

2013 Dolomites Research Week on Approximation



Lecture 2:

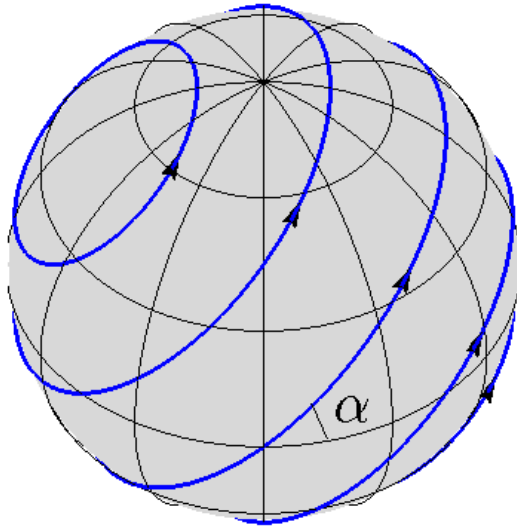
Reconstruction and decomposition of vector fields on the sphere with applications

Grady B. Wright

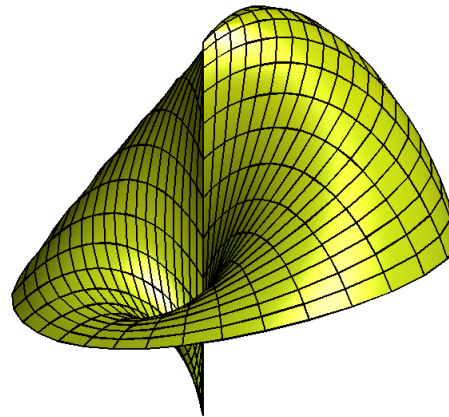
Boise State University

- Consider tangent vector field for solid body rotation:

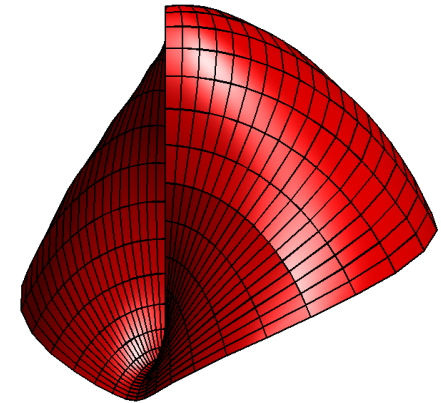
$$\mathbf{u} = [u_s, v_s]^T = u_0[(\cos \alpha \cos \theta + \sin \alpha \cos \lambda \sin \theta), -\sin \alpha \sin \lambda]^T$$



$$r(\lambda, \theta) = 1 + u_s(\lambda, \theta)$$



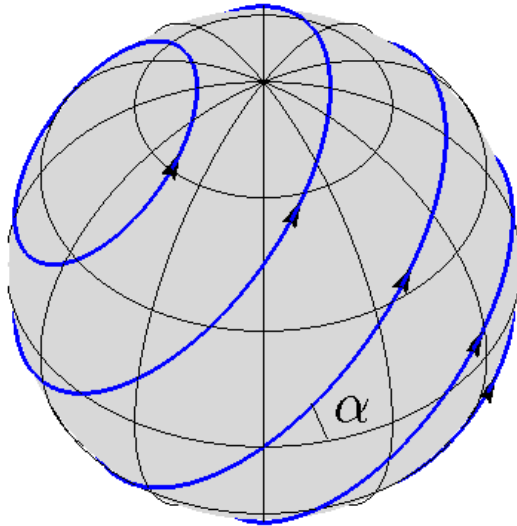
$$r(\lambda, \theta) = 1 + v_s(\lambda, \theta)$$



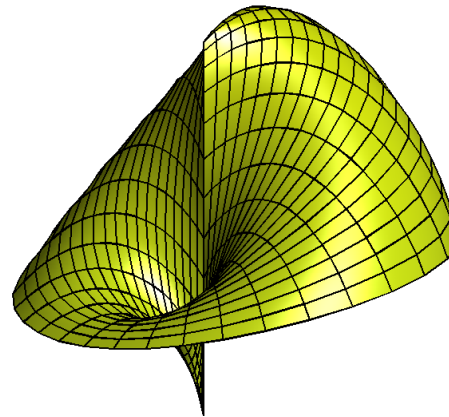
Components of \mathbf{u} are discontinuous at poles!

- Consider tangent vector field for solid body rotation:

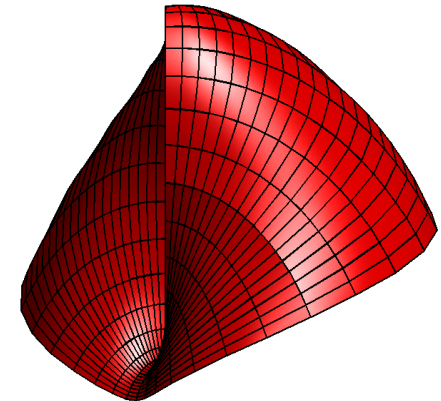
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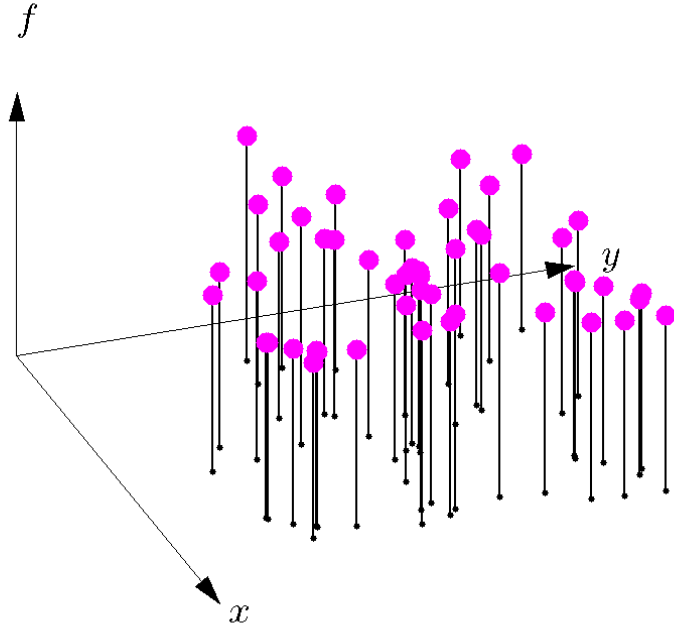


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Components of \mathbf{u} are discontinuous at poles!

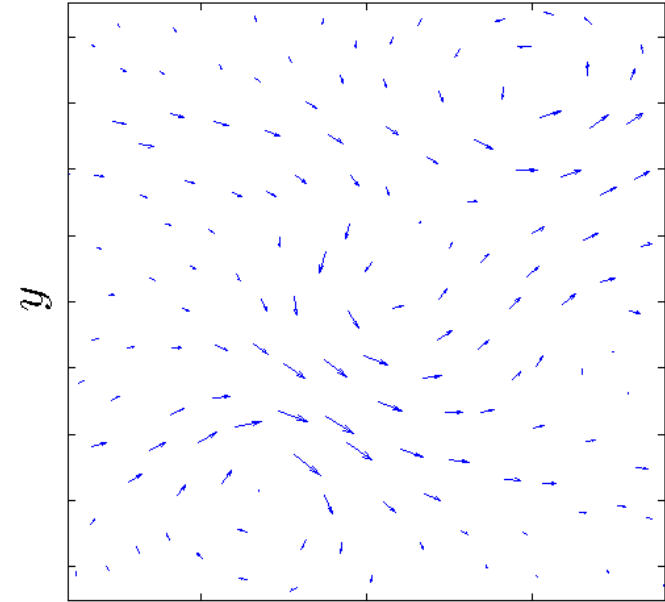
- Purpose of this tutorial:** Describe a reconstruction method for tangent vector fields that:
 - Is entirely free of any coordinate singularities.
 - Can accurately reconstruct a vector field from *scattered samples* of the field.
 - Analytically preserves certain physical properties of the field.
 - Can be differentiated to accurately compute the divergence and vorticity of the field.
- The construction is based on *positive definite matrix valued kernels*



Scalar interpolant:

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

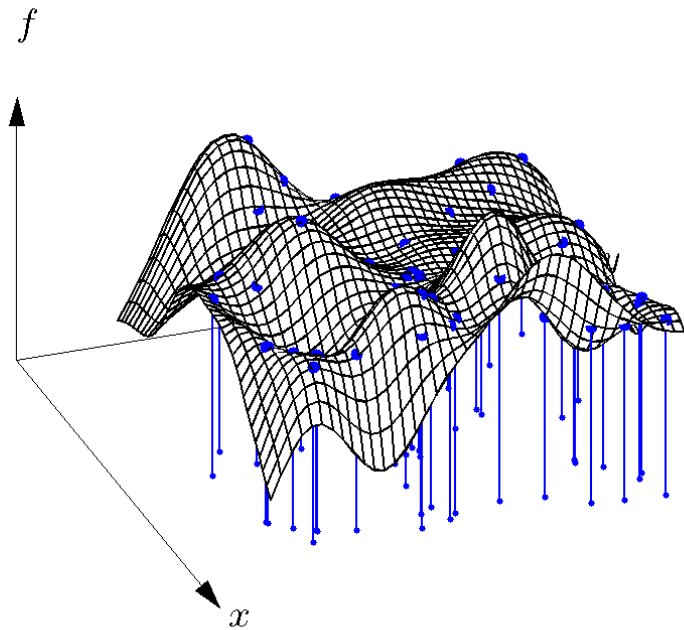
- ϕ is scalar-valued “radial” kernel.
- Nodes can be “scattered”.
- Interpolation matrix is positive definite.
- Can impose additional constraints on g .
- Form of the interpolant does not depend on the topology of domain.



