

# Customized Approximation with Radial Basis Functions\*

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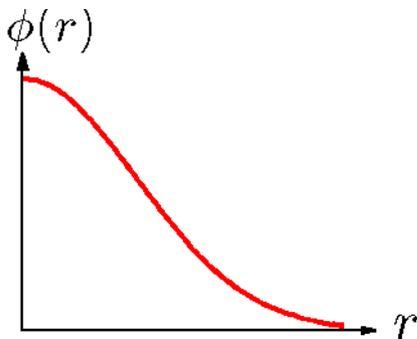
**Collaborators:** Edward J. Fuselier (USMA), Francis J. Narcowich (Texas A&M), and  
Joseph D. Ward (Texas A&M)

\* This work supported by NSF-CMG grant ATM-0620090

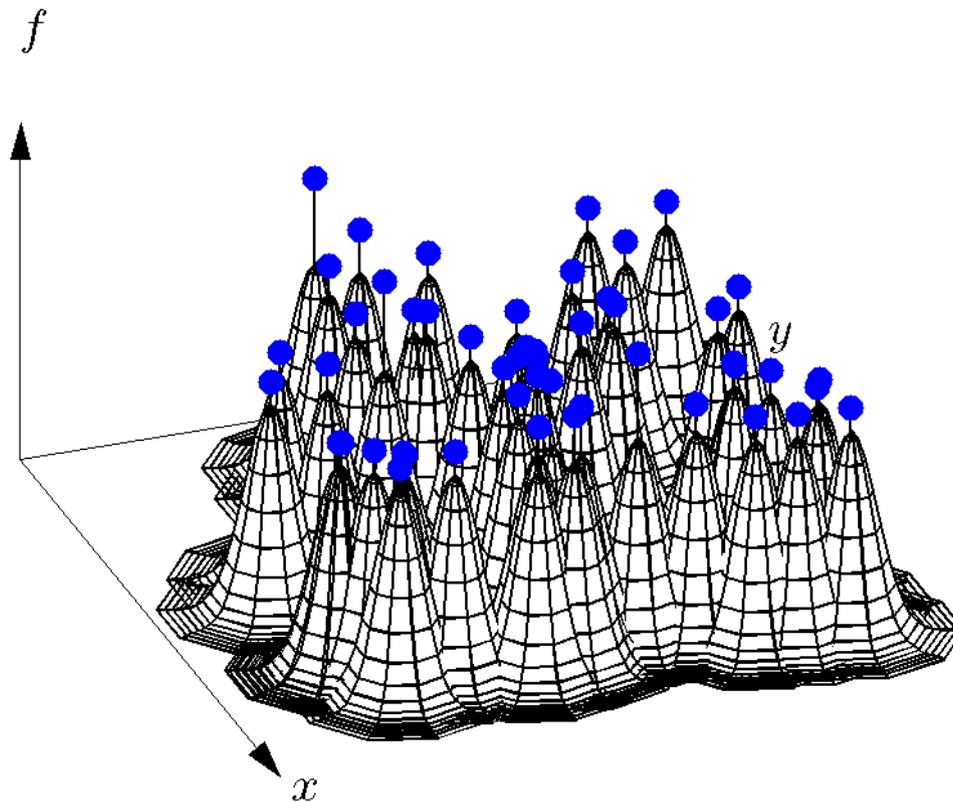
- Very quick review of Radial Basis Functions (RBFs) interpolation
- Customizing RBF approximation for vector fields:
  - Developed by **Narcowich and Ward (1994)**
  - **Divergence-free** vector fields
    - fluid flows, (static) magnetic fields
  - **Curl-free** vector fields
    - gravity fields, (static) electric fields
- RBF approximation of vector fields tangent to the surface of the sphere:
  - **Surface divergence-free** approximation
  - **Surface curl-free** approximation
  - **Helmholtz-Hodge** decomposition
- Geophysical applications

# Scalar RBF interpolation

Key idea: linear combination of **translates** and **rotations** of a **single radial function**:



1-D:  $\phi(|x-x_j|)$       > 1-D:  $\phi(\|\mathbf{x}-\mathbf{x}_j\|_2)$

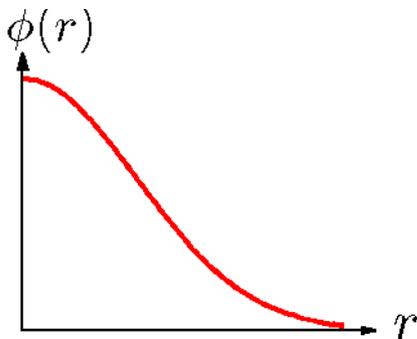


Interpolant:  $s(\mathbf{x}) = \sum_{j=1}^N \beta_j \phi(\|\mathbf{x}-\mathbf{x}_j\|)$ ,  $s(\mathbf{x}_k) = f_k$ ,  $k=1, \dots, N$

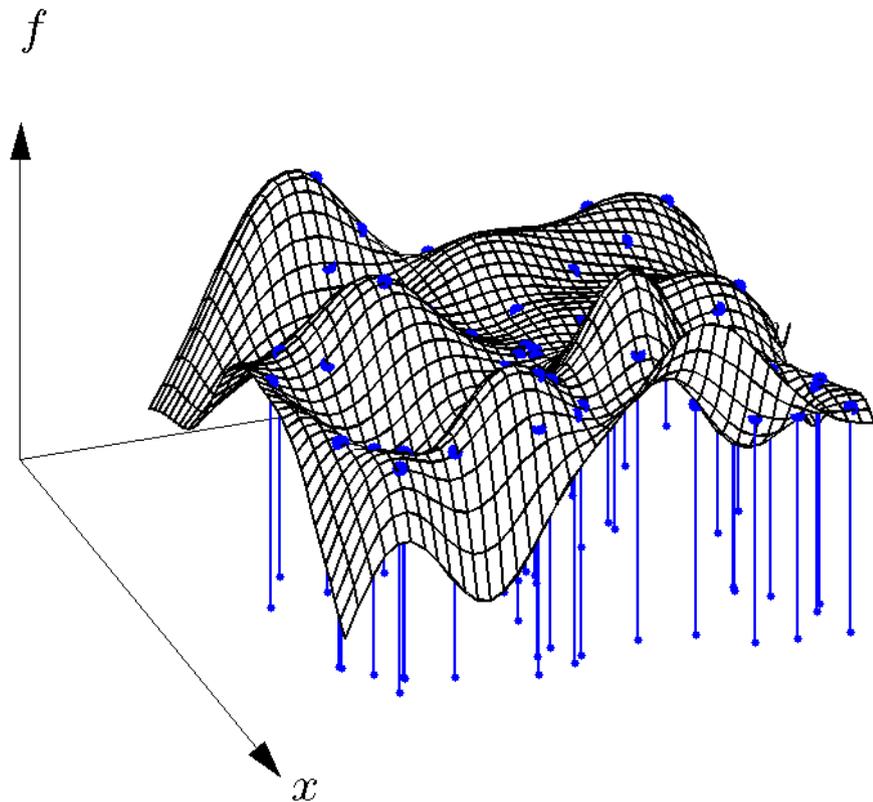
Expansion coefficients: 
$$\begin{bmatrix} \phi(\|\mathbf{x}_1-\mathbf{x}_1\|) & \phi(\|\mathbf{x}_1-\mathbf{x}_2\|) & \cdots & \phi(\|\mathbf{x}_1-\mathbf{x}_N\|) \\ \phi(\|\mathbf{x}_2-\mathbf{x}_1\|) & \phi(\|\mathbf{x}_2-\mathbf{x}_2\|) & \cdots & \phi(\|\mathbf{x}_2-\mathbf{x}_N\|) \\ \vdots & \vdots & \ddots & \vdots \\ \phi(\|\mathbf{x}_N-\mathbf{x}_1\|) & \phi(\|\mathbf{x}_N-\mathbf{x}_2\|) & \cdots & \phi(\|\mathbf{x}_N-\mathbf{x}_N\|) \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_N \end{bmatrix} = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_N \end{bmatrix},$$

# Scalar RBF interpolation

Key idea: linear combination of **translates** and **rotations** of a **single radial function**:



1-D:  $\phi(|x - x_j|)$       > 1-D:  $\phi(\|\mathbf{x} - \mathbf{x}_j\|_2)$

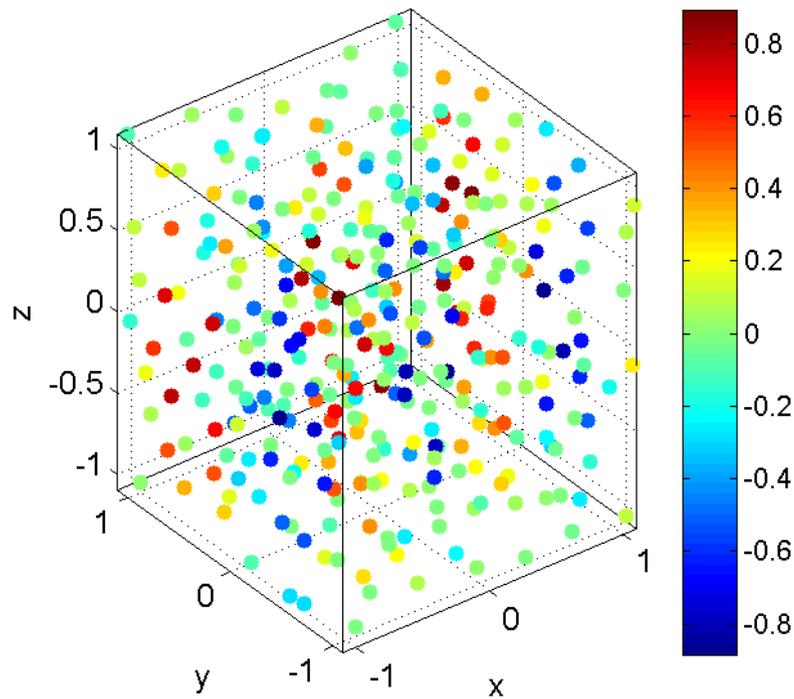


Interpolant:  $s(\mathbf{x}) = \sum_{j=1}^N \beta_j \phi(\|\mathbf{x} - \mathbf{x}_j\|)$ ,  $s(\mathbf{x}_k) = f_k$ ,  $k=1, \dots, N$

