

# **RBF-FD: New Computational Opportunities in the Geo-Fluid Modeling**

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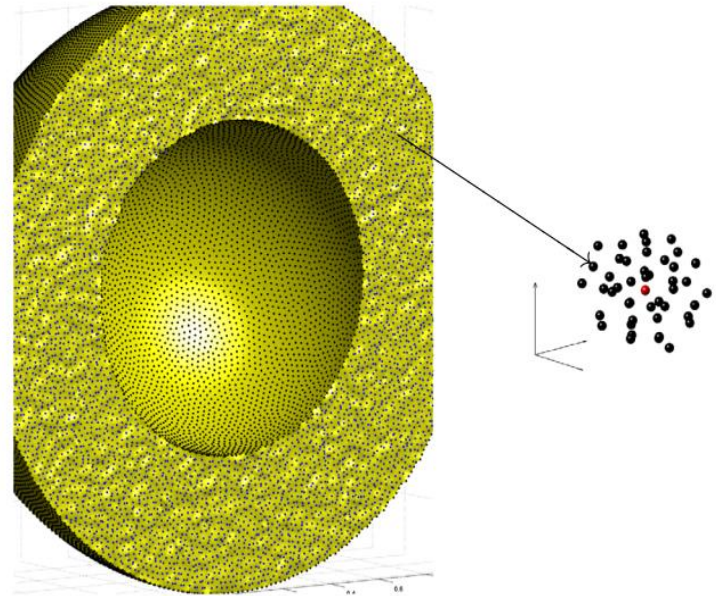
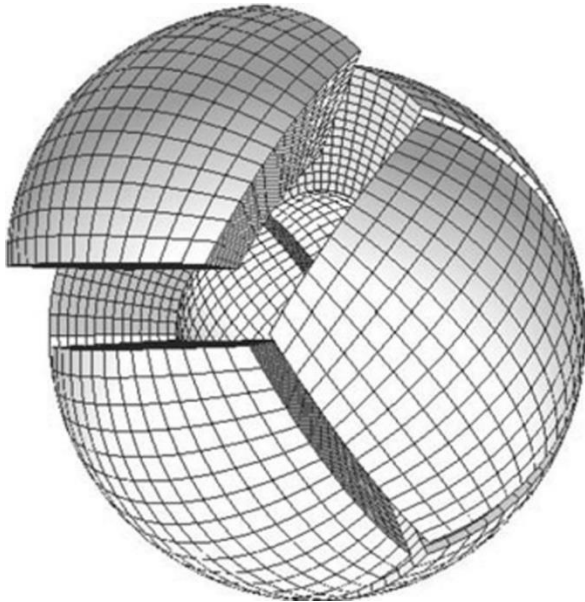
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University of Colorado-Boulder

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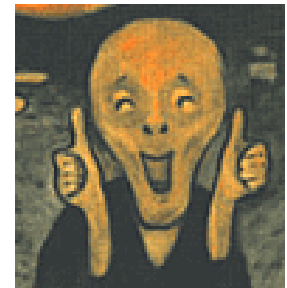
# Current vs. Future Spatial Discretization for Modeling PDEs



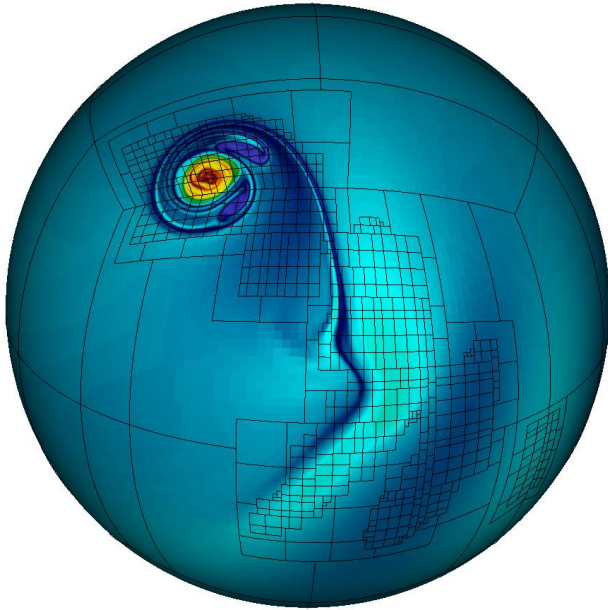
**Current**: All methods define elements or volumes. Requires mappings/transformations. Easier in 2D, computation in 3D is nightmarish.



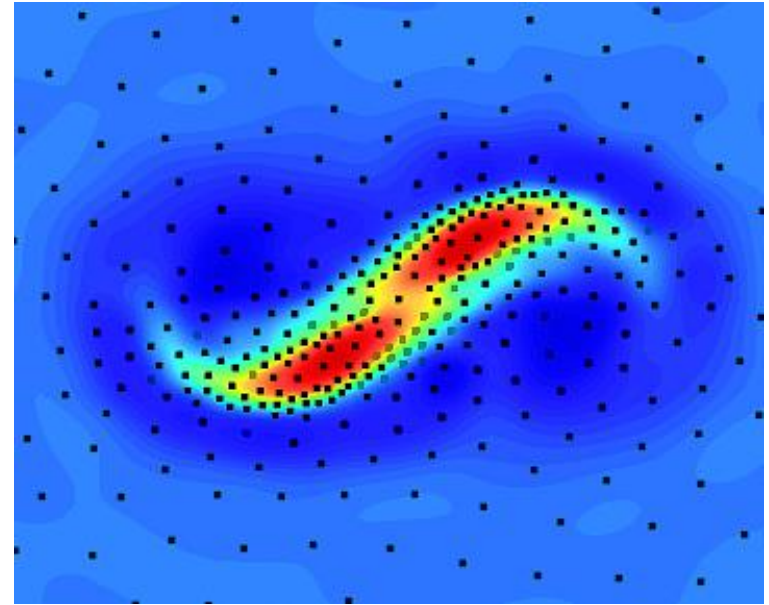
**Future**: Simply scatter nodes in any dimensional space. No connectivities, thus no mappings/transformations. To go from 2D to 3D, changing the code is much more simple.