Vector interpolant:

$$\mathbf{s}(\mathbf{x}) = \sum_{j=1}^N \Phi(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j, \quad \mathbf{s}(\mathbf{x}_k) = \mathbf{u}_k, \quad k = 1, \dots, N$$

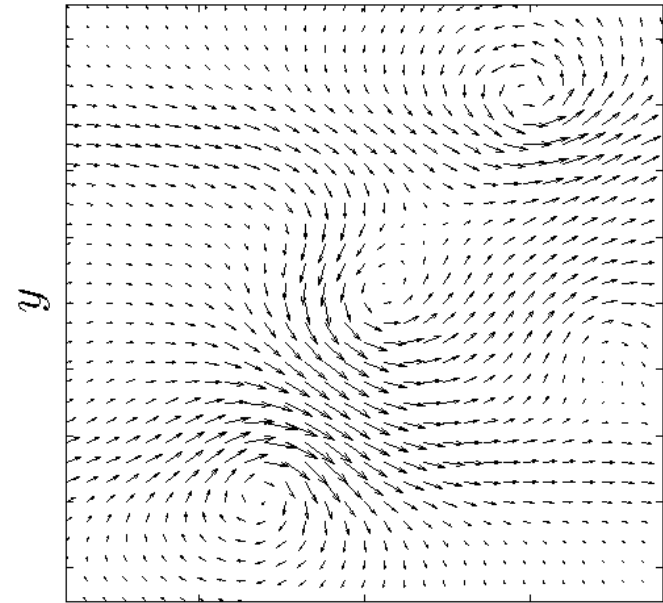
- Φ is *matrix-valued* kernel based on scalar ϕ .
- Nodes can be “scattered”.
- Interpolation matrix is positive definite.
- Can impose additional constraints on \mathbf{s} .
- Form of the interpolant does depend on the topology of domain.



Scalar interpolant:

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

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Vector interpolant:

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- Customizing RBF approximations for vector fields:

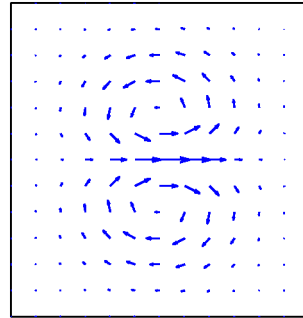
$$\mathbf{s}(\mathbf{x}) = \sum_{j=1}^N \Phi(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j, \quad \mathbf{s}(\mathbf{x}_k) = \mathbf{u}_k, \quad k = 1, \dots, N$$

- Developed by [Narcowich and Ward \(1994\)](#)

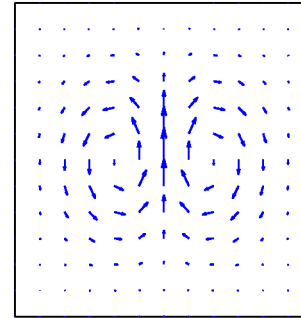
F. J. Narcowich and J. D. Ward. Generalized Hermite interpolation via matrix-valued conditionally positive definite functions. *Math. Comp.*, 63:661-687, 1994.

- **Divergence-free** vector fields: fluid flows, (static) magnetic fields

Columns of **div-free**
matrix-valued kernel:



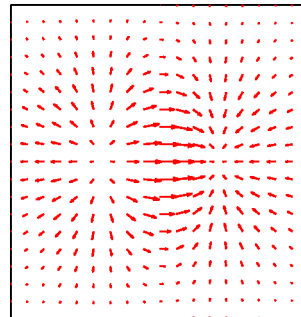
$$\Phi_{\text{div}} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



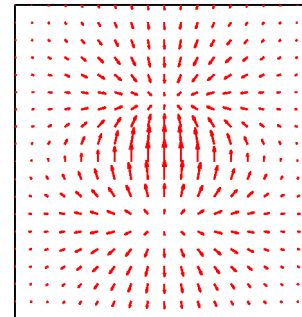
$$\Phi_{\text{div}} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

- **Curl-free** vector fields: gravity fields, (static) electric fields

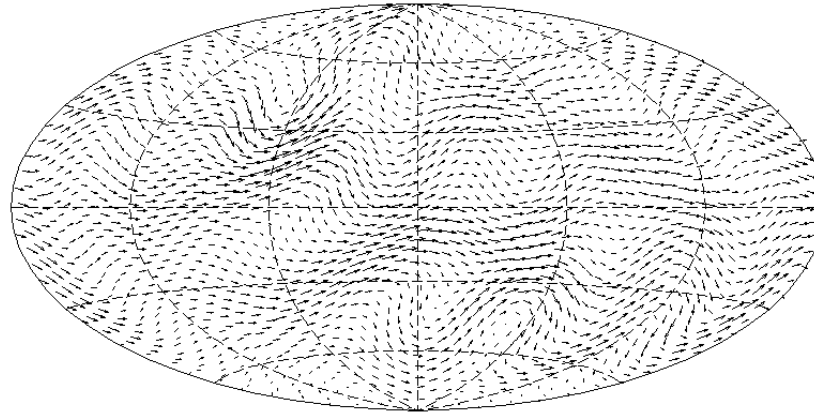
Columns of **curl-free**
matrix-valued kernel:



$$\Phi_{\text{curl}} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



$$\Phi_{\text{curl}} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$



- Kernel approximation of vector fields tangent to the surface of the sphere:
 - Surface divergence-free approximation
 - Surface curl-free approximation
 - Helmholtz-Hodge decomposition
 - F.J. Narcowich, J.D. Ward, and G.B.W. Divergence-free RBFs on Surfaces. *J. Fourier Anal. Appl.* 13 (2007), 643-663.
 - E.J. Fuselier, F.J. Narcowich, J.D. Ward, and G.B.W. Error and stability estimates for surface-divergence free RBF interpolants on the sphere. *Math. Comp.*, 78 (2009), 2157-2186.
 - E.J. Fuselier and G.B.W. Stability and error estimates for vector field interpolation and decomposition on the sphere with RBFs. *SIAM J. Numer. Anal.*, 47 (2009), 3213-3239.
- Geophysical applications

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

$$(P_{\mathbf{x}} \nabla_c) \cdot \mathbf{u}_c = \nabla_c \cdot \mathbf{u}_c - \mathbf{x} \cdot \nabla(\mathbf{u}_c \cdot \mathbf{x})$$

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

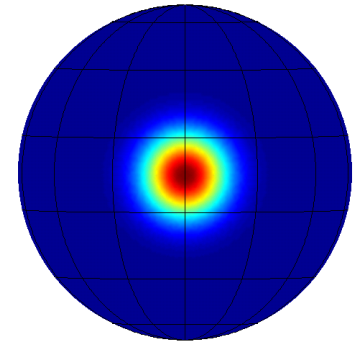
$$\mathbf{u}_c = \mathbf{x} \times (P_{\mathbf{x}} \nabla_c f) = Q_{\mathbf{x}} P_{\mathbf{x}} (\nabla_c f) = Q_{\mathbf{x}} (\nabla_c f)$$

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

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

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

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



$$\begin{aligned}\Psi_{\text{div}}(\mathbf{x}, \mathbf{y}) &= (Q_{\mathbf{x}} \nabla_{\mathbf{x}})(Q_{\mathbf{y}} \nabla_{\mathbf{y}})^T \phi(\|\mathbf{x} - \mathbf{y}\|) \\ &= Q_{\mathbf{x}}(\nabla \nabla^T \phi(\|\mathbf{x} - \mathbf{y}\|))Q_{\mathbf{y}}\end{aligned}$$

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

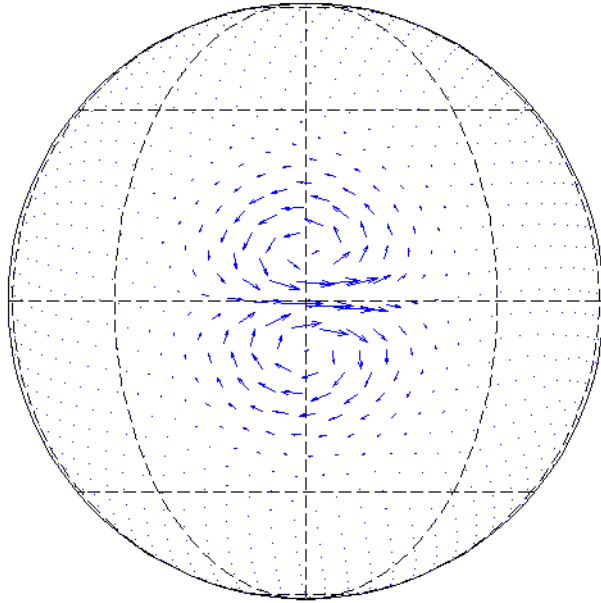
$$\begin{aligned}\Psi_{\text{div}}(\mathbf{x}, \mathbf{y})\mathbf{c} &= [Q_{\mathbf{x}}(\nabla \nabla^T \phi(\|\mathbf{x} - \mathbf{y}\|))Q_{\mathbf{y}}] \mathbf{c} \\ &= Q_{\mathbf{x}} \nabla [\nabla^T (\phi(\|\mathbf{x} - \mathbf{y}\|)Q_{\mathbf{y}}\mathbf{c})] \\ &= Q_{\mathbf{x}}(\nabla f).\end{aligned}$$

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

- Idea can be extended to other smooth orientable manifolds.