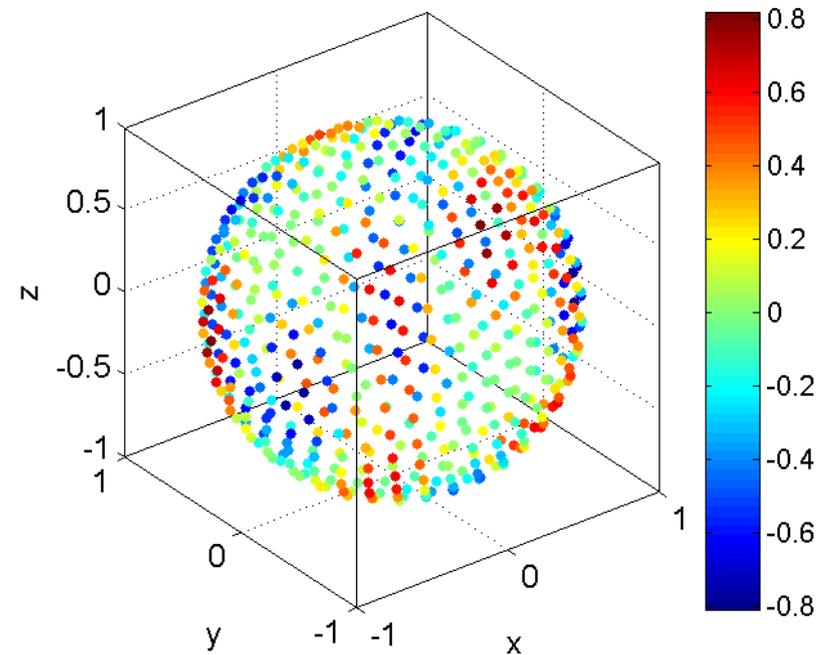
Expansion coefficients: 
$$\begin{bmatrix} \phi(\|\mathbf{x}_1 - \mathbf{x}_1\|) & \phi(\|\mathbf{x}_1 - \mathbf{x}_2\|) & \cdots & \phi(\|\mathbf{x}_1 - \mathbf{x}_N\|) \\ \phi(\|\mathbf{x}_2 - \mathbf{x}_1\|) & \phi(\|\mathbf{x}_2 - \mathbf{x}_2\|) & \cdots & \phi(\|\mathbf{x}_2 - \mathbf{x}_N\|) \\ \vdots & \vdots & \ddots & \vdots \\ \phi(\|\mathbf{x}_N - \mathbf{x}_1\|) & \phi(\|\mathbf{x}_N - \mathbf{x}_2\|) & \cdots & \phi(\|\mathbf{x}_N - \mathbf{x}_N\|) \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_N \end{bmatrix} = \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_N \end{bmatrix},$$

Guaranteed  
**positive-definite**  
for appropriate  
 $\phi(r)$

Interpolation on  $\mathbb{R}^3$



Interpolation on  $S^2 \subset \mathbb{R}^3$



- **Scalar** RBF interpolant does not change:

$$s(\mathbf{x}) = \sum_{j=1}^N \beta_j \phi(\|\mathbf{x} - \mathbf{x}_j\|), \quad s(\mathbf{x}_k) = f_k, \quad k=1, \dots, N$$

- **Divergence-free** and **curl-free** RBF interpolants do!

# (Surface) Div, Grad, Curl, and all that

Workshop on Petascale  
Computing, NCAR 2008

Spherical Coords.

Cartesian Coords.

Point:  $(\lambda, \theta, 1)$

$(x, y, z)$

Unit vectors:  $\hat{\mathbf{i}}$  = longitudinal  
 $\hat{\mathbf{j}}$  = latitudinal  
 $\hat{\mathbf{k}}$  = radial

$\hat{\mathbf{i}}$  =  $x$ -direction  
 $\hat{\mathbf{j}}$  =  $y$ -direction  
 $\hat{\mathbf{k}}$  =  $z$ -direction

Unit tangent vectors:  $\hat{\mathbf{i}}, \hat{\mathbf{j}}$

$$\zeta = \frac{1}{\sqrt{1-z^2}} \begin{bmatrix} -y \\ x \\ 0 \end{bmatrix}, \quad \mu = \frac{1}{\sqrt{1-z^2}} \begin{bmatrix} -zx \\ -zy \\ 1-z^2 \end{bmatrix}$$

Unit normal vector:  $\hat{\mathbf{k}}$

$$\mathbf{x} = x\hat{\mathbf{i}} + y\hat{\mathbf{j}} + z\hat{\mathbf{k}}$$

Gradient of scalar  $g$ :  $\mathbf{u}_s = \nabla_s g = \frac{1}{\cos\theta} \frac{\partial g}{\partial \lambda} \hat{\mathbf{i}} + \frac{\partial g}{\partial \theta} \hat{\mathbf{j}}$

$$\mathbf{u}_c = P_{\mathbf{x}}(\nabla_c g) = P_{\mathbf{x}} \left( \frac{\partial g}{\partial x} \hat{\mathbf{i}} + \frac{\partial g}{\partial y} \hat{\mathbf{j}} + \frac{\partial g}{\partial z} \hat{\mathbf{k}} \right)$$

Surface divergence of  $\mathbf{u}$ :  $\nabla_s \cdot \mathbf{u}_s = \nabla_s^T \mathbf{u}_s = \frac{1}{\cos\theta} \frac{\partial u_s}{\partial \lambda} + \frac{\partial v_s}{\partial \theta}$

$$(\nabla_c P_{\mathbf{x}})^T \mathbf{u}_c = \nabla_c^T (P_{\mathbf{x}} \mathbf{u}_c)$$

Curl of a scalar  $f$ :  $\mathbf{u}_s = \hat{\mathbf{k}} \times (\nabla_s f) = -\frac{\partial f}{\partial \theta} \hat{\mathbf{i}} + \frac{1}{\cos\theta} \frac{\partial f}{\partial \lambda} \hat{\mathbf{j}}$

$$\mathbf{u}_c = \mathbf{x} \times (P_{\mathbf{x}} \nabla_c f) = Q_{\mathbf{x}} P_{\mathbf{x}} (\nabla_c f) = Q_{\mathbf{x}} (\nabla_c f)$$

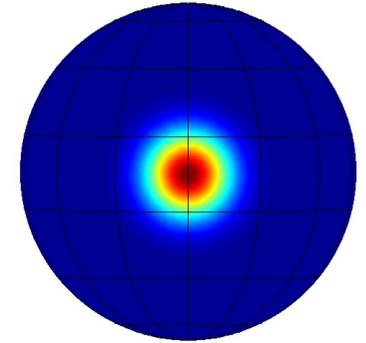
Surface curl of a vector  $\mathbf{u}$ :  $\hat{\mathbf{k}} \cdot (\nabla_s \times \mathbf{u}_s) = -\nabla_s^T (\hat{\mathbf{k}} \times \mathbf{u}_s)$

$$(Q_{\mathbf{x}} \nabla_c)^T \mathbf{u}_c = \nabla_c^T (Q_{\mathbf{x}}^T \mathbf{u}_c) = -\nabla_c^T (Q_{\mathbf{x}} \mathbf{u}_c)$$

where  $P_{\mathbf{x}} = I - \mathbf{x}\mathbf{x}^T = \begin{bmatrix} 1-x^2 & -xy & -xz \\ -xy & 1-y^2 & -yz \\ -xz & -yz & 1-z^2 \end{bmatrix}$  and  $Q_{\mathbf{x}} = \begin{bmatrix} 0 & -z & y \\ z & 0 & -x \\ -y & x & 0 \end{bmatrix}$

- Developed by **Narcowich, Ward, and Wright (2007)**

- Use extrinsic (Cartesian) coordinates,  $\mathbf{x} = (x, y, z) \in \mathbb{S}^2$ .
- Start with a **radial function** centered  $\mathbf{x}_0 \in \mathbb{S}^2$ :  $\phi(\|\mathbf{x} - \mathbf{x}_0\|)$
- Construct 3-by-3 *matrix-valued* function



$$\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0) = -Q_{\mathbf{x}}(\nabla\nabla^T \phi(\|\mathbf{x} - \mathbf{x}_0\|))Q_{\mathbf{x}_0}^T.$$

- If  $\mathbf{c} = (c_1, c_2, c_3)^T$  is tangent to  $\mathbb{S}^2$  at  $\mathbf{x}_0$  then  $\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0)\mathbf{c}$  is tangent to  $\mathbb{S}^2$  at  $\mathbf{x}$ .
- Furthermore,

$$\begin{aligned}\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0)\mathbf{c} &= [-Q_{\mathbf{x}}(\nabla\nabla^T \phi(\|\mathbf{x} - \mathbf{x}_0\|))Q_{\mathbf{x}_0}^T] \mathbf{c} \\ &= Q_{\mathbf{x}} \nabla [\nabla^T (\phi(\|\mathbf{x} - \mathbf{x}_0\|)Q_{\mathbf{x}_0}\mathbf{c})] \\ &= Q_{\mathbf{x}}(\nabla f).\end{aligned}$$