## Current vs. Future Local Refinement



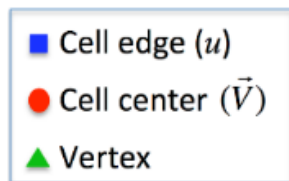
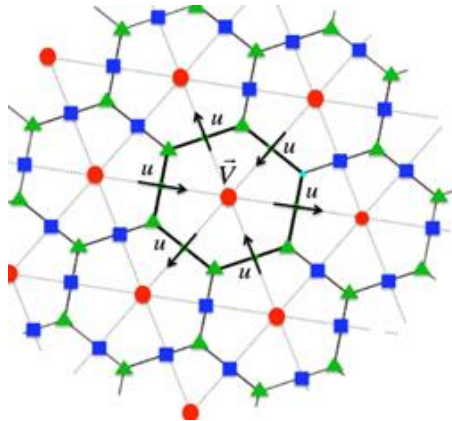
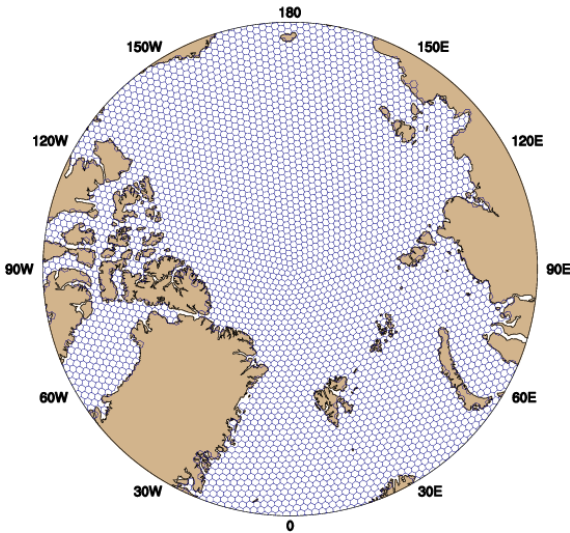
**Current**: Mesh refinement does not follow the shape of the feature, here, trying to capture a cyclone. Thus less effective in terms of accuracy and computational cost.



**Future**: Since nodes can be placed wherever needed due to no meshes, refinement occurs where most needed, here according to the gradient of the vorticity. Thus, much less pts. and computation are needed.

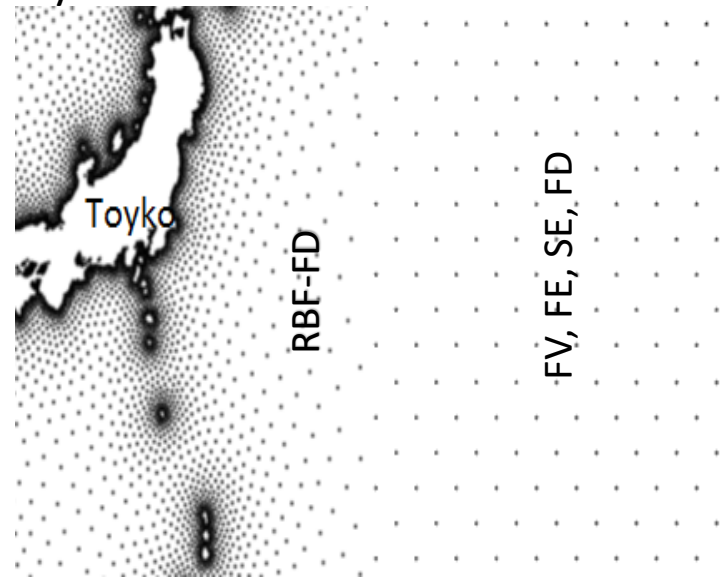
# Current vs. Future Treatment of Boundaries, etc.

MPASOcean60km.nc



**Current:** Uses Voronoi mesh can not conform to coastal topography. Need to keep track of hexagon edges, centers, and vertices.

Allows for hybridization with other numerical methods



**Future:** RBF-FD can easily conform to coastlines and only needs: 1) point locations and 2) the distances between them. Then in open ocean, one can use whatever (FV, FE, SE, FD).



# Shallow water wave equations

Simplest equations to describe the evolution of the horizontal structure of a fluid in response to forcings, such as gravity and rotation.

## Basic Properties

- Set of nonlinear hyperbolic equations derived from physical conservation laws
- Horizontal scales of motion  $\gg$  Vertical scales of motion
- Vertical velocity and all derivatives in vertical not present
- It is a 2D model.