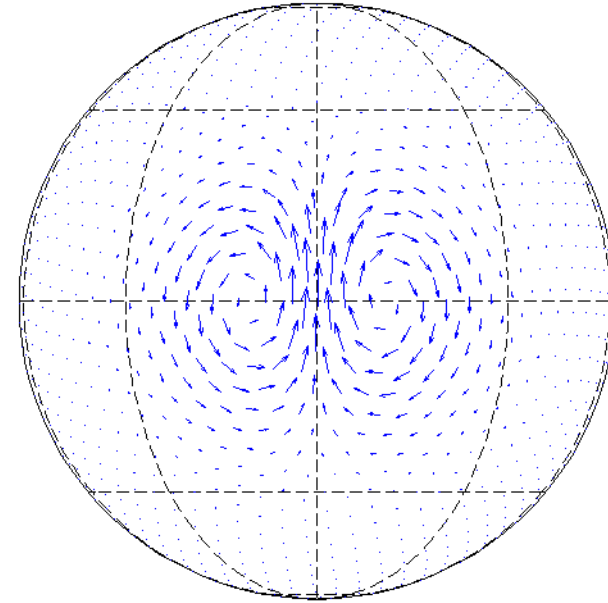
- Illustration of new basis (orthographic projection):

Zonal basis

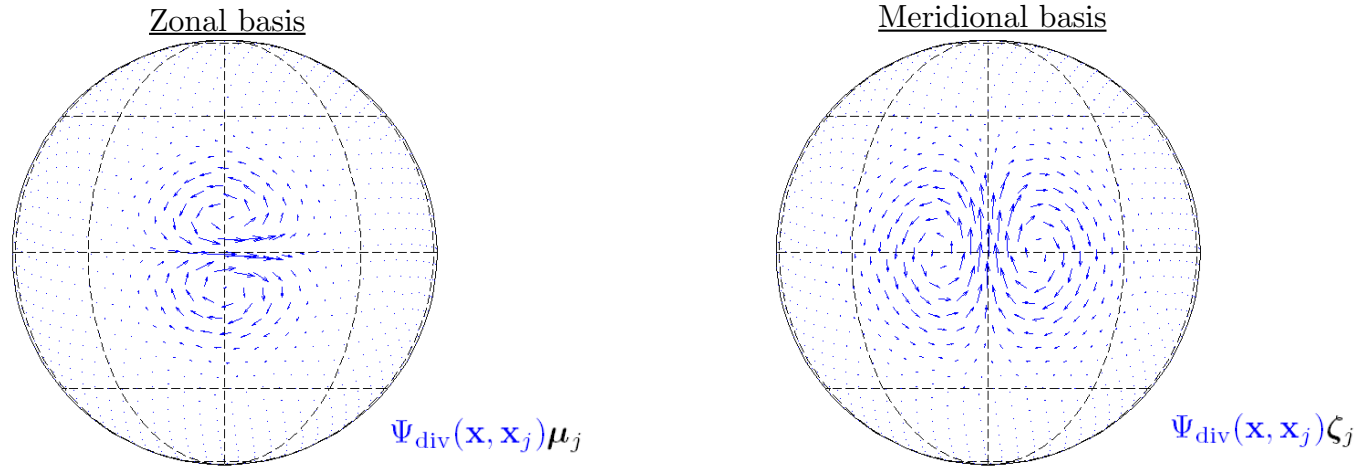


$$\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) \boldsymbol{\mu}_j$$

Meridional basis



$$\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) \boldsymbol{\zeta}_j$$



Procedure:

1. For each distinct node \mathbf{x}_j , center a $\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j)\boldsymbol{\mu}_j$ and $\Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j)\boldsymbol{\zeta}_j$.
2. Linearly combine these *vector-valued* functions to satisfy the interpolation conditions.

$$\mathbf{s}(\mathbf{x}) = \sum_{j=1}^N \Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) \underbrace{[a_j \boldsymbol{\mu}_j + b_j \boldsymbol{\zeta}_j]}_{\mathbf{c}_j}$$

3. Solve $2N$ -by- $2N$ linear system for unknown coefficients.

Interpolant will

- exist and be unique (linear system positive definite),
- be free of any pole singularity,
- be tangent to the sphere,
- be surface divergence-free.

- Strategy: Show the $\Psi_{\text{div}}(\mathbf{x}, \mathbf{y})$ is **positive definite** on \mathbb{S}^2 , i.e.:

$$\sum_{j,k} \mathbf{c}_k^T \Psi_{\text{div}}(\mathbf{x}_k, \mathbf{x}_j) \mathbf{c}_j > 0$$

for all distinct finite point sets $X = \{\mathbf{x}_j\}_{j=1}^N \subset \mathbb{S}^2$ and all \mathbf{c}_j tangent to \mathbb{S}^2 at \mathbf{x}_j .

- Key: All positive definite radial kernels are zonal on \mathbb{S}^2 with the spectral representation:

$$\phi(\|\mathbf{x} - \mathbf{y}\|) = \psi(\mathbf{x} \cdot \mathbf{y}) = \sum_{\ell=0}^{\infty} \hat{\psi}(\ell) \sum_{m=-\ell}^{\ell} Y_{\ell}^m(\mathbf{x}) Y_{\ell}^m(\mathbf{y}) \quad (\text{with } \hat{\psi}(\ell) > 0)$$

- Thus we can write $\Psi_{\text{div}}(\mathbf{x}, \mathbf{y}) := (Q_{\mathbf{x}} \nabla_{\mathbf{x}})(Q_{\mathbf{y}} \nabla_{\mathbf{y}})^T \phi(\|\mathbf{x} - \mathbf{y}\|)$ as

$$\begin{aligned} \Psi_{\text{div}}(\mathbf{x}, \mathbf{y}) &= (Q_{\mathbf{x}} \nabla_{\mathbf{x}})(Q_{\mathbf{y}} \nabla_{\mathbf{y}})^T \left(\sum_{\ell=0}^{\infty} \hat{\psi}(\ell) \sum_{m=-\ell}^{\ell} Y_{\ell}^m(\mathbf{x}) Y_{\ell}^m(\mathbf{y}) \right) \\ &= \sum_{\ell=1}^{\infty} \hat{\psi}(\ell) \sum_{m=-\ell}^{\ell} [(Q_{\mathbf{x}} \nabla_{\mathbf{x}}) Y_{\ell}^m(\mathbf{x})] [(Q_{\mathbf{y}} \nabla_{\mathbf{y}})^T Y_{\ell}^m(\mathbf{y})] = \sum_{\ell=1}^{\infty} \ell(\ell+1) \hat{\psi}(\ell) \sum_{m=-\ell}^{\ell} \mathbf{Y}_{\ell}^m(\mathbf{x}) [\mathbf{Y}_{\ell}^m(\mathbf{y})]^T \end{aligned}$$

- Using this result we immediately obtain:

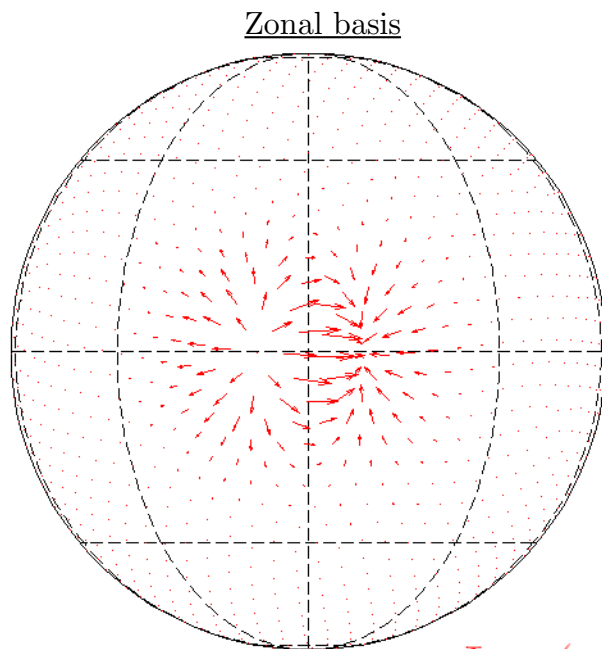
$$\begin{aligned} \sum_{j,k=1}^N \mathbf{c}_k^T \Psi_{\text{div}}(\mathbf{x}_k, \mathbf{x}_j) \mathbf{c}_j &= \sum_{j,k} \sum_{\ell=1}^{\infty} \ell(\ell+1) \hat{\psi}(\ell) \sum_{m=-\ell}^{\ell} \mathbf{c}_k^T \mathbf{Y}_{\ell}^m(\mathbf{x}_k) [\mathbf{Y}_{\ell}^m(\mathbf{x}_j)]^T \mathbf{c}_j \\ &= \sum_{\ell=1}^{\infty} \ell(\ell+1) \hat{\psi}(\ell) \sum_{m=-\ell}^{\ell} \left| \sum_{k=1}^N \mathbf{c}_k^T \mathbf{Y}_{\ell}^m(\mathbf{x}_k) \right|^2 \geq 0. \end{aligned}$$