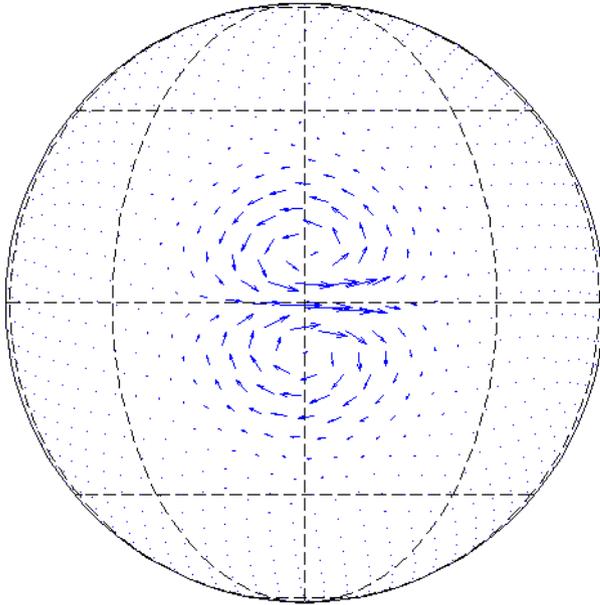
Thus,  $\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0)\mathbf{c}$  is **surface divergence-free**.

- Idea can be extended to other smooth manifolds.

# Surface div-free RBF interpolation

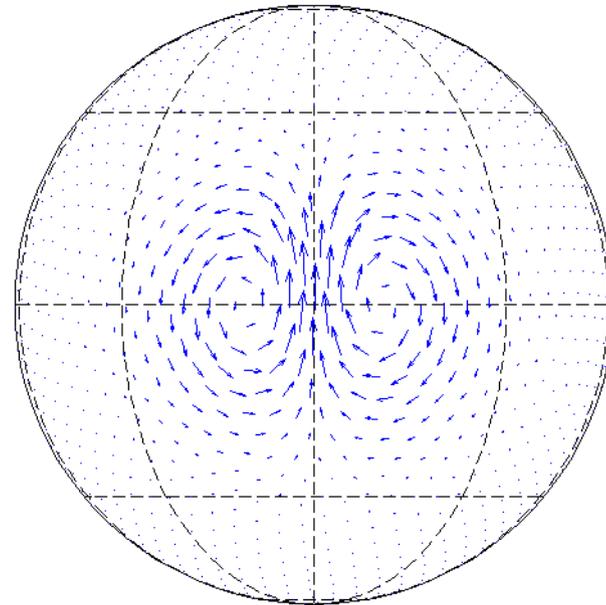
- Illustration of new basis (orthographic projection):

Zonal basis



$$\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0)\zeta_0$$

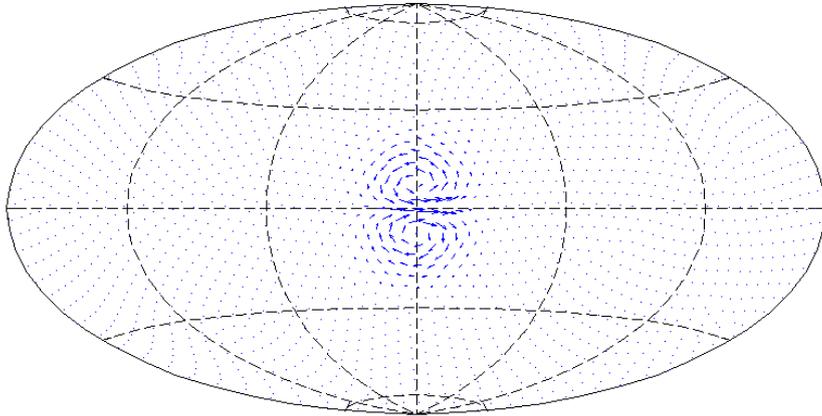
Meridional basis



$$\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0)\mu_0$$

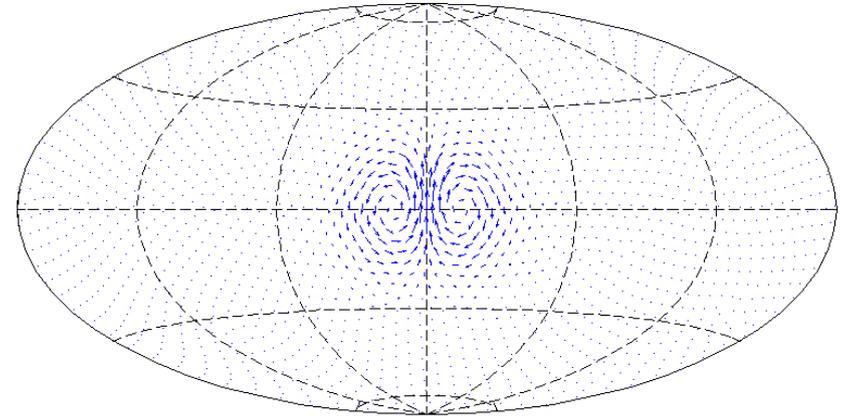
- Illustration of new basis (Hammer-Aitoff projection):

Zonal basis



$$\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0)\zeta_0$$

Meridional basis



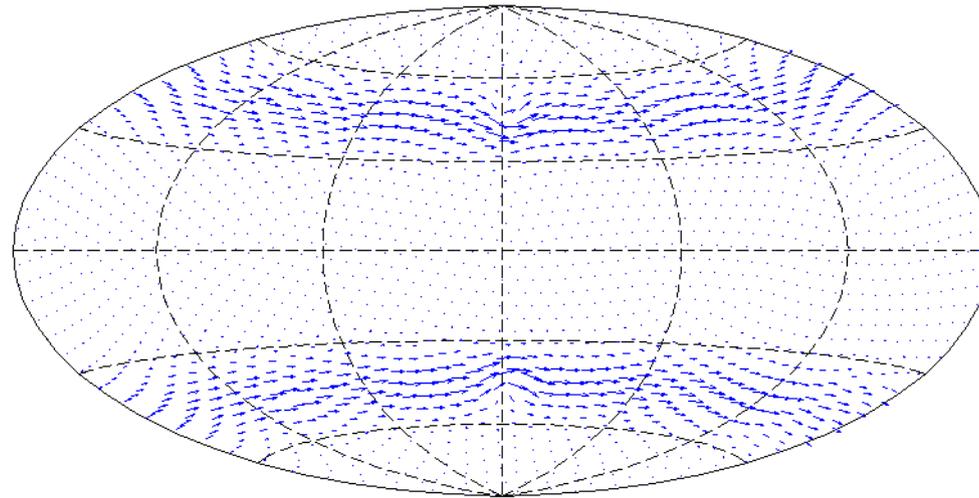
$$\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0)\mu_0$$

- Construction similar to the scalar RBF interpolant :
  1. For each node  $\mathbf{x}_j$ , center a  $\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j)\mu_j$  and  $\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j)\zeta_j$ .
  2. Linearly combine these *vector-valued* functions to satisfy the interpolation conditions.
  3. Interpolant will be **surface divergence-free**.
  4. AND **free of any pole singularity**.

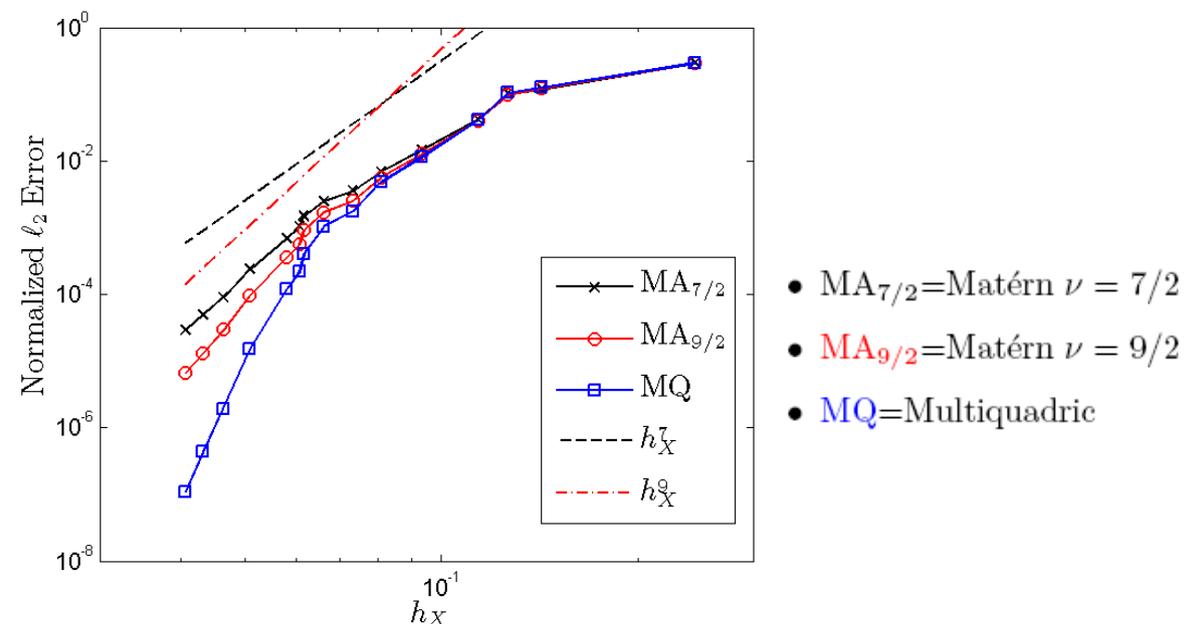
  - Slight modification needed to uniquely solve for the interpolation coefficients.

