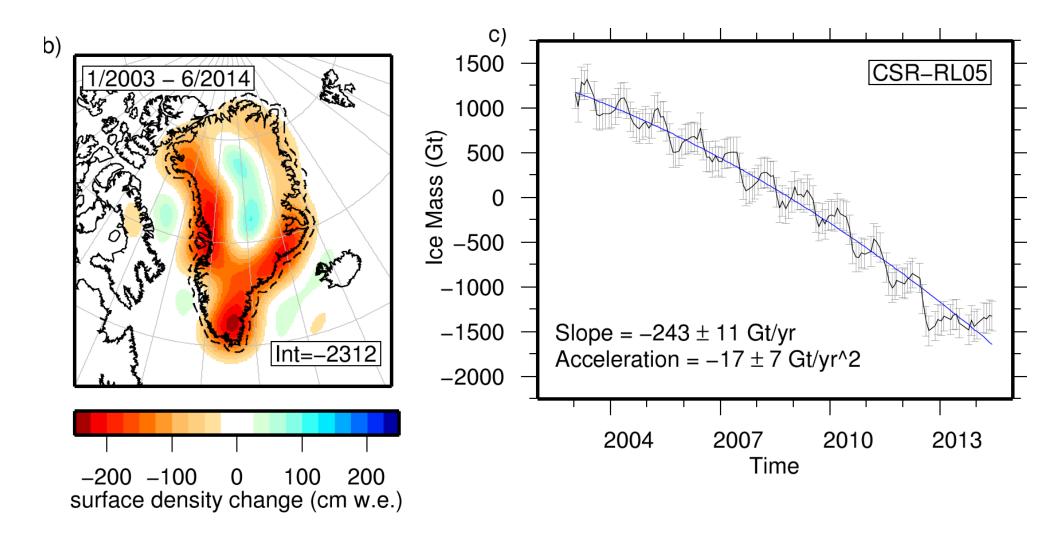
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The solution depends on the localization eigenvalue at the same rank:

$$\hat{s}_{\alpha} = \lambda_{\alpha}^{-1} \int_{R} dg_{\alpha} \, d\Omega.$$

Application 2 : Time-variable gravity



If we simply worked with the available data we'd be using a **boxcar** window:

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This estimate is **biased** (unless $S_l = S$ or $R = \Omega$), *coupling* over the *entire* band.

Its bias, variance, and thus mean-squared error depend, again, on D:

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The multitaper estimate uses a small L for the Slepian windows $g_{\alpha}(\mathbf{r})$ over R,

$$\hat{S}_{l}^{\mathrm{MT}} = \sum_{\alpha} \lambda_{\alpha} \left(\frac{1}{2l+1} \sum_{m} \left| \int_{\Omega} g_{\alpha}(\mathbf{r}) \, d(\mathbf{r}) \, Y_{lm}(\mathbf{r}) \, d\Omega \right|^{2} \right).$$

It returns a **spectrally bandlimited** (to $\pm L$) average of the true spectral power while being sensitive to a **spatially localized** patch R of data.

Spectral and spatial concentration trade off via the **Shannon number**, which is the sole parameter to be chosen by the analyst:

$$N = (L+1)^2 \frac{A}{4\pi}.$$

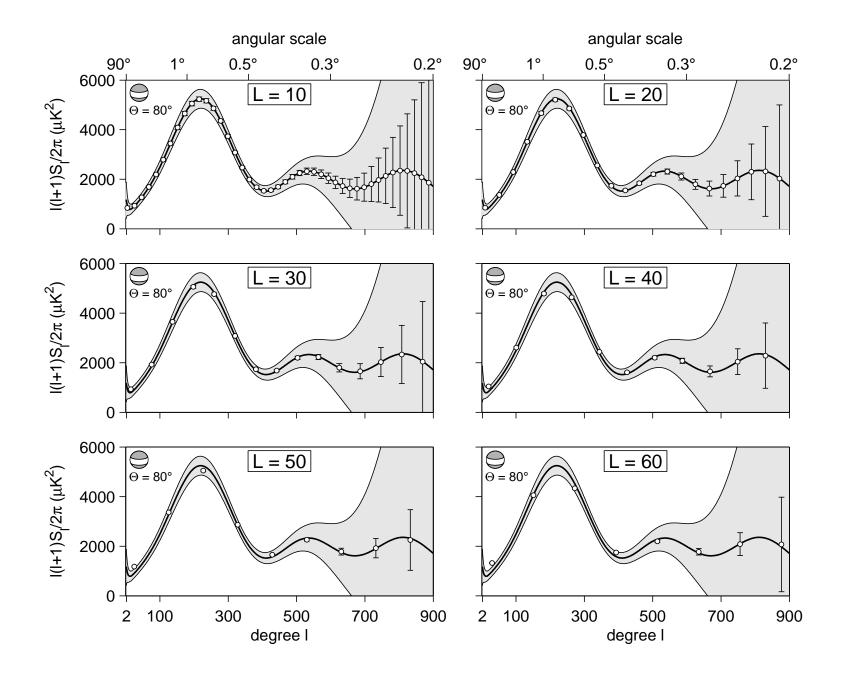
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This dictates the deliberate **bias** of the estimate. More tapers \rightarrow more bias, but the **covariance** matrix of the estimates *between* tapers is almost **diagonal**.