## Areas of Application

- Atmospheric flows
- Tsunami prediction
- Planetary flows
- Storm surge
- Dam breaking

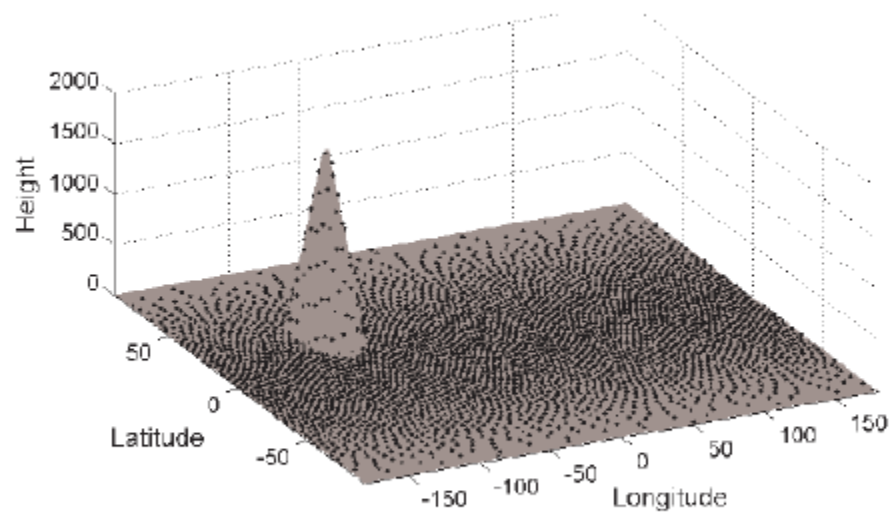


Netherlands Overflowing



Jupiter's atmosphere

# Flow over a $C^0$ Cone Mountain with RBF-FD (Flyer et al., JCP, 2012)



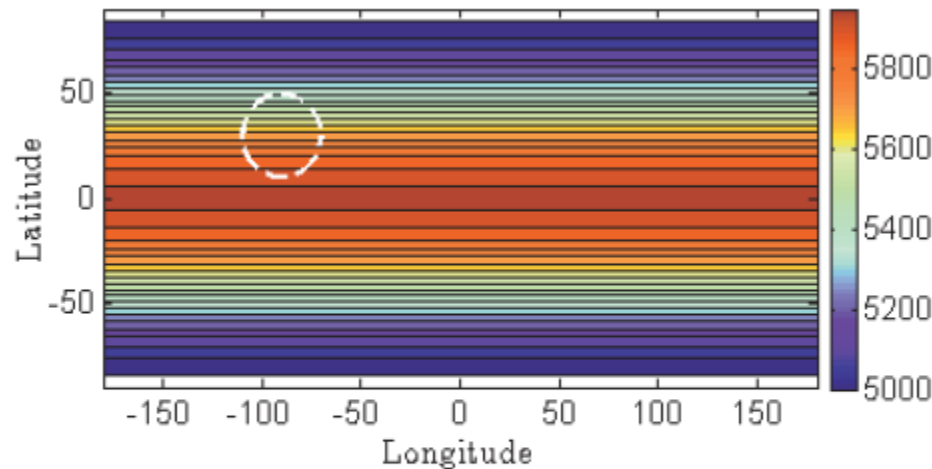
Cone Mountain

Shallow Water Equations

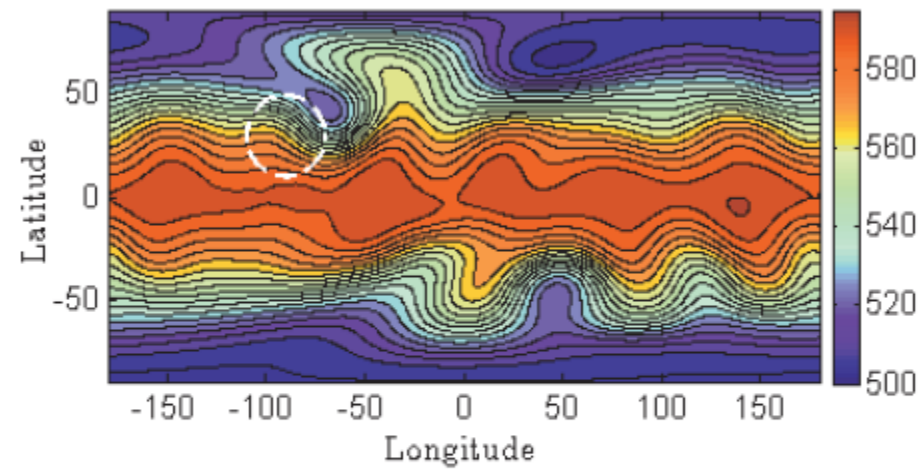
$$\frac{\partial \mathbf{u}}{\partial t} = -(\mathbf{u} \cdot \nabla) \mathbf{u} - f(\mathbf{x} \times \mathbf{u}) - g \nabla h$$