# Surface div-free RBF interpolation: examples

- Smooth divergence-free test field sampled at “scattered” nodes on the sphere.



- Error in the RBF reconstructed field vs. node spacing (log-log scale):

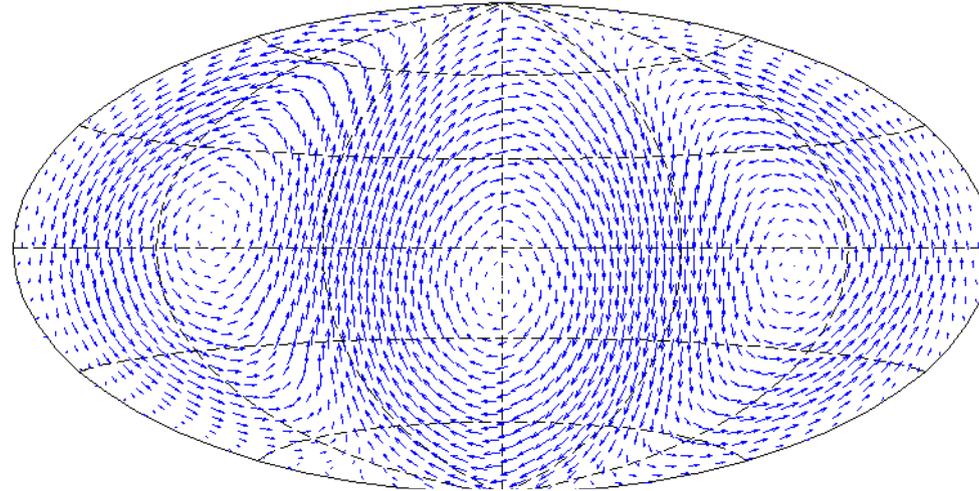


- Dashed and dashed-dotted lines predicted error rates for Matérn (Fuselier *et. al.* 2008.)

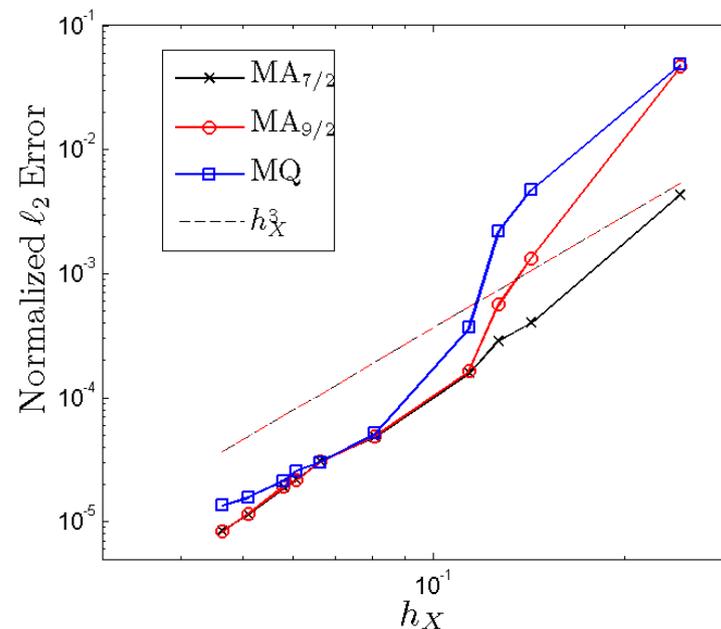
- Divergence of recovered field = 0.

# Surface div-free RBF interpolation: examples

- Less smooth divergence-free test field sampled at “scattered” nodes on the sphere.



- Error in the RBF reconstructed field vs. node spacing (log-log scale):



- MA<sub>7/2</sub>=Matérn  $\nu = 7/2$
- MA<sub>9/2</sub>=Matérn  $\nu = 9/2$
- MQ=Multiquadric

- Dashed and dashed-dotted lines predicted error rates for Matérn (Fuselier *et. al.* 2008.)

- Divergence of recovered field = 0.

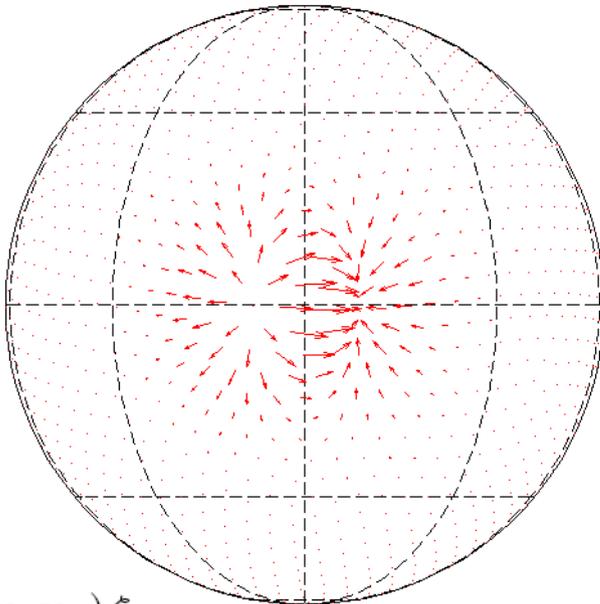
- **Surface curl-free** basis:

- Use extrinsic (Cartesian) coordinates,  $\mathbf{x} = (x, y, z) \in \mathbb{S}^2$ .
- $\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0) = -P_{\mathbf{x}}(\nabla\nabla^T\phi(\|\mathbf{x} - \mathbf{x}_0\|))P_{\mathbf{x}_0}^T$ .
- If  $\mathbf{c} = (c_1, c_2, c_3)^T$  is tangent to  $\mathbb{S}^2$  at  $\mathbf{x}_0$  then  $\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0)\mathbf{c}$  is tangent to  $\mathbb{S}^2$  at  $\mathbf{x}$ .
- Furthermore,  $\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0)\mathbf{c} = P_{\mathbf{x}}\nabla \underbrace{[-\nabla^T\phi(\|\mathbf{x} - \mathbf{x}_0\|)]P_{\mathbf{x}_0}\mathbf{c}}_g = P_{\mathbf{x}}(\nabla g)$ .

Thus,  $\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0)\mathbf{c}$  is **surface curl-free**.

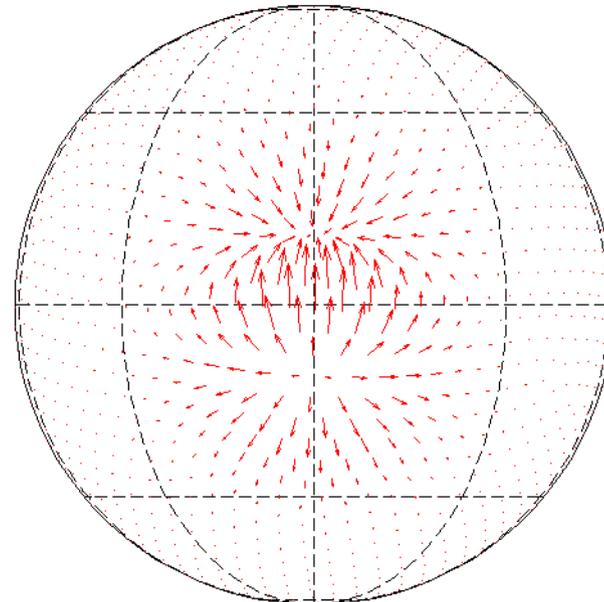
- Illustration of new basis (orthographic projection):

Zonal basis



$$\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0)\zeta_0$$

Meridional basis



$$\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0)\mu_0$$

# Helmholtz-Hodge Decomposition

- Any vector field tangent to the sphere can be *uniquely* decomposed into **surface divergence-free** and **surface curl-free** components:

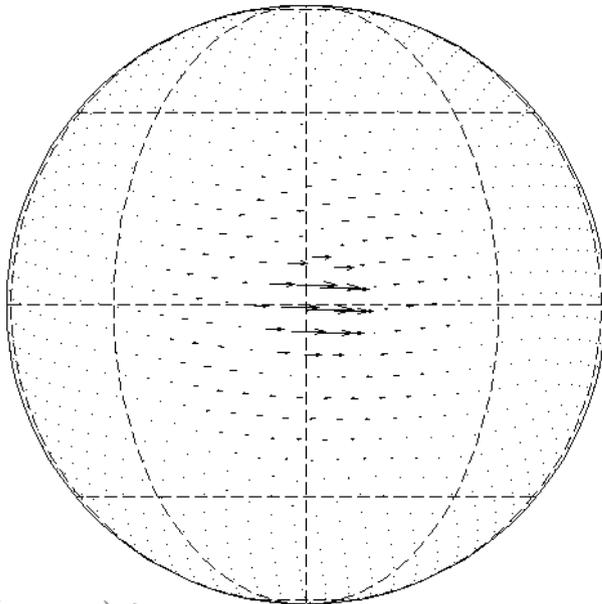
$$\begin{aligned}\mathbf{u}(\mathbf{x}) &= \mathbf{u}_{\text{div}}(\mathbf{x}) + \mathbf{u}_{\text{curl}}(\mathbf{x}) \\ &= Q_{\mathbf{x}}\nabla\psi(\mathbf{x}) + P_{\mathbf{x}}\nabla\chi(\mathbf{x})\end{aligned}$$