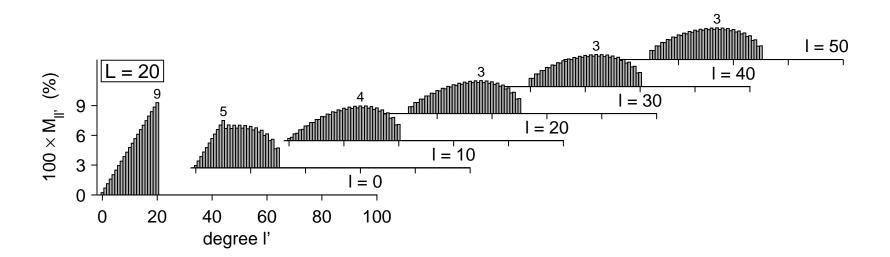
Thus, weighted averaging of estimates made with many different tapers reduces the estimation variance. And with *eigenvalue weighting*, the bias is strictly limited to the bandwidth L, and independent of the shape of the region R.



Using the choice of the **eigenvalues** λ of D as weights of the multitaper spectral estimate, the **multitaper coupling matrix** is

$$K_{ll'} = \frac{2l'+1}{(L+1)^2} \sum_{p}^{L} (2p+1) \left(\begin{array}{ccc} l & p & l' \\ 0 & 0 & 0 \end{array} \right)^2,$$

which — amazingly — depends only upon the chosen bandwidth L.