- **Surface curl-free** basis:

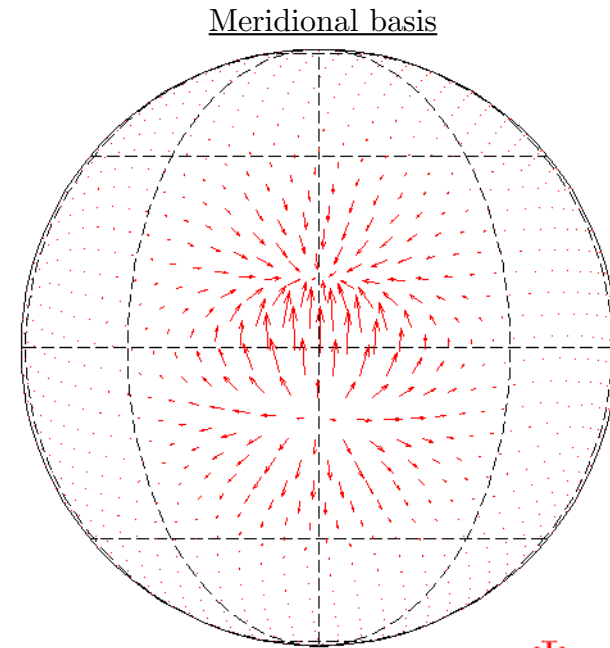
- Use extrinsic (Cartesian) coordinates, $\mathbf{x} = (x, y, z) \in \mathbb{S}^2$.
- $\Psi_{\text{curl}}(\mathbf{x}, \mathbf{y}) = (P_{\mathbf{x}} \nabla_{\mathbf{x}})(P_{\mathbf{y}} \nabla_{\mathbf{y}}^T) \phi(\|\mathbf{x} - \mathbf{y}\|) = -P_{\mathbf{x}}(\nabla \nabla^T \phi(\|\mathbf{x} - \mathbf{y}\|))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

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

- Illustration of new basis (orthographic projection):



$$\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j) \boldsymbol{\mu}_j$$



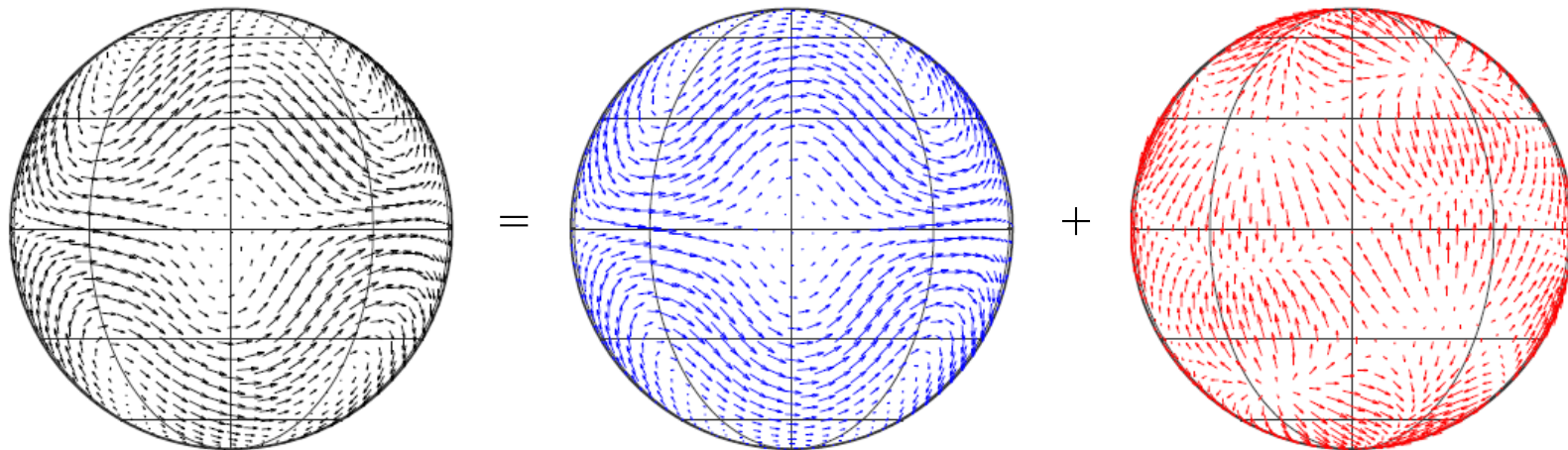
$$\Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j) \boldsymbol{\zeta}_j$$

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

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

ψ = stream function and χ = velocity potential

- Example:



- **Goal**: Construct an interpolant that mimics the Helmholtz-Hodge decomposition.

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

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

ψ = stream function and χ = velocity potential

- **RBF interpolant mimicking the Helmholtz-Hodge decomposition**: $\mathbf{x} = (x, y, z) \in \mathbb{S}^2$

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

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

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

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

ψ = stream function and χ = velocity potential

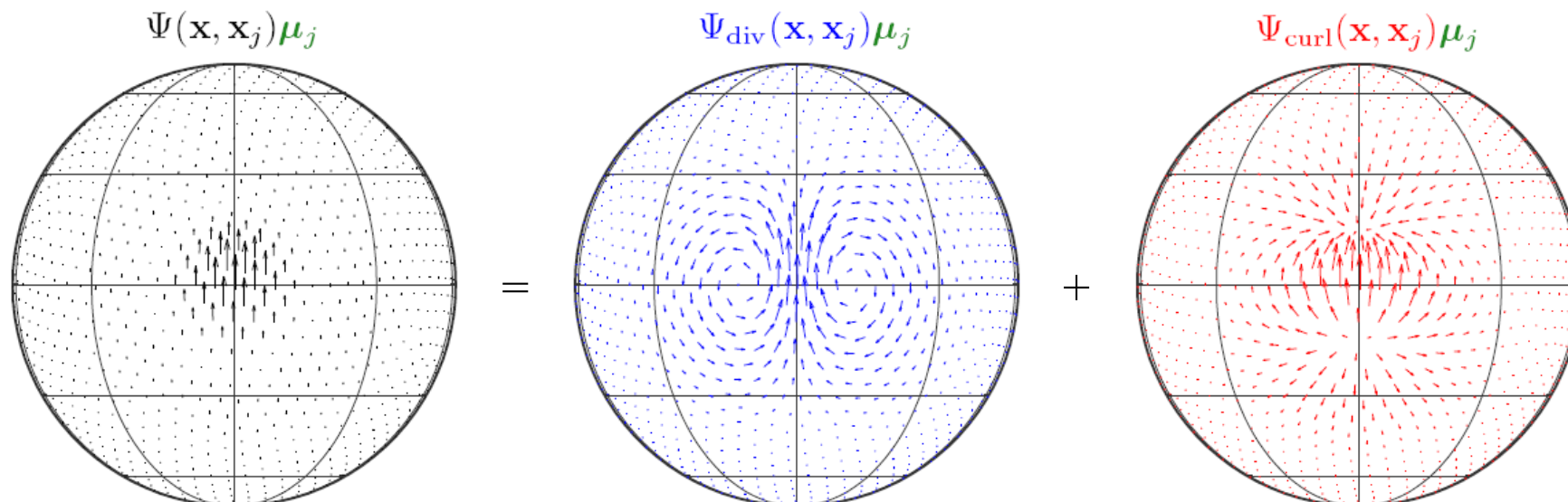
- **RBF interpolant mimicking the Helmholtz-Hodge decomposition**: $\mathbf{x} = (x, y, z) \in \mathbb{S}^2$

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

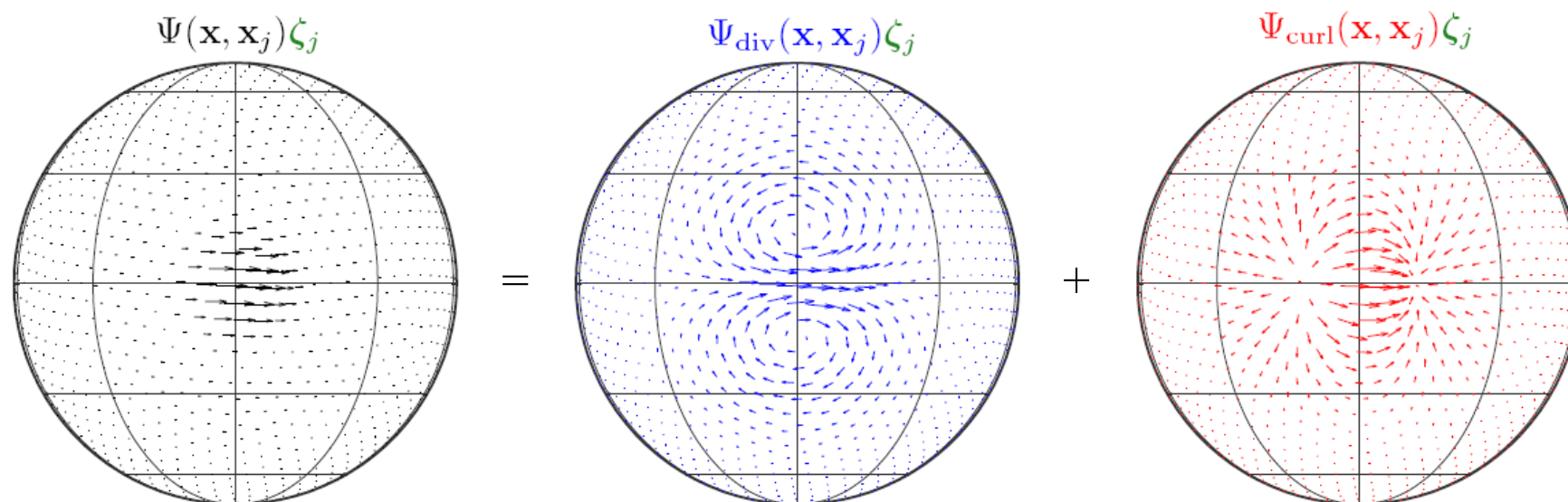
- Can get a stream function and velocity potential for the interpolant!