$\psi$  = stream function and  $\chi$  = velocity potential

- Matrix-valued Helmholtz-Hodge RBF:

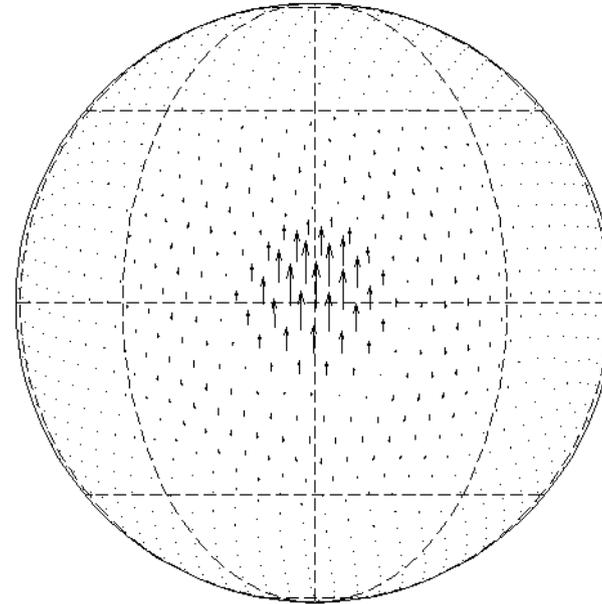
$$\Psi(\mathbf{x}, \mathbf{x}_0) = \Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_0) + \Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0)$$

Zonal basis



$$\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0)\zeta_0$$

Meridional basis



$$\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_0)\mu_0$$

- Any vector field tangent to the sphere can be *uniquely* decomposed into **surface divergence-free** and **surface curl-free** components:

$$\begin{aligned}\mathbf{u}(\mathbf{x}) &= \mathbf{u}_{\text{div}}(\mathbf{x}) + \mathbf{u}_{\text{curl}}(\mathbf{x}) \\ &= Q_{\mathbf{x}} \nabla \psi(\mathbf{x}) + P_{\mathbf{x}} \nabla \chi(\mathbf{x})\end{aligned}$$

$\psi$  = stream function and  $\chi$  = velocity potential

- Helmholtz-Hodge RBF interpolant of  $\mathbf{f}$  sampled at  $\mathbf{x}_j$ :

$$\begin{aligned}\mathbf{s}(\mathbf{x}) &= \sum_{j=1}^N \Psi(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j && \text{(where } \mathbf{s}(\mathbf{x}_j) = \mathbf{u}_j, j = 1, \dots, N\text{)} \\ &= \sum_{j=1}^N [\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) + \Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j)] \mathbf{c}_j \\ &= \underbrace{\sum_{j=1}^N \Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j}_{\approx \mathbf{u}_{\text{div}}} + \underbrace{\sum_{j=1}^N \Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j}_{\approx \mathbf{u}_{\text{curl}}}\end{aligned}$$

- Can get an approximation to the **surface divergence-free** and **surface curl-free** components!

- Any vector field tangent to the sphere can be *uniquely* decomposed into **surface divergence-free** and **surface curl-free** components:

$$\begin{aligned}\mathbf{u}(\mathbf{x}) &= \mathbf{u}_{\text{div}}(\mathbf{x}) + \mathbf{u}_{\text{curl}}(\mathbf{x}) \\ &= Q_{\mathbf{x}} \nabla \psi(\mathbf{x}) + P_{\mathbf{x}} \nabla \chi(\mathbf{x})\end{aligned}$$

$\psi$  = stream function and  $\chi$  = velocity potential

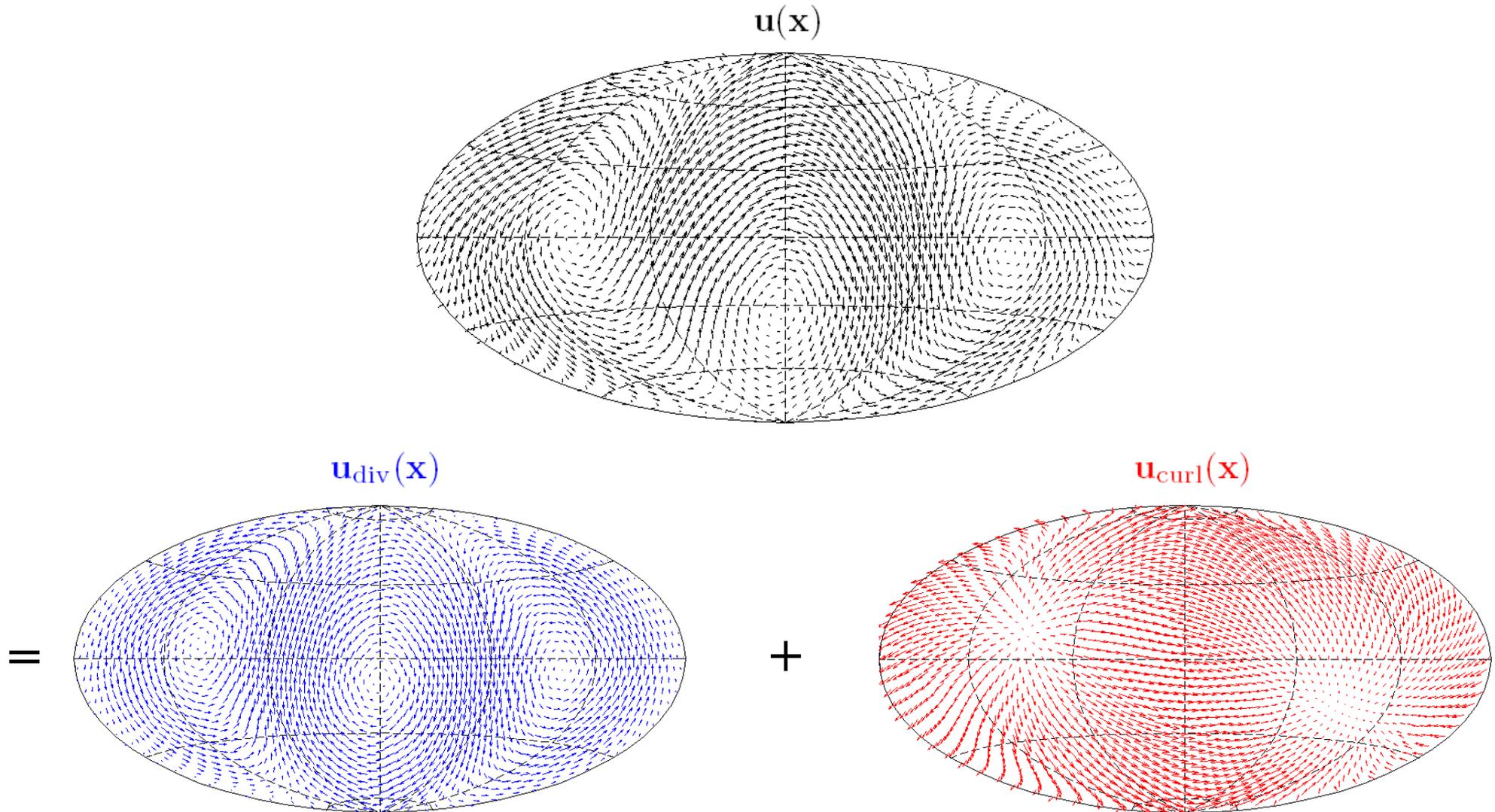
- Helmholtz-Hodge RBF interpolant of  $\mathbf{f}$  sampled at  $\mathbf{x}_j$ :

$$\begin{aligned}\mathbf{s}(\mathbf{x}) &= \sum_{j=1}^N \Psi(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j && \text{(where } \mathbf{s}(\mathbf{x}_j) = \mathbf{u}_j, j = 1, \dots, N\text{)} \\ &= \sum_{j=1}^N [\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) + \Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j)] \mathbf{c}_j \\ &= \underbrace{Q_{\mathbf{x}} \nabla \left[ \sum_{j=1}^N \nabla^T \phi(\|\mathbf{x} - \mathbf{x}_j\|) Q_{\mathbf{x}_j}^T \mathbf{c}_j \right]}_{\text{stream function for } \mathbf{s}} + \underbrace{P_{\mathbf{x}} \nabla \left[ \sum_{j=1}^N \nabla^T \phi(\|\mathbf{x} - \mathbf{x}_j\|) P_{\mathbf{x}_j}^T \mathbf{c}_j \right]}_{\text{velocity potential for } \mathbf{s}}\end{aligned}$$

- Can get a stream function and velocity potential for the interpolant.