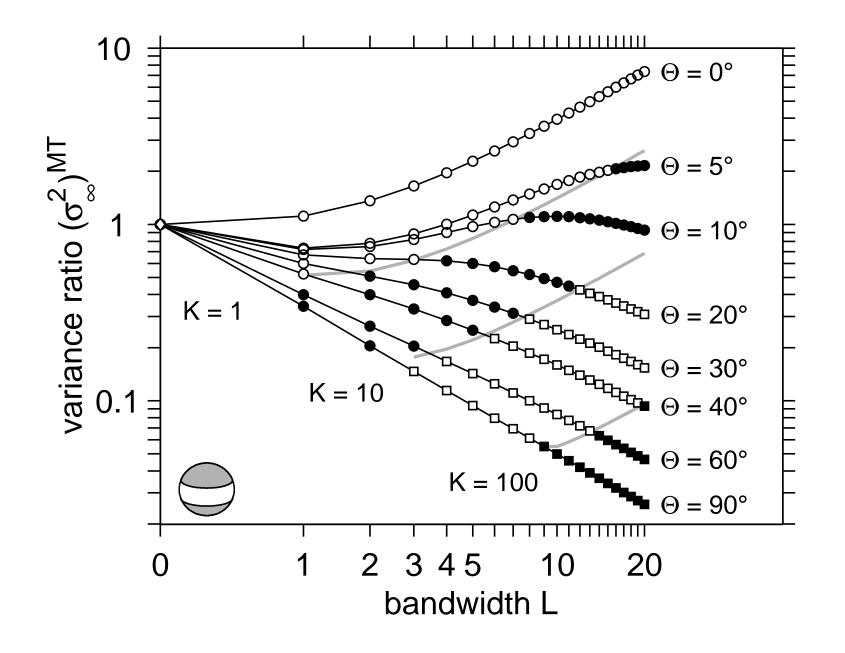


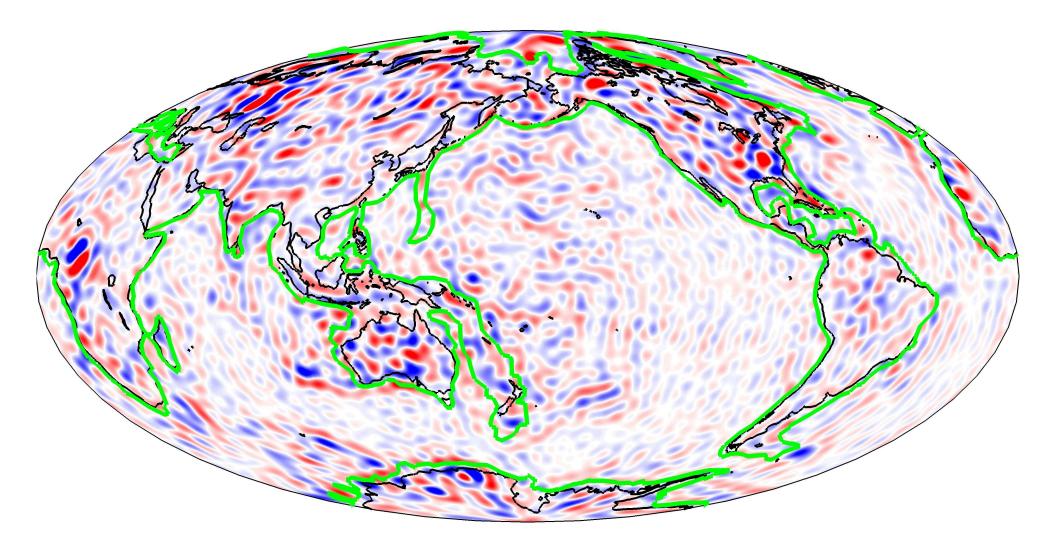
The **covariance** between the multitaper estimates is relatively simple when the spectra is **moderately colored** (compared to the bandwidth L of the estimator):

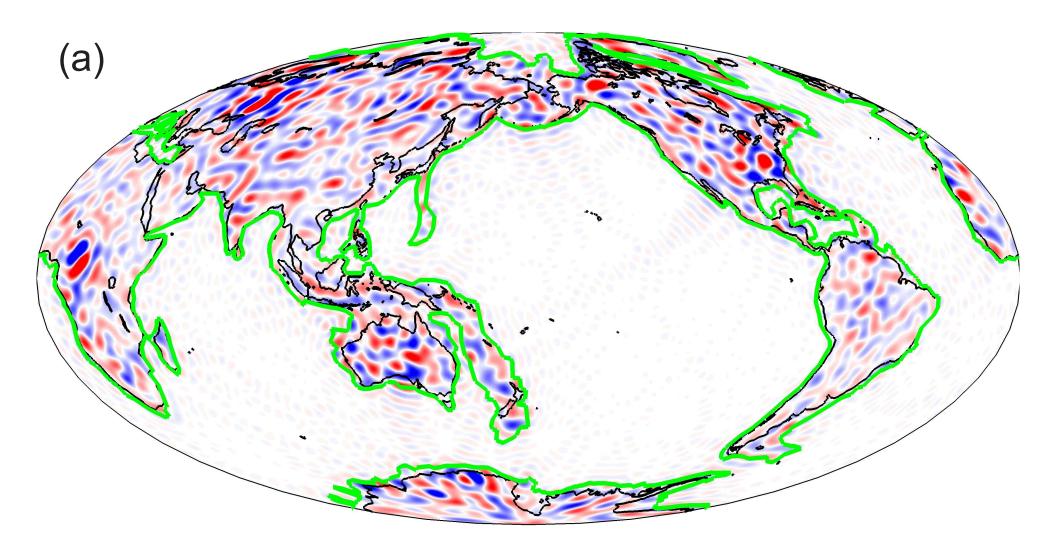
$$\Sigma_{ll'}^{\rm MT} = \frac{1}{2\pi} (S_l + N_l) (S_{l'} + N_{l'}) \sum_p (2p+1) \Gamma_p \left(\begin{array}{ccc} l & p & l' \\ 0 & 0 & 0 \end{array} \right)^2, \quad (13)$$