$$\text{Interpolant: } s(\mathbf{x}) = \sum_{j=1}^N \Psi(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j = \sum_{j=1}^N \Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j + \sum_{j=1}^N \Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j, \text{ where } \mathbf{c}_j = a_j \boldsymbol{\mu}_j + b_j \boldsymbol{\zeta}_j$$

Meridional



Zonal



$$\text{Interpolant: } \mathbf{s}(\mathbf{x}) = \sum_{j=1}^N \Psi(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j = \sum_{j=1}^N \Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j + \sum_{j=1}^N \Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j, \text{ where } \mathbf{c}_j = a_j \boldsymbol{\mu}_j + b_j \boldsymbol{\zeta}_j$$

- Construction of the interpolant identical to the **div-free** construction.
- Can show for any distinct node set, **existence and uniqueness are guaranteed**.
- Error estimates for reconstructing and decomposing vector field $\mathbf{u} = \mathbf{u}_{\text{div}} + \mathbf{u}_{\text{curl}}$:

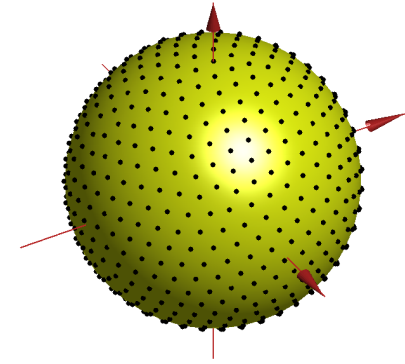
Key quantities:

$$\text{Separation radius: } q_X := \frac{1}{2} \min_{i \neq j} d(\mathbf{x}_i, \mathbf{x}_j)$$

$$\text{Mesh norm: } h_X := \sup_{\mathbf{x} \in \mathbb{S}^2} \min_{\mathbf{x}_j \in X} d(\mathbf{x}, \mathbf{x}_j)$$

$$\text{Mesh ratio: } \rho_X = h_X / q_X$$

$$\text{Decay Fourier transform } \phi : \hat{\phi} \sim (1 + \|\boldsymbol{\xi}\|_2^2)^{-(\tau + \frac{3}{2})}$$



Theorem 1 (Fuselier and Wright 2009):

If $\mathbf{u} \in H^\beta(\mathbb{S}^2)$, with $\beta \geq \tau > 1$ then we have

$$\begin{aligned} \|\mathbf{u} - \mathbf{s}\|_{L^2(\mathbb{S}^2)} &\leq Ch_X^\tau \|\mathbf{u}\|_{H^\tau(\mathbb{S}^2)}, \\ \|\mathbf{u}_{\text{div}} - \mathbf{s}_{\text{div}}\|_{L^2(\mathbb{S}^2)} &\leq Ch_X^\tau \|\mathbf{u}\|_{H^\tau(\mathbb{S}^2)}, \\ \|\mathbf{u}_{\text{curl}} - \mathbf{s}_{\text{curl}}\|_{L^2(\mathbb{S}^2)} &\leq Ch_X^\tau \|\mathbf{u}\|_{H^\tau(\mathbb{S}^2)}. \end{aligned}$$

Interpolant: $\mathbf{s}(\mathbf{x}) = \sum_{j=1}^N \Psi(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j = \sum_{j=1}^N \Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j + \sum_{j=1}^N \Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j$, where $\mathbf{c}_j = a_j \boldsymbol{\mu}_j + b_j \boldsymbol{\zeta}_j$

- Construction of the interpolant identical to the **div-free** construction.
- Can show for any distinct node set, **existence and uniqueness are guaranteed**.
- Error estimates for reconstructing and decomposing vector field $\mathbf{u} = \mathbf{u}_{\text{div}} + \mathbf{u}_{\text{curl}}$:

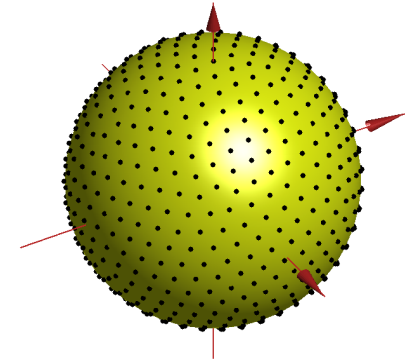
Key quantities:

$$\text{Separation radius: } q_X := \frac{1}{2} \min_{i \neq j} d(\mathbf{x}_i, \mathbf{x}_j)$$

$$\text{Mesh norm: } h_X := \sup_{\mathbf{x} \in \mathbb{S}^2} \min_{\mathbf{x}_j \in X} d(\mathbf{x}, \mathbf{x}_j)$$

$$\text{Mesh ratio: } \rho_X = h_X / q_X$$

$$\text{Decay Fourier transform } \phi : \hat{\phi} \sim (1 + \|\boldsymbol{\xi}\|_2^2)^{-(\tau + \frac{3}{2})}$$



Theorem 2 (Fuselier and Wright 2009):

If $\mathbf{u} \in H^{2\tau}(\mathbb{S}^2)$, with $\tau > 1$ then we have

$$\begin{aligned} \|\mathbf{u} - \mathbf{s}\|_{L^2(\mathbb{S}^2)} &\leq Ch_X^{2\tau} \|\mathbf{u}\|_{H^{2\tau}(\mathbb{S}^2)}, \\ \|\mathbf{u}_{\text{div}} - \mathbf{s}_{\text{div}}\|_{L^2(\mathbb{S}^2)} &\leq Ch_X^{2\tau} \|\mathbf{u}\|_{H^{2\tau}(\mathbb{S}^2)}, \\ \|\mathbf{u}_{\text{curl}} - \mathbf{s}_{\text{curl}}\|_{L^2(\mathbb{S}^2)} &\leq Ch_X^{2\tau} \|\mathbf{u}\|_{H^{2\tau}(\mathbb{S}^2)}. \end{aligned}$$

$$\text{Interpolant: } \mathbf{s}(\mathbf{x}) = \sum_{j=1}^N \Psi(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j = \sum_{j=1}^N \Psi_{\text{div}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j + \sum_{j=1}^N \Psi_{\text{curl}}(\mathbf{x}, \mathbf{x}_j) \mathbf{c}_j, \text{ where } \mathbf{c}_j = a_j \boldsymbol{\mu}_j + b_j \boldsymbol{\zeta}_j$$

- Construction of the interpolant identical to the **div-free** construction.
- Can show for any distinct node set, **existence and uniqueness are guaranteed**.
- Error estimates for reconstructing and decomposing vector field $\mathbf{u} = \mathbf{u}_{\text{div}} + \mathbf{u}_{\text{curl}}$:

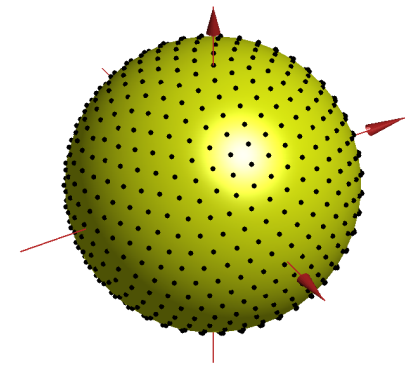
Key quantities:

$$\text{Separation radius: } q_X := \frac{1}{2} \min_{i \neq j} d(\mathbf{x}_i, \mathbf{x}_j)$$

$$\text{Mesh norm: } h_X := \sup_{\mathbf{x} \in \mathbb{S}^2} \min_{\mathbf{x}_j \in X} d(\mathbf{x}, \mathbf{x}_j)$$

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$$\text{Decay Fourier transform } \phi : \hat{\phi} \sim (1 + \|\boldsymbol{\xi}\|_2^2)^{-(\tau + \frac{3}{2})}$$



Theorem 3 (Fuselier and Wright 2009):

If $\mathbf{u} \in H^\beta(\mathbb{S}^2)$, with $1 \leq \beta \leq \tau$ then we have