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

GA RBFs  
stencil size = 31

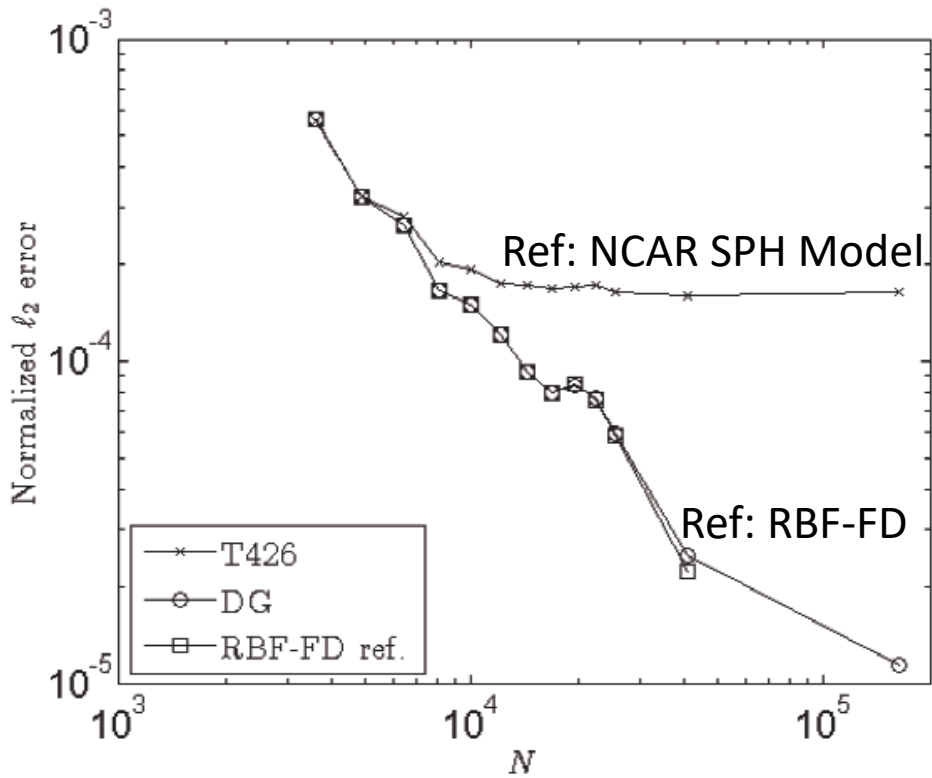


$h, t = 0$  days



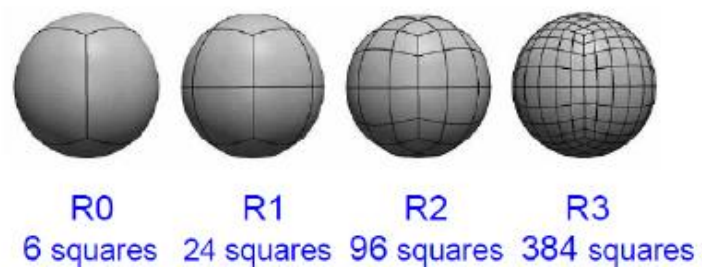
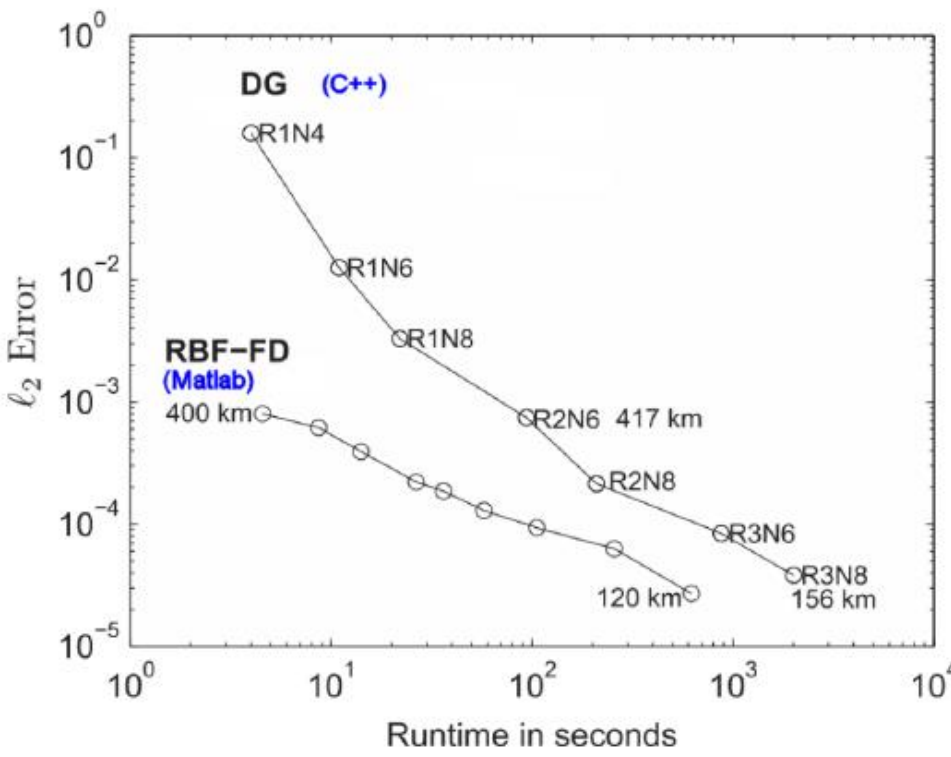
$h, t = 15$  days

# Convergence and Cost Efficiency of RBF-FD



NCAR SPH Model: 182,329 SPH bases (30km)  
RBF-FD gave first evidence that this model, the standard of comparison, was not so accurate.