# Helmholtz-Hodge Decomposition: Example 1

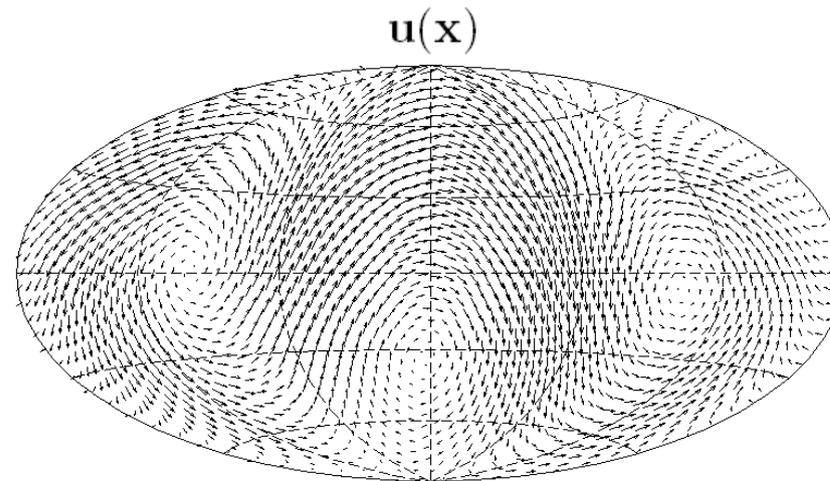
- Tangent vector field  $\mathbf{u}(\mathbf{x})$  sampled at “scattered” nodes on the sphere.



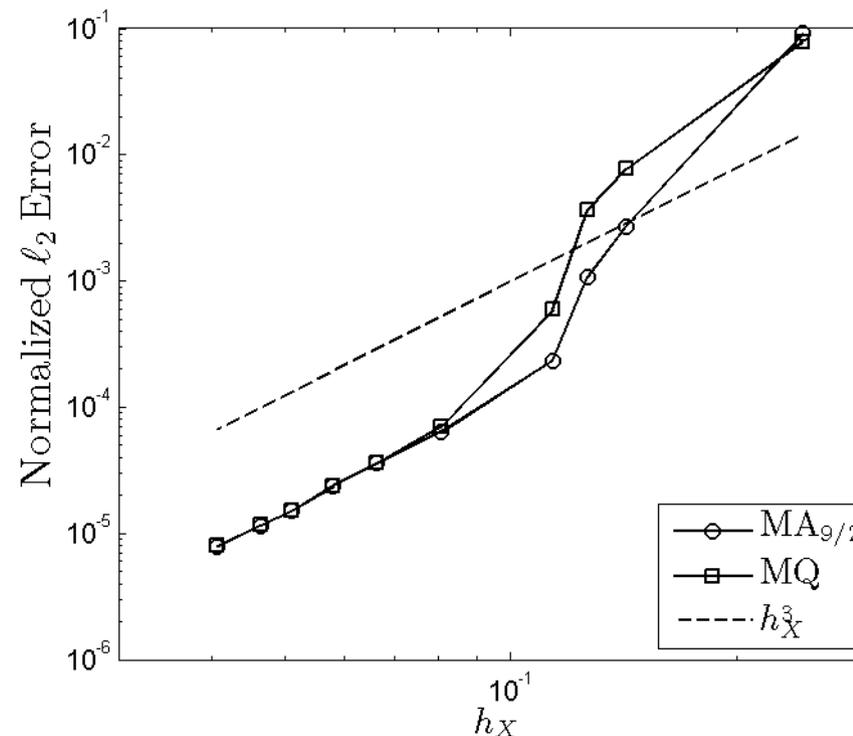
- Only  $\mathbf{u}$  is sampled;  $\mathbf{u}_{\text{div}}$  and  $\mathbf{u}_{\text{curl}}$  are not known to the interpolant.

# Helmholtz-Hodge Decomposition: Example 1

- Tangent vector field  $\mathbf{u}(\mathbf{x})$  sampled at “scattered” nodes on the sphere.



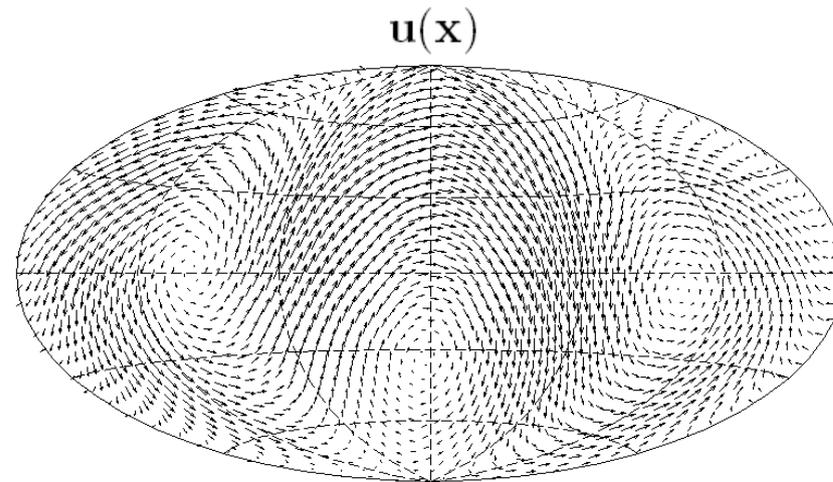
- Error in the RBF reconstructed field vs. node spacing (log-log scale):



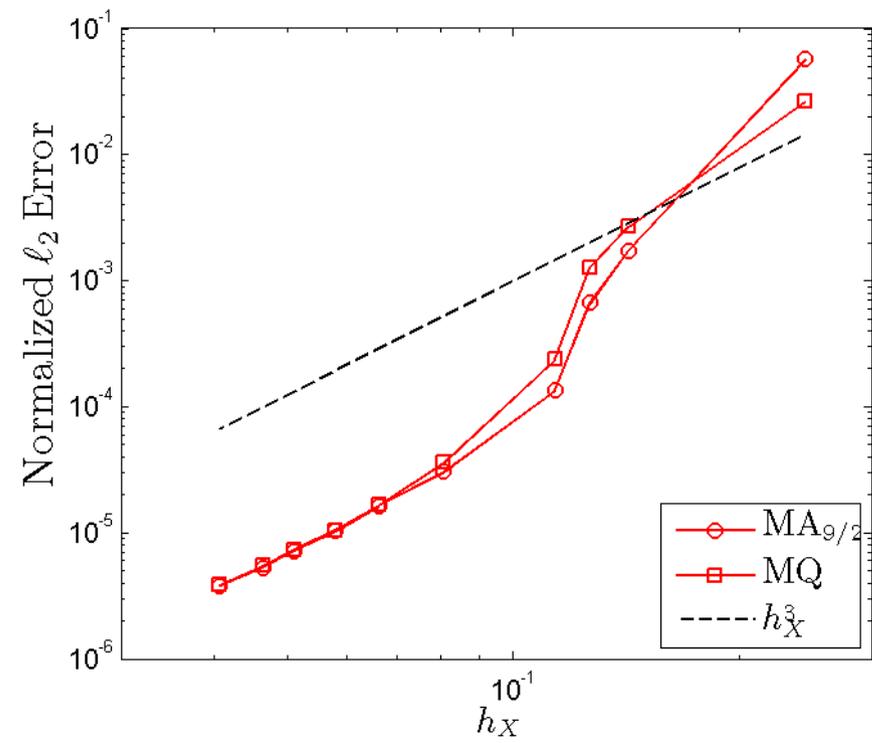
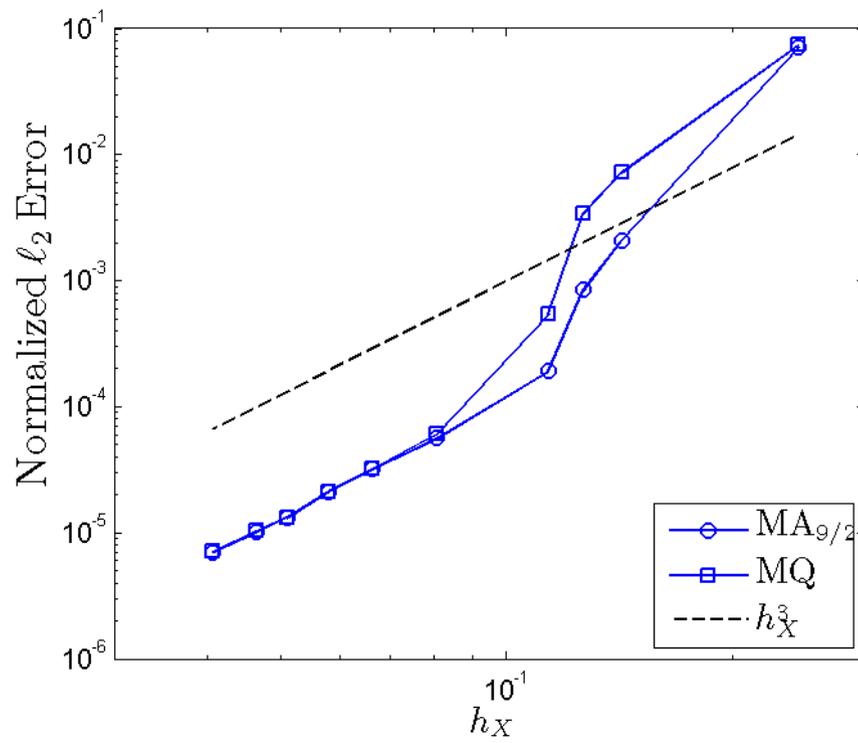
- Dashed line predicted error rate for Matérn (Fuselier and Wright 2008)

# Helmholtz-Hodge Decomposition: Example 1

- Tangent vector field  $\mathbf{u}(\mathbf{x})$  sampled at “scattered” nodes on the sphere.