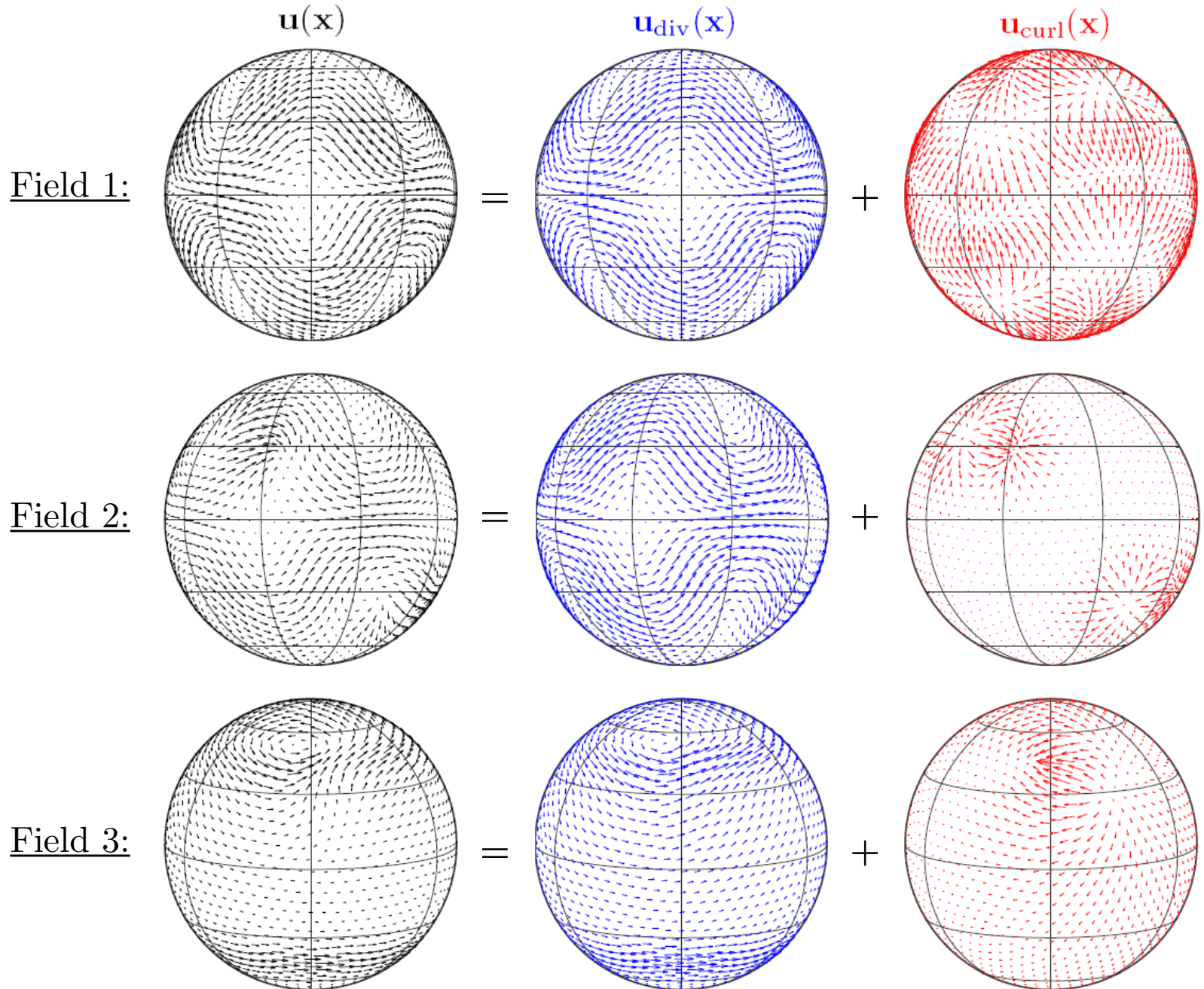
$$\|\mathbf{u} - \mathbf{s}\|_{L^2(\mathbb{S}^2)} \leq C \rho_X^{\tau - \beta} h_X^\beta \|\mathbf{u}\|_{H^\beta(\mathbb{S}^2)},$$

$$\|\mathbf{u}_{\text{div}} - \mathbf{s}_{\text{div}}\|_{L^2(\mathbb{S}^2)} \leq C \rho_X^{\tau - \beta} h_X^\beta \|\mathbf{u}\|_{H^\beta(\mathbb{S}^2)},$$

$$\|\mathbf{u}_{\text{curl}} - \mathbf{s}_{\text{curl}}\|_{L^2(\mathbb{S}^2)} \leq C \rho_X^{\tau - \beta} h_X^\beta \|\mathbf{u}\|_{H^\beta(\mathbb{S}^2)}.$$

Ex: Accuracy of reconstruction and decomposition

- Three fields of varying smoothness sampled at “scattered” nodes on the sphere:

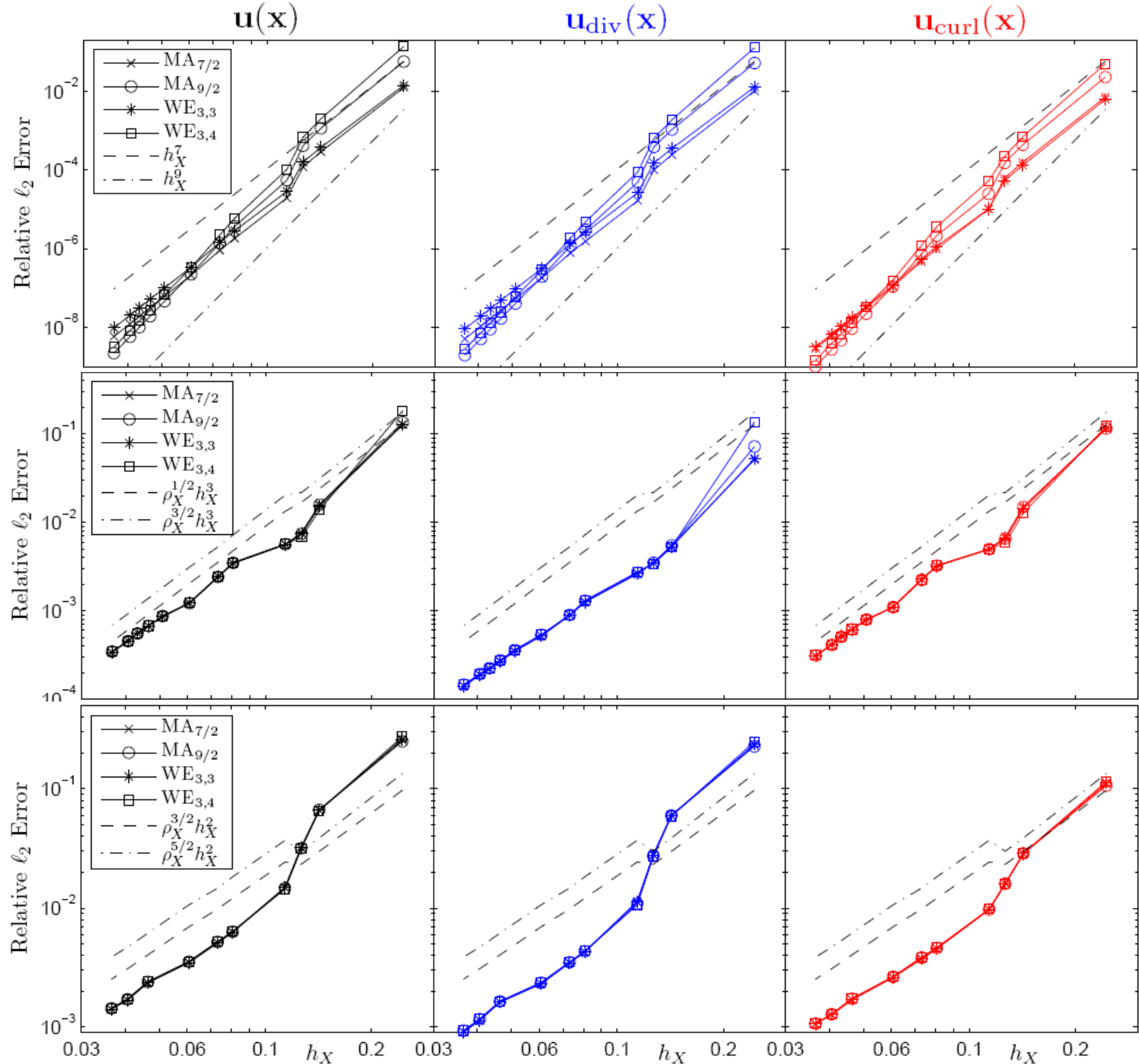


- Only the $\mathbf{u}(\mathbf{x})$ fields (first column) are sampled.