Performance on Intel i7 CPU



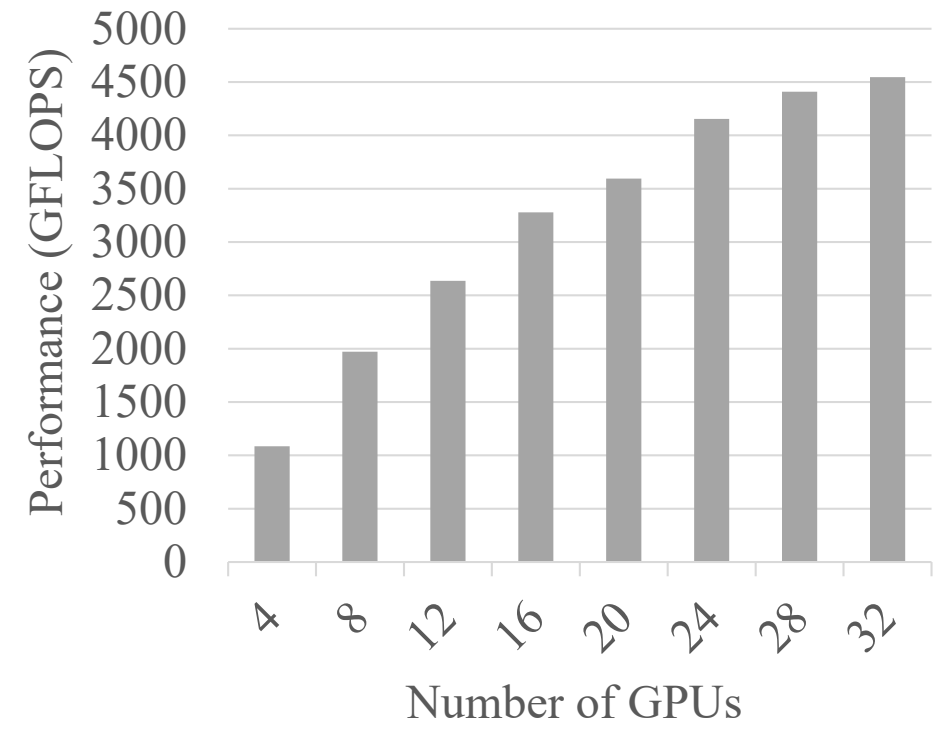
R = Number of subdivisions of each cube face  
N = Degree of Legendre poly. in each square

# Multi – CPU and Multi – GPU performance: 2.6M nodes on sphere (15km)

(Elliott et al., 2017)

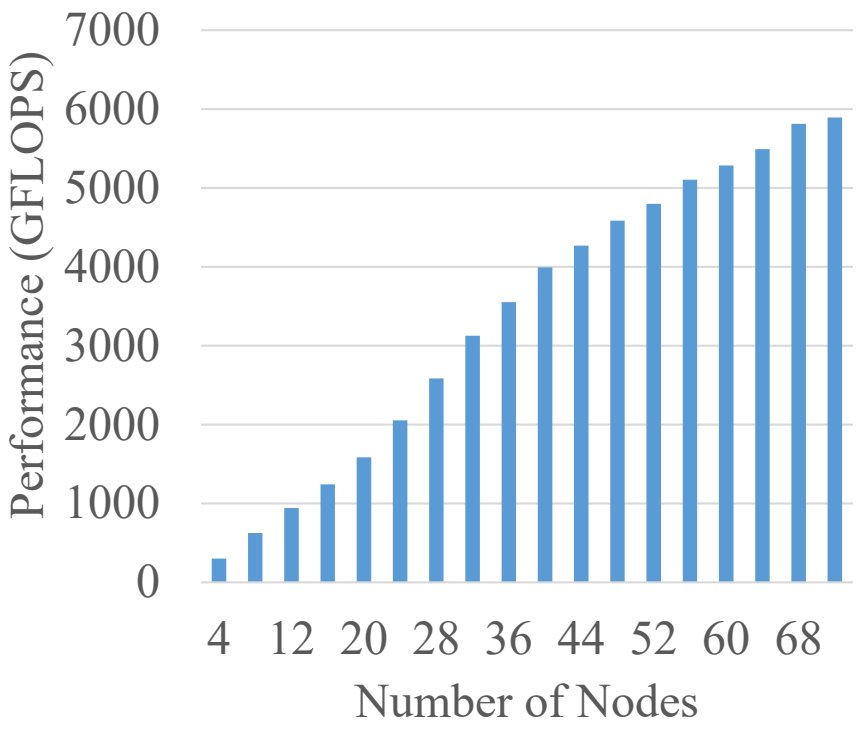
Latest GPU and CPU architectures for HPC

NVIDIA PSG P100



4.5 Teraflops

Intel Broadwell CPU  
36 cores/node, 72 nodes, 2592 cores



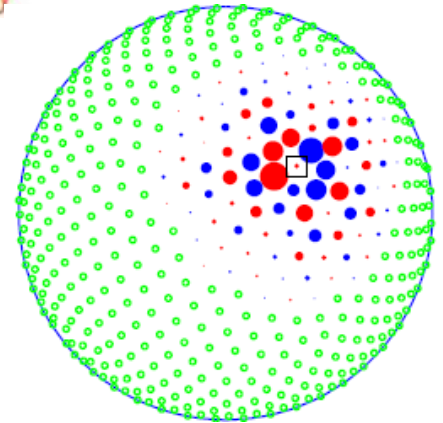
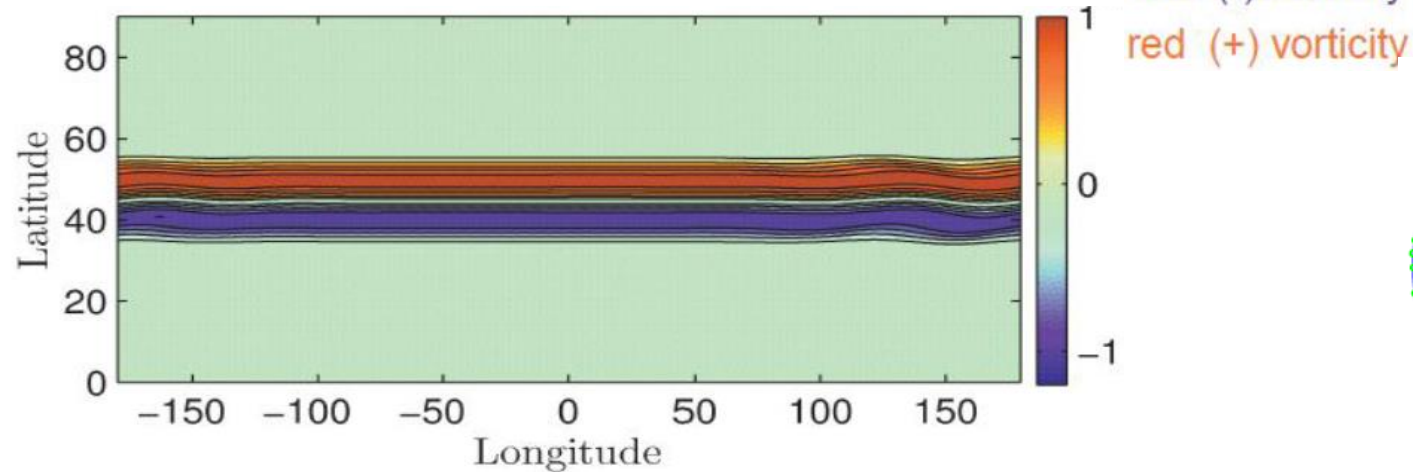
6 Teraflops