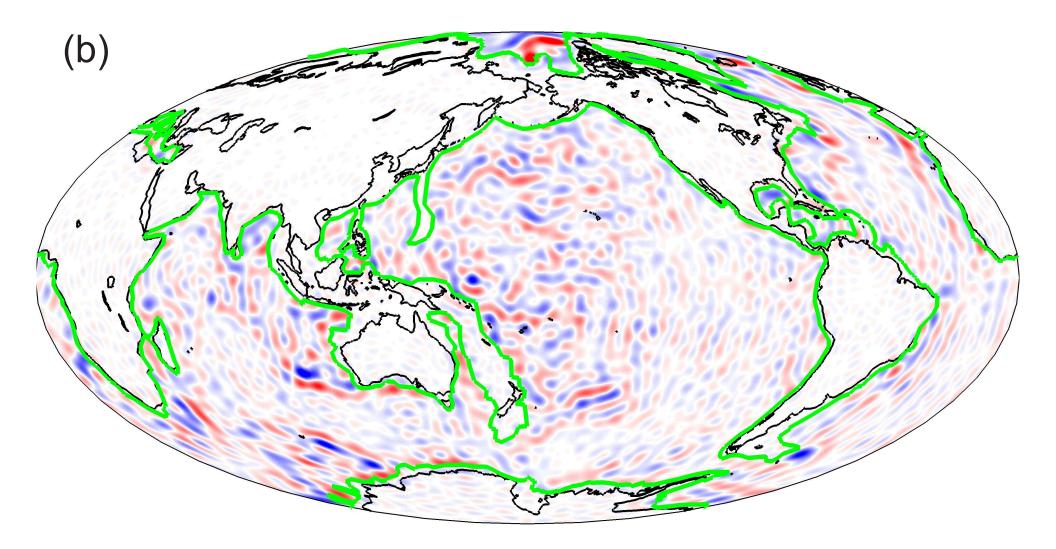
$$\Gamma_{p} = \frac{1}{K^{2}} \sum_{ss'}^{L} \sum_{uu'}^{L} (2s+1)(2s'+1)(2u+1)(2u'+1) \sum_{e}^{2L} (-1)^{p+e}(2e+1)B_{e} \\ \times \left\{ \begin{array}{ccc} s & e & s' \\ u & p & u' \end{array} \right\} \left(\begin{array}{ccc} s & e & s' \\ 0 & 0 & 0 \end{array} \right) \left(\begin{array}{ccc} u & e & u' \\ 0 & 0 & 0 \end{array} \right) \left(\begin{array}{ccc} s & p & u' \\ 0 & 0 & 0 \end{array} \right) \left(\begin{array}{ccc} u & p & s' \\ 0 & 0 & 0 \end{array} \right), \quad (14)$$

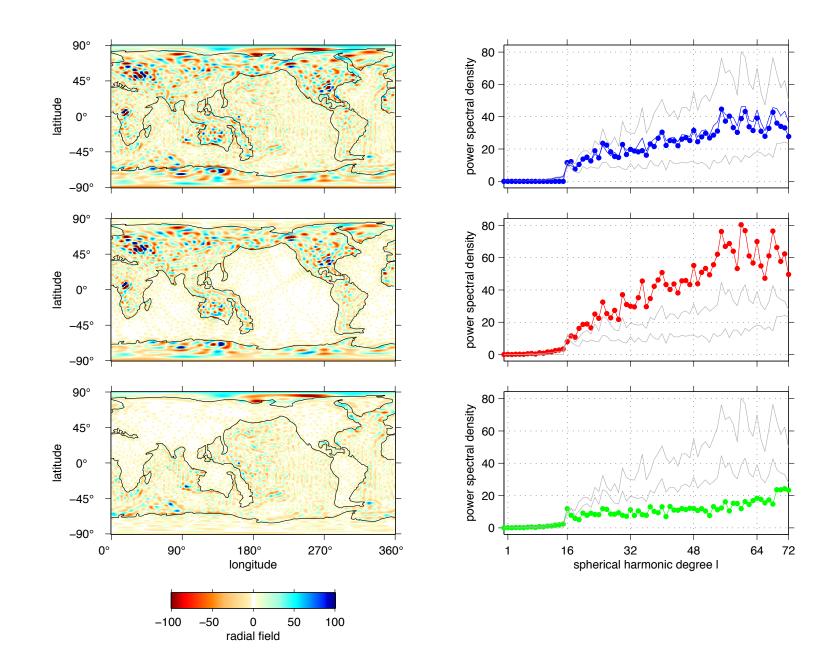
with B_e the boxcar power, which depends on the **shape** of the region of interest, and the sums over angular degrees are limited by Wigner 3-j selection rules. The term in curly braces is a Wigner 6-j symbol. Ugly, but computable.











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- The **Slepian multitaper method** yields a smoothed and thus **biased** estimate of the spectrum, but it requires neither iteration nor large-scale matrix inversion. Its **variance is much lower** than that of any other method, and the only parameter that needs to be specified by the analyst is the **Shannon number**, or the space-bandwidth product diagnostic of the spatiospectral concentration.