Ex: Accuracy of reconstruction and decomposition

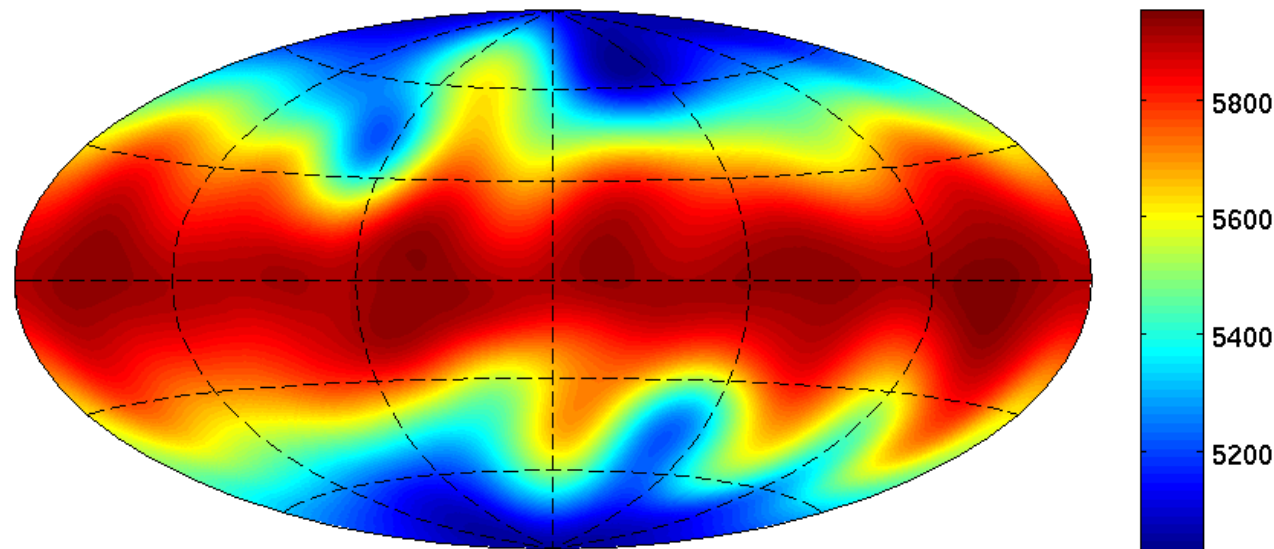
- Error in the RBF reconstructed and decomposed field vs. node spacing (log-log scale):

- Dashed lines correspond to predicted error rates from (Fuselier and Wright 2009)



- Numerical simulation of flow over an isolated mountain.

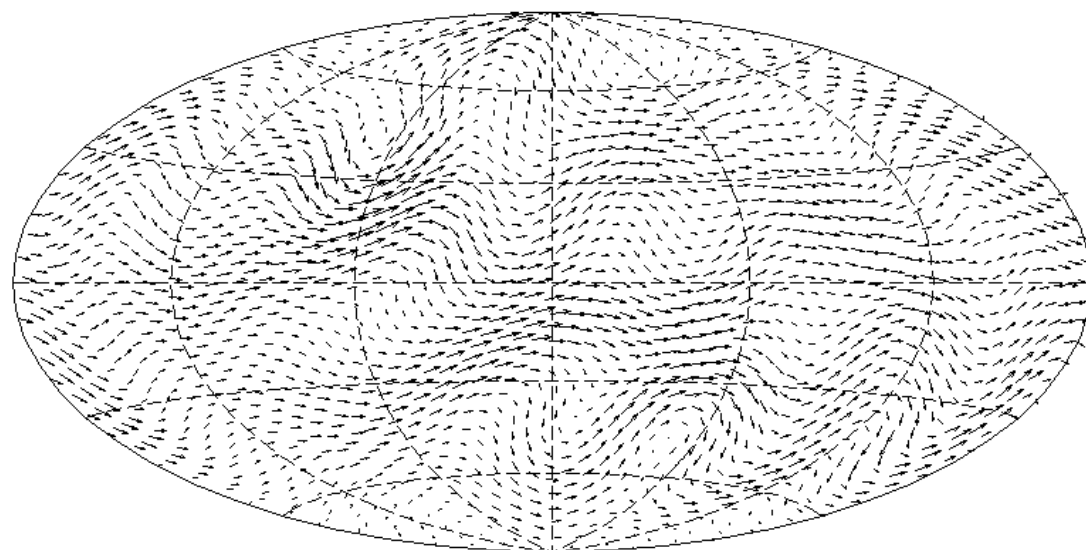
geopotential field $t=15$ days



Solution details

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

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

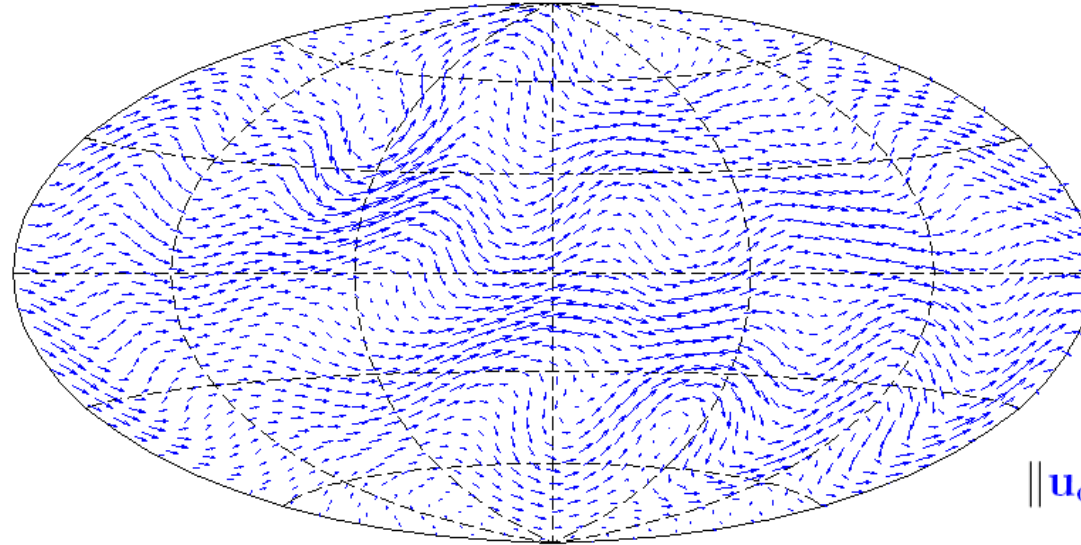


Sample details

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

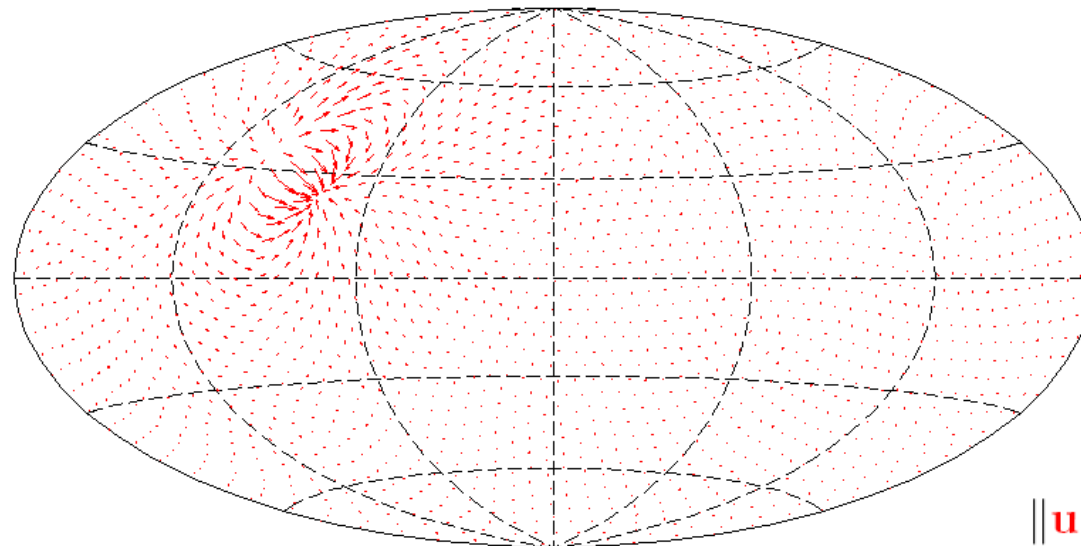
- Numerical simulation of flow over an isolated mountain.

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



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

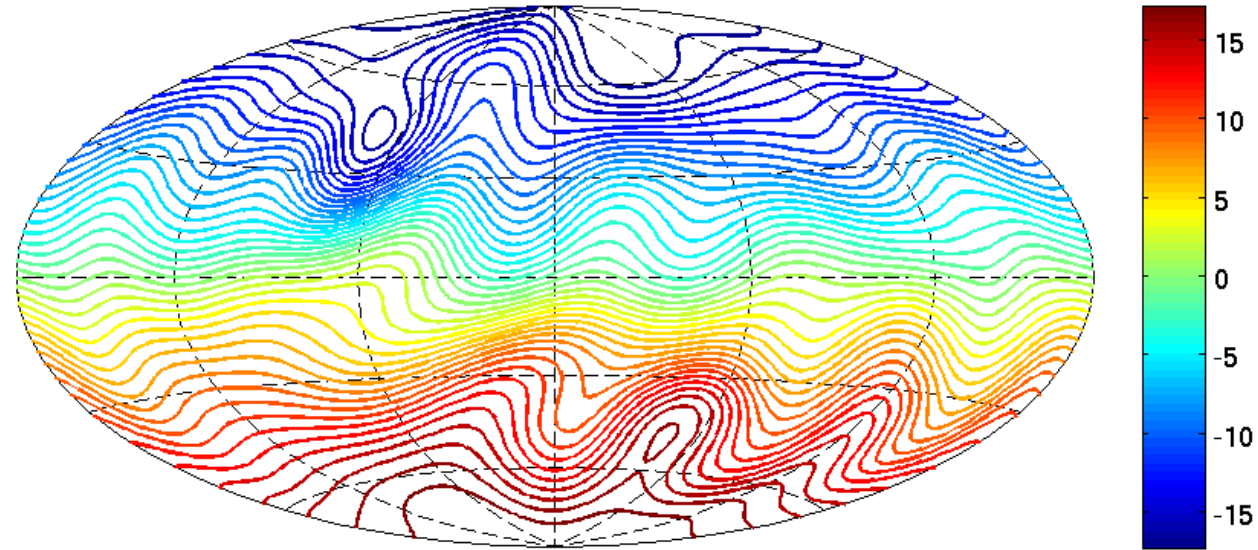
RBF reconstructed **curl-free** velocity field $t=15$ days