Both are > **100X** speedups over the highest achieved performance by the previous single device GPU implementation.

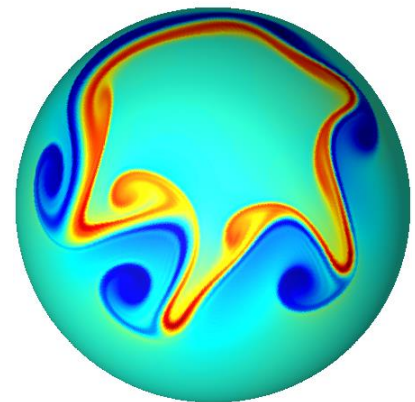
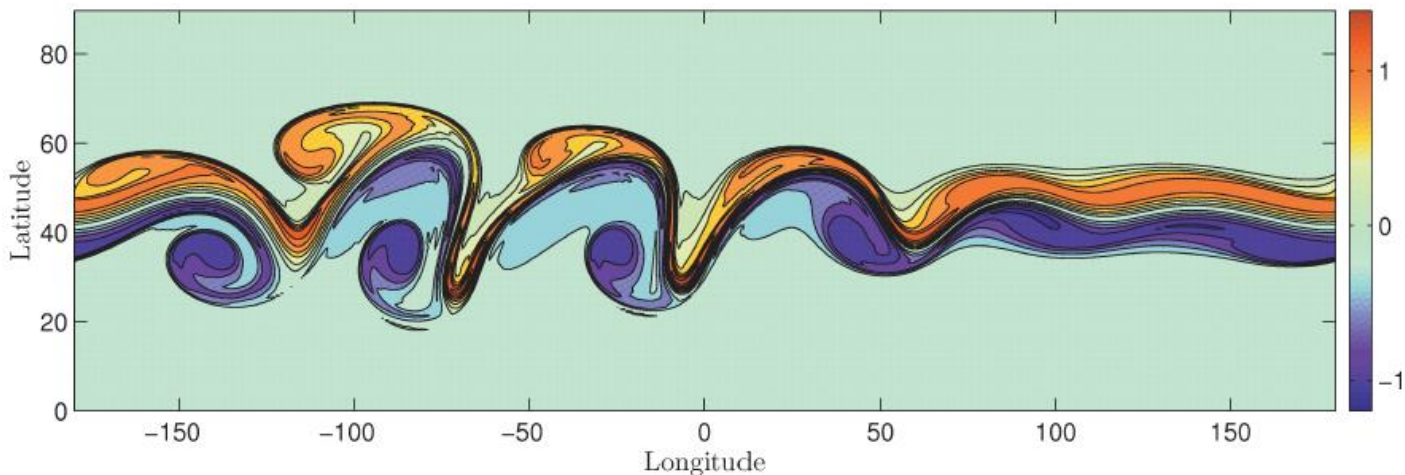