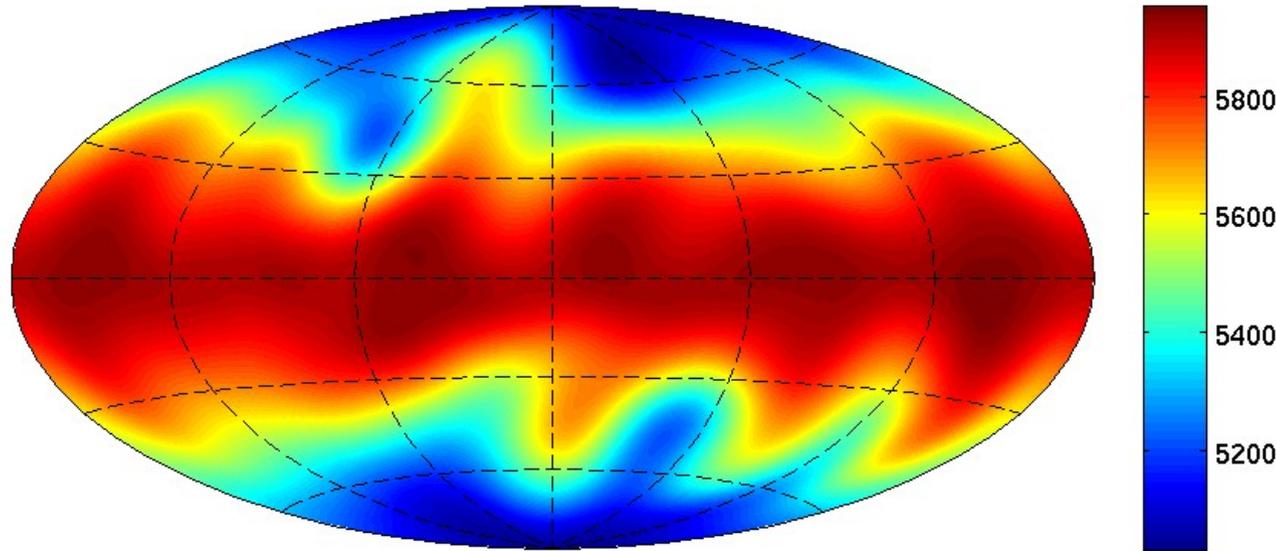
- Error in the RBF reconstruction of the surface **div-free** and **curl-free** parts vs. node spacing:



# Helmholtz-Hodge Decomposition: Example 2

- Test case 5 (flow over an isolated mountain) from Williamson *et. al.* JCP (1992).

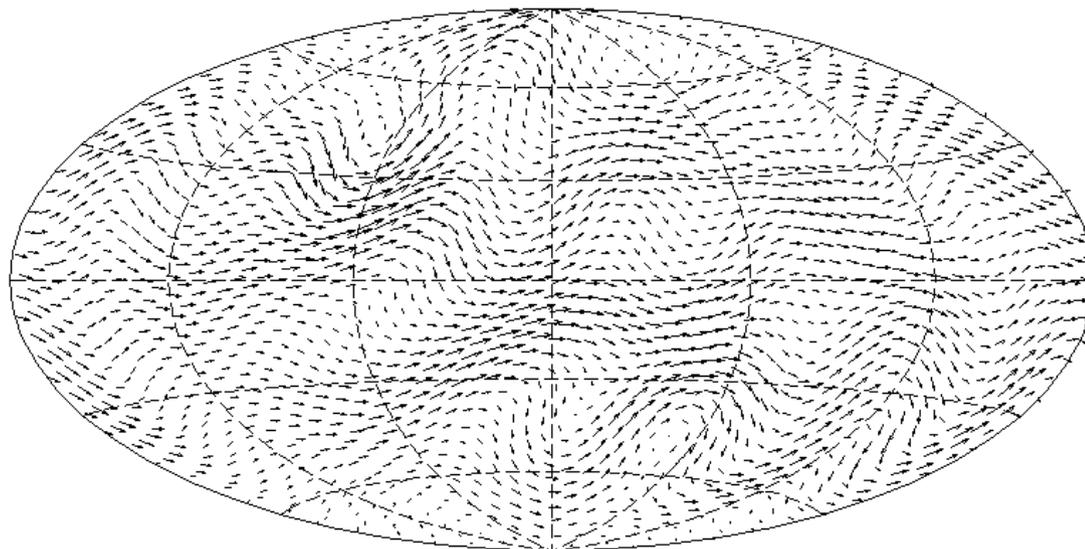
Height field  $t=15$  days



## Solution details

- GME SWM (Majewski *et. al.* MWR 2002)
- Icosahedral grid point model (92162 grid points).

Velocity field  $\mathbf{u}$   $t=15$  days



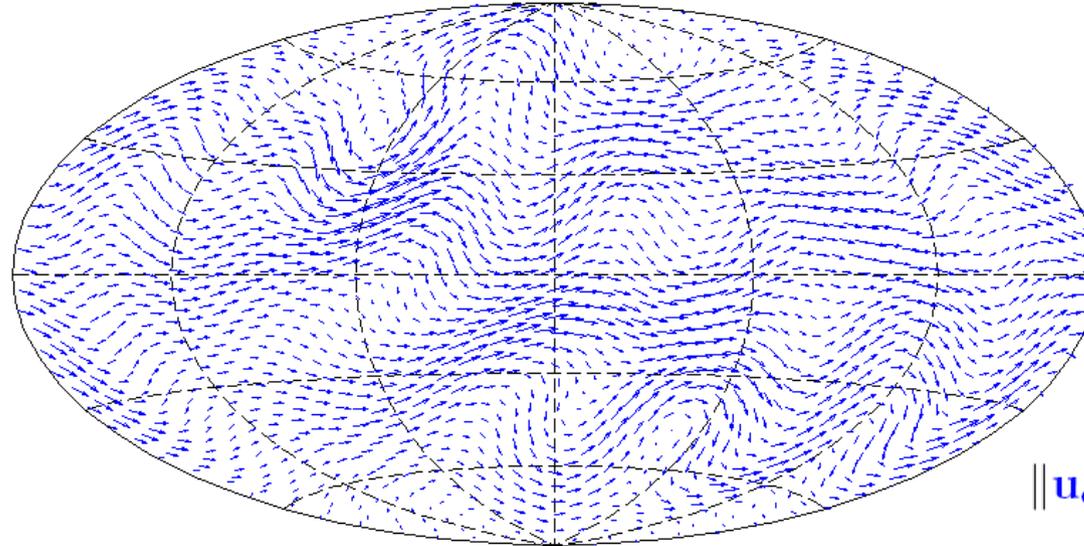
## Sample details

- Sample velocity field of GME solution at  $N=1849$  “scattered” nodes.
- Construct Helmholtz-Hodge RBF interpolant using Matérn  $MA_{9/2}$  RBF

# Helmholtz-Hodge Decomposition: Example 2

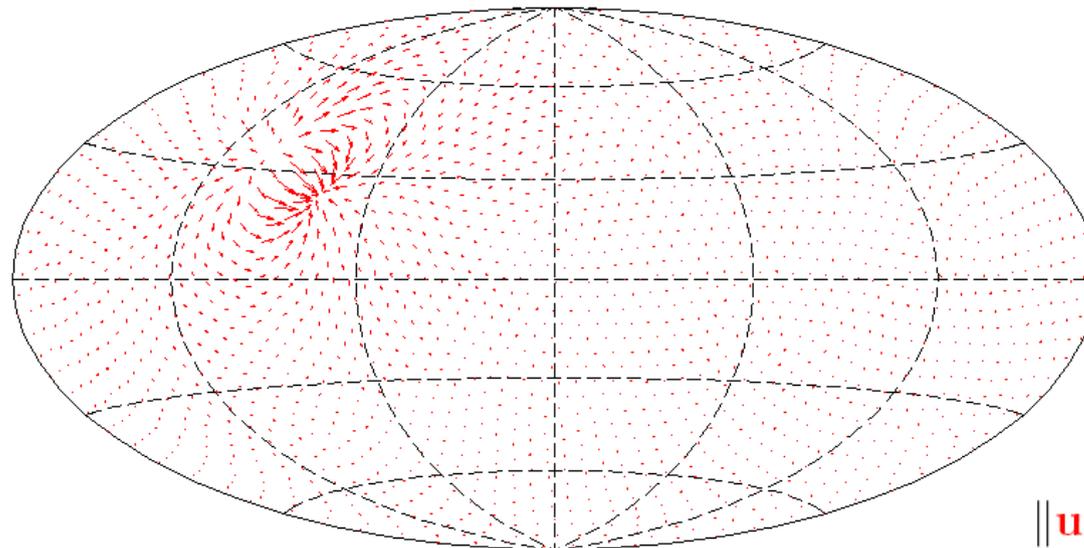
- Test case 5 (flow over an isolated mountain) from Williamson *et. al.* JCP (1992).