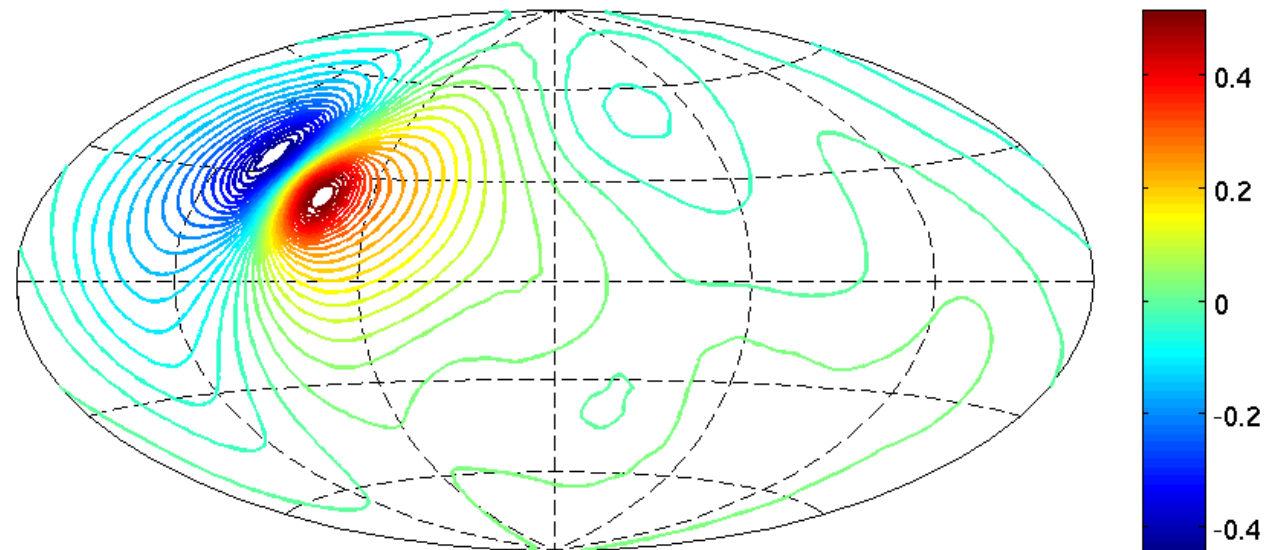
$$\|\mathbf{u}_{\text{curl}}\|_2 = 4.3 \text{ m/s}$$

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

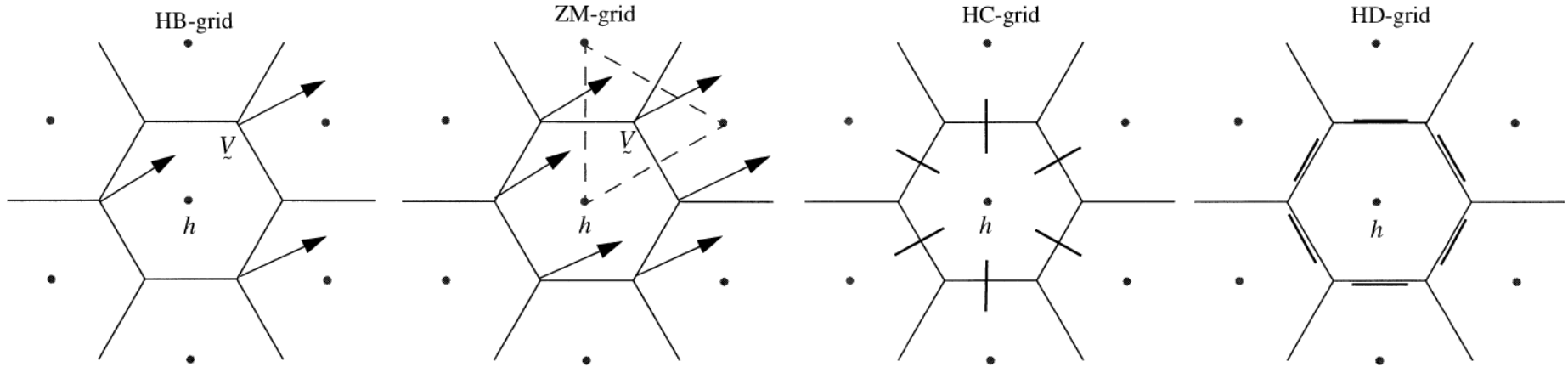
RBF reconstructed [stream function](#) $t=15$ days



RBF reconstructed [velocity potential](#) $t=15$ days

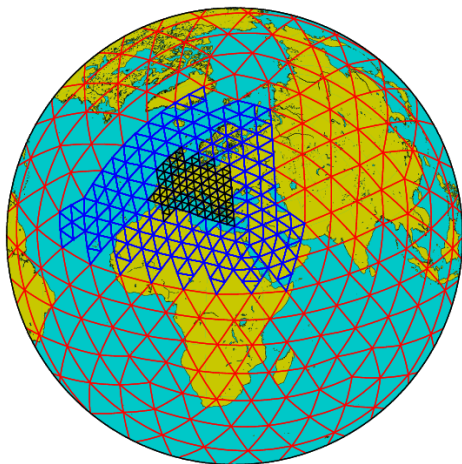


- Local reconstruction of velocity fields on staggered grids:

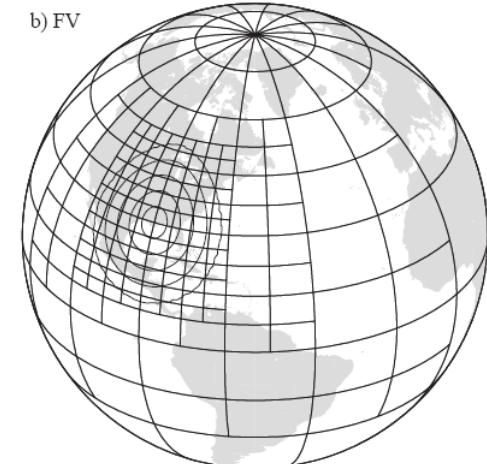
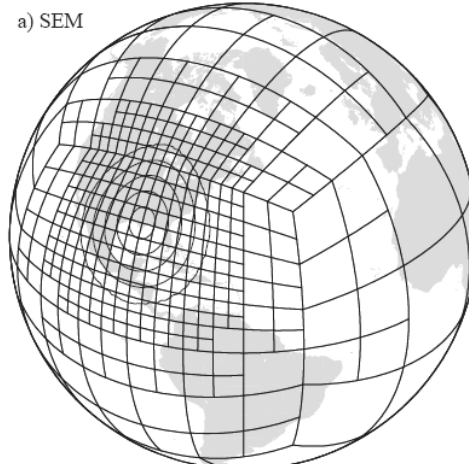


Ringler, T.D., and D.A. Randall, 2002: The ZM-grid: An alternative to the Z-grid, *Monthly Weather Review*, 130, 1411-1422.

- Local reconstruction of velocity fields for adaptive mesh refinement (AMR)



<http://icon.enes.org/>



St-Cyr, A., C. Jablonowski, J. M. Dennis, H. M. Tufo and S. J. Thomas, A Comparison of Two Shallow Water Models with Non-Conforming Adaptive Grids, *Mon. Wea. Rev.*, 136, 1898-1922, 2008.

- Show connection between surface divergence and curl free RBF interpolants and divergence and curl free spherical harmonics.
- Develop numerical method for Navier Stokes equations on a rotating sphere:

Spherical coords:

$$\frac{\partial}{\partial t} \mathbf{u}_s + \nabla_s^* p = -\nabla_{\mathbf{u}_s}^* \mathbf{u}_s + \nu \Delta_s^* \mathbf{u}_s - f \hat{\mathbf{k}} \times \mathbf{u}_s + \mathbf{g}_s,$$

$$\nabla_s^* \cdot \mathbf{u}_s = 0,$$

Cartesian coords:

$$\frac{\partial}{\partial t} \mathbf{u} + P_x \nabla p = -P_x (((P_x \nabla) \otimes \mathbf{u})^T \mathbf{u}) + \nu Q_x \nabla (\mathbf{x} \cdot (\nabla \times \mathbf{u})) - f Q_x \mathbf{u} + \mathbf{g},$$

$$\nabla \cdot \mathbf{u} = 0,$$

- Develop other decompositions (e.g. add radial component, toroidal/poloidal).
- Develop fast algorithms for computing and evaluating the interpolants.
- Extend to more general orientable manifolds:
 - Oblate/prolate spheroids and ellipsoids being the most applicable.
- Extend to bounded domains (work already under way).