# Shallow water wave equations on the sphere: Evolution of a highly unstable wave

Day 3: Initial Signs of Instability

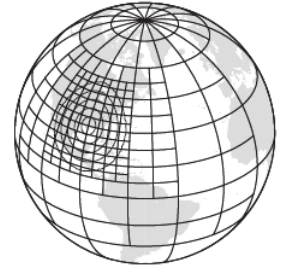


Day 6: Unstable vortex dynamics

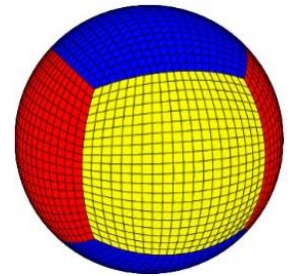


# Vorticity at $5^\circ \times 5^\circ$

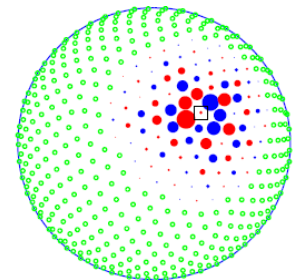
Finite Volume



Spectral Element



Discontinuous Galerkin

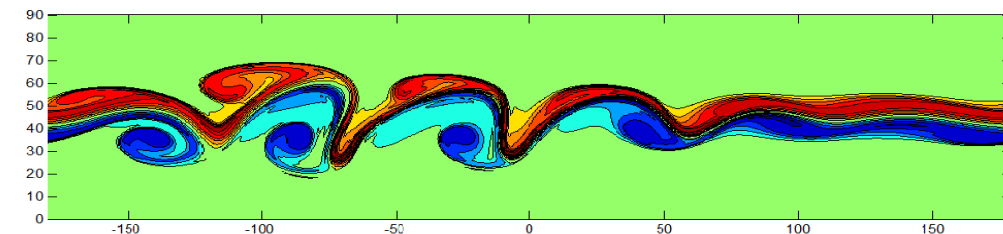
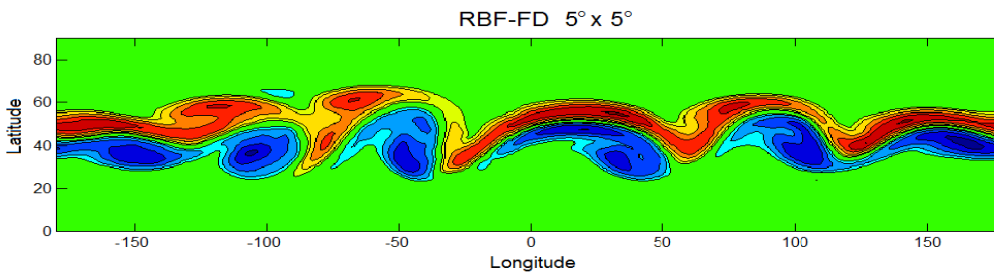
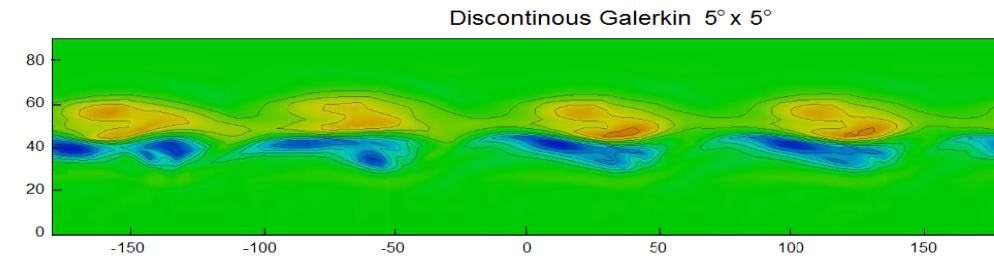
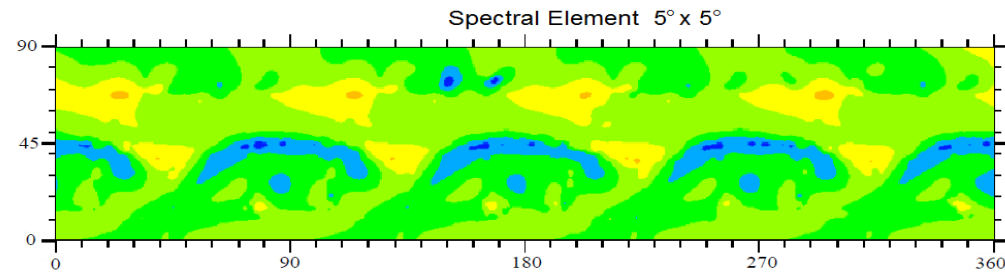
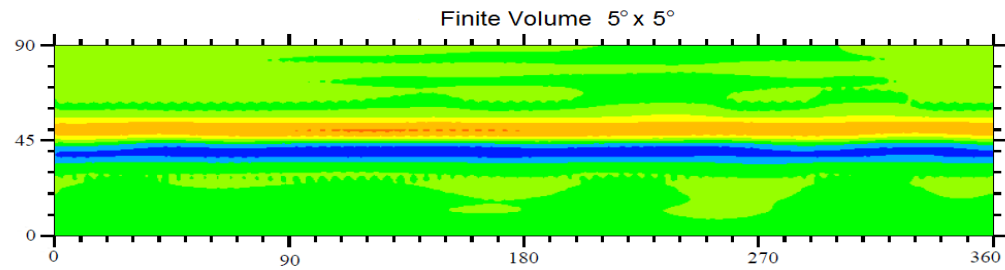


RBF-FD



“Truth”  $0.35^\circ \times 0.35^\circ$

DG, SE, RBF-FD



## 2D Compressible Navier-Stokes (Flyer, Barnett, Wicker, *JCP*, 2016)

First paper in literature to consider using polyharmonic spline (PHS) RBF with high-order polynomials.

WHY? Possible explanation:

From a historical perspective, before RBF-FD, applications of RBFs were global.

1. If PHS RBFs were used, they were used in conjunction with low-order polynomials.

Role of polynomials  $\longrightarrow$  guarantee non-singularity of RBF interpolation matrix for unusual node layouts.

The role of capturing the physics was left to the RBFs.

2. Using high-order polynomials on a global scale can be dangerous

$\longrightarrow$  Runge phenomena near boundaries.

RBF-FD gives the approximation at the center of the stencil and not at the edges.