RBF reconstructed **div-free** velocity field  $t=15$  days



$$\|\mathbf{u}_{\text{div}}\|_2 = 40.3 \text{ m/s}$$

RBF reconstructed **curl-free** velocity field  $t=15$  days

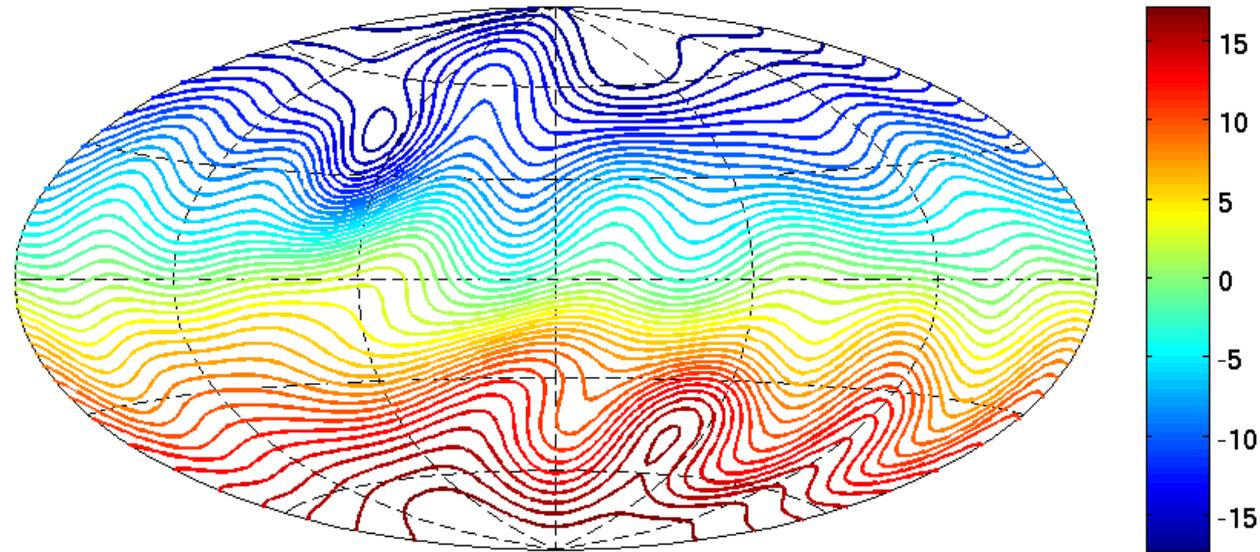


$$\|\mathbf{u}_{\text{curl}}\|_2 = 4.3 \text{ m/s}$$

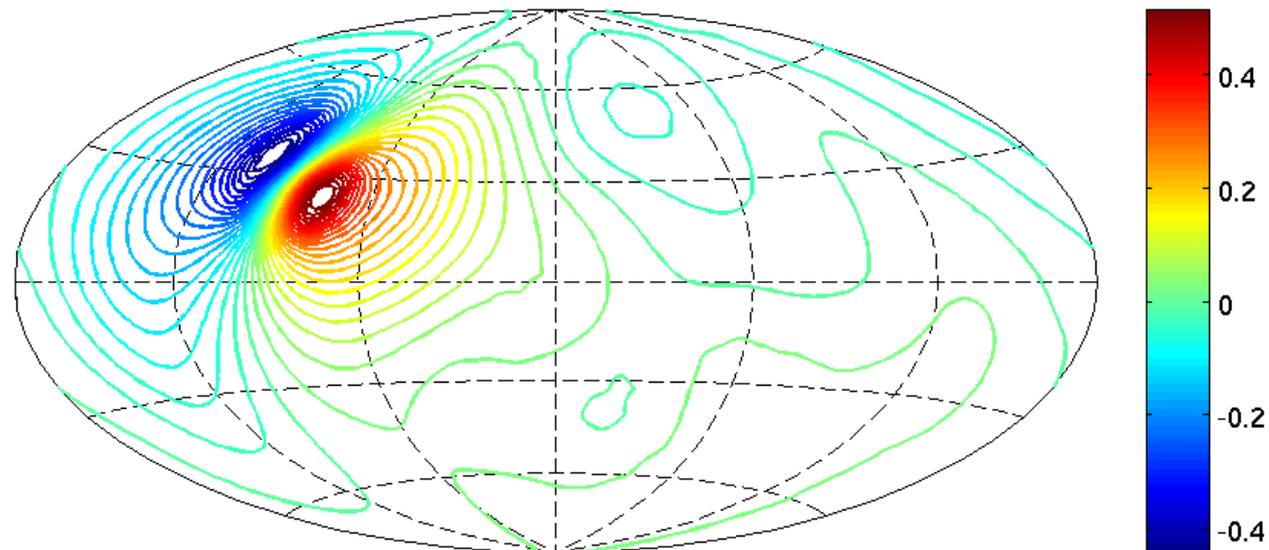
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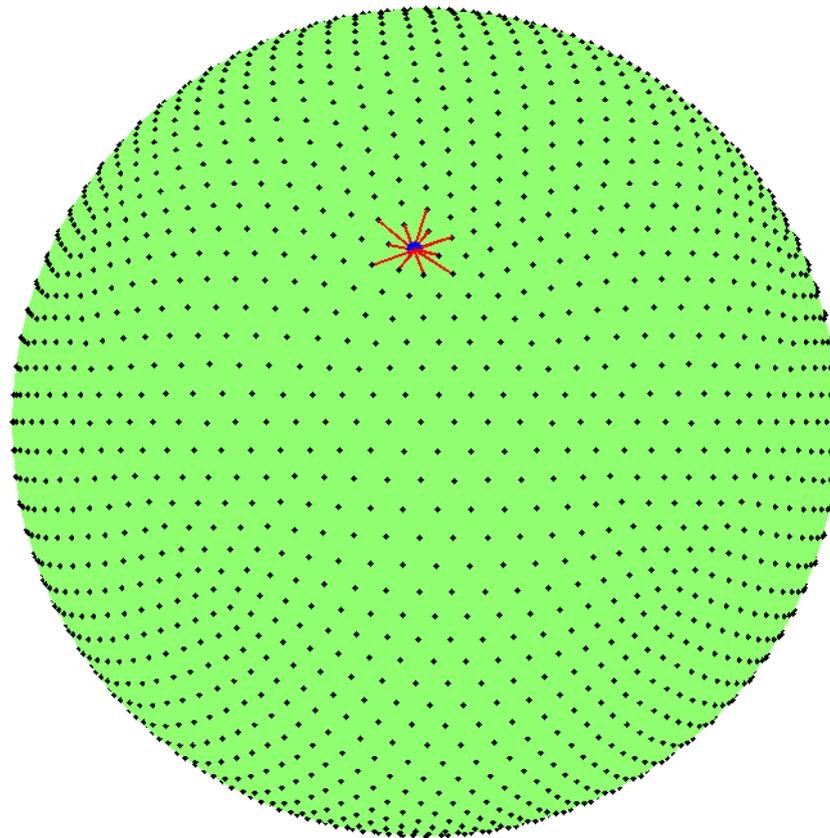
RBF reconstructed stream function  $t=15$  days



RBF reconstructed velocity potential  $t=15$  days



- This technique can also be used on patches of the sphere to locally approximate:
  - 1) Surface **divergence-free** or **curl-free** vector fields
  - 2) General vector fields tangent to the sphere
  - 3) Stream functions and velocity potentials
  - 4) Vorticity
  - 5) **Divergence**
- Illustration of a local approximation

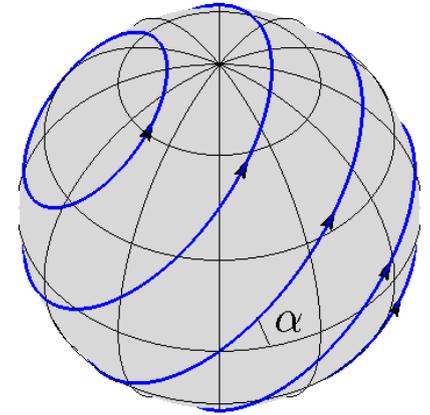


# Example: linear advection

- Governing PDE to simulate:  $\frac{\partial h}{\partial t} + \nabla \cdot (\mathbf{u}h) = 0$

$$\mathbf{u} = [u_s, v_s]^T = u_0[(\cos \alpha \cos \theta + \sin \alpha \cos \lambda \sin \theta), -\sin \alpha \sin \lambda]^T$$

$$\mathbf{u} = [u_c, v_c, w_c]^T = u_0[-y \cos \alpha, x \cos \alpha + y \sin \alpha, -y \sin \alpha]^T$$

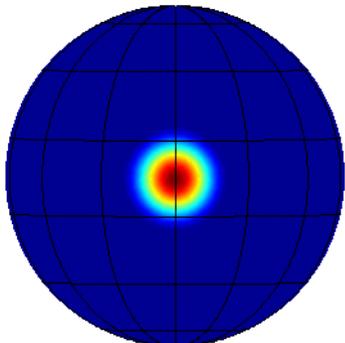


- Simulation strategy (*method of lines*):

- Rewrite advection equation as:

$$\frac{\partial h}{\partial t} = -\nabla \cdot \underbrace{(\mathbf{u}h)}_{\mathbf{f}}$$

- Approximate  $\mathbf{f}$  locally at each node point using Helmholtz-Hodge RBF interpolant.
- Compute divergence of local interpolants.
- Advance the system in time.
- **Example:** advection of a cosine bell (Test Case 1 of Williamson *et. al.* JCP (1992)).



- Icosahedral grid,  $N=40962$  nodes
- 13 point local approximation
- Standard RK4 with 30 min. time step
- MQ RBF

Simulation

- Can approximate vector fields that are either **surface divergence** or **curl-free** with RBFs.
  - For **div-free** fields, RBF reconstructed field will be **div-free**.
  - For **curl-free** fields, RBF reconstructed field will be **curl-free**.
- Can approximate general vector fields tangent to the sphere with RBFs. From the RBF reconstructed field, we can
  - Approximate the **div-free** and **curl-free** parts.
  - Generate a **stream function** and **velocity potential** for the field.
  - Approximate **vorticity** and **divergence** of the field.
- All this can also be done on **local patches** of the sphere.
  - Local technique offers nice **scaling** abilities (**petascale?**).
  - Can be used in many current geophysical models as a local interpolation method for vector fields.
  - Can also possibly be used as the basis functions for new computational models.
- Approximating “real” data.