# On the "Optimal" Choice of Trial Functions for Modelling Potential Fields

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EGU 2015 1 / 14

## Outline



- Spherical Harmonics
- Slepian Functions (due to Simons et al.)
- Radial Basis Functions
- Locally Supported Functions
- The Algorithm RFMP
- 3 Numerical Examples
  - Multiscale Approximation of the Potential
  - Gravity Inversion
  - Mass Transports in the Amazon Area



2 / 14

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# **Trial Functions**

Examples of trial functions which are available on the sphere are

- Spherical harmonics (ideal frequency localization, no space localization) click for details
- Slepian functions •• click for details
- Radial basis functions (aka reproducing kernel based functions)
  » click for details
- Locally supported functions (very low frequency localization, high space localization) - click for details

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# Spherical Harmonics



#### Pros:

- Established system with known physical interpretation
- Efficient numerical codes available

#### Cons:

- Instabilities for irregular data grids or regional approximation
- Maximum degree is the only parameter to control the resolution
- Large data sets cause large systems of equations
- Local noise becomes global example



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## Local noise becomes global

Data are locally perturbed at a cap in the North and we interpolate the data by spherical harmonics (left-hand) and by radial basis functions (right-hand). The plots show the approximation errors.



# Slepian Functions (due to Simons et al.)



#### Pros:

- Functions can (theoretically) be calculated for every arbitrary region
- Analysis of regional effects is easily possible

## Cons:

- Functions are bandlimited, i.e. polynomials
- Resolution of an eigenvalue problem required (but with stable approach for simple geometries)



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# **Radial Basis Functions**

$$\mathcal{K}(\xi,\eta) = \sum_{n=0}^{\infty} \sum_{j=-n}^{n} k_n Y_{n,j}(\xi) Y_{n,j}(\eta) = \sum_{n=0}^{\infty} k_n \frac{2n+1}{4\pi} P_n(\xi \cdot \eta), \ \xi,\eta \in \Omega.$$



#### Pros:

Local noise remains local.

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- Resolution can be locally controlled.
- Irregular data sets can be handled (by using splines).
- Multiresolution analysis is possible (by using wavelets).

#### Cons:

- Splines: large data sets require solution of large systems of equations.
- Wavelets: quadrature rule needed, but difficult for irregular data sets.



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## Locally Supported Functions

 $B_{h,k}(\xi) = \frac{(\xi \cdot \eta - h)^k}{(1-h)^k} \cdot \chi_{[h,1]}(\xi \cdot \eta), \ \xi \in \Omega$  variable,  $\eta \in \Omega$  fixed centre



Pros:

- Space-limited functions can e.g. fade out certain areas
- Numerically very easy to implement
- Smoothness can be controlled

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### Cons:

- Spectral properties are complicated (way out: up-functions by Schreiner)
- No eigenfunctions of typical geodetic/geophysical equations



EGU 2015 8 / 14

# RFMP – A Best Basis Algorithm for Inverse Problems

We search a function F such that  $\mathcal{F}F = y$ , where  $y \in \mathbb{R}^{l}$  is given.

- $\bullet\,$  Choose a so-called dictionary  ${\cal D}$  of trial functions that might be useful
- Iteratively construct an expansion of the unknown signal F as follows:

• If 
$$F_n = \sum_{k=1}^n \alpha_k \, d_k$$

with  $\alpha_k \in \mathbb{R}$  and  $d_k \in \mathcal{D}$  has already been constructed, then add another summand

$$F_{n+1} = F_n + \alpha_{n+1} d_{n+1}$$

such that the (regularized) data misfit

$$\|y - \mathcal{F}F\|_{\mathbb{R}^{l}}^{2} + \lambda \|F\|_{\mathcal{H}(\Omega)}^{2}$$



is minimized.

— A back to beginning

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## Multiscale Approximation of the Potential

We approximate the EGM 2008 potential from irregularly distributed samples (more data on the continents) with the RFMP (30 000 iterations). The dictionary contains spherical harmonics and localized trial functions (scaling function and wavelets at regular grid of centres). The approximation is finally split up into the contributions of different trial functions (for details, see Michel and Telschow 2014).

coarse to fine approximation added details of scale J = 0, 1, 2

centres of added wavelets (dots) and coefficients (colour)



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## approximation









#### added details











# We can handle large, irregular data grids <u>and</u> we obtain a multiresolution analysis! <a>for explanations</a> <a>for explanati

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EGU 2015 11 / 14

centres

# Near-surface mass anomalies from EGM2008

density approximation (from degree 3) and centres of chosen RBFs



## Water mass transport in the Amazon area in 2008



## References

- D. Fischer, V. Michel: Sparse regularization of inverse gravimetry case study: spatial and temporal mass variations in South America, *Inverse Probl.*, 28 (2012), 065012 (34pp).
  - D. Fischer, V. Michel: Automatic best-basis selection for geophysical tomographic inverse problems, *Geophys. J. Int.*, **193** (2013), 1291-1299.
- V. Michel: Lectures on Constructive Approximation Fourier, Spline, and Wavelet Methods on the Real Line, the Sphere, and the Ball, textbook, 326 pages, Birkhäuser, New York, 2013.
- V. Michel: RFMP An iterative best basis algorithm for inverse problems in the geosciences, in: Handbook of Geomathematics (W. Freeden, M.Z. Nashed, and T. Sonar, eds.), 2nd edition, accepted, 2013.
- V. Michel, R. Telschow: A Non-linear Approximation Method on the Sphere, International Journal on Geomathematics, **5** (2014), 195-224.