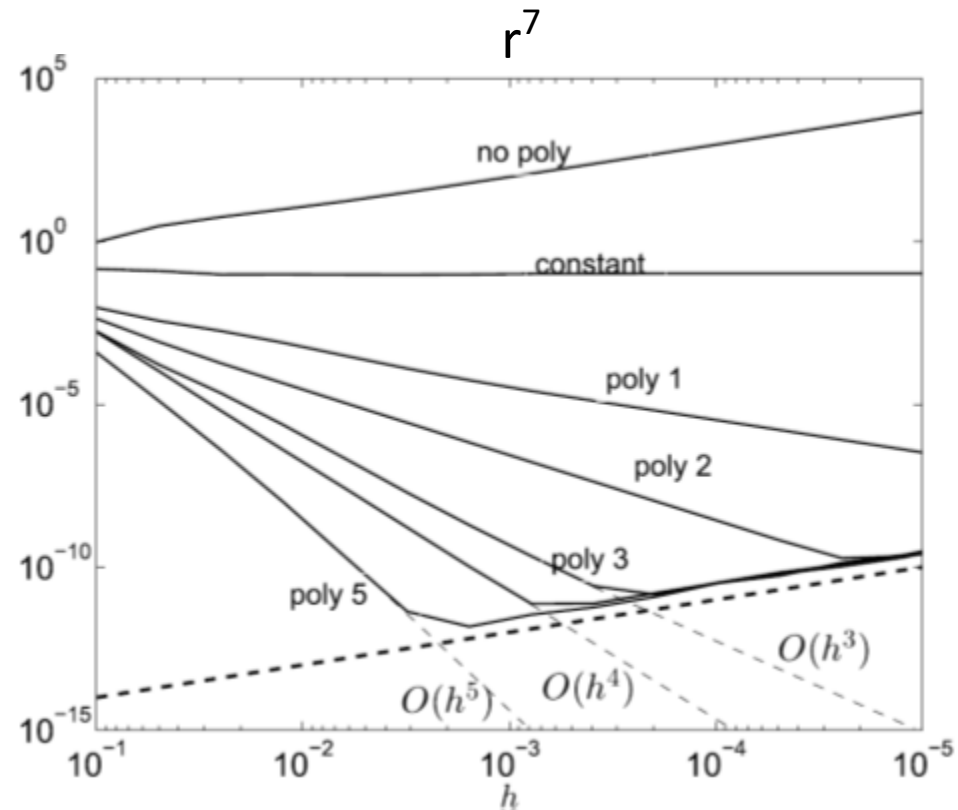
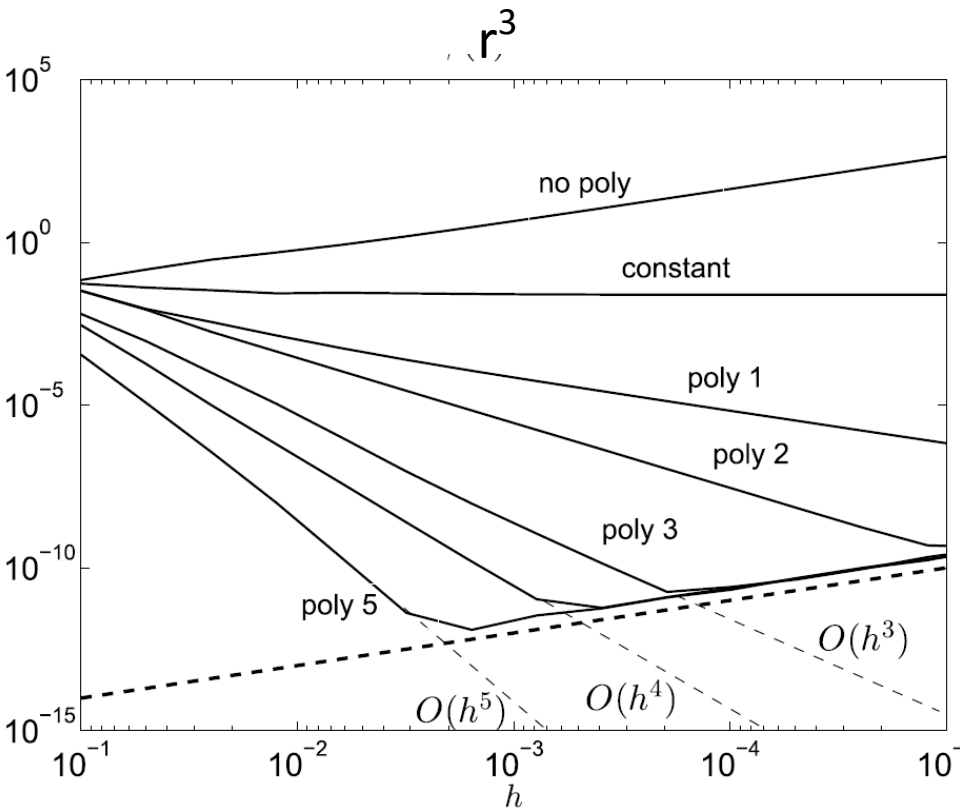
3. PHS RBFs were not as nearly as popular as infinitely smooth RBFs for PDEs.

For the high computation price of global RBFs, you want the fast convergence and accuracy.

Let's briefly explore PHS RBF-FD convergence and accuracy before test cases.

# Polynomials in Control (Flyer, Barnett, Wicker, JCP, 2016)

$L_2$  error in approximating  $d/dx$  of  $f(x, y) = 1 + \sin(4x) + \cos(3x) + \sin(2y)$   
near the center of a 37 node hex. stencil, using  $r^3$  and  $r^7$  with corresponding polynomials



Dashed line machine round-off error of  $10^{-15}/h$

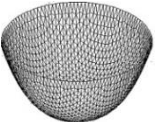
# 2D Compressible Navier-Stokes

$$\frac{\partial \mathbf{u}}{\partial t} = -(\mathbf{u} \cdot \nabla) \mathbf{u} - c_p \theta \nabla P - g \mathbf{k} + \mu \Delta \mathbf{u},$$

$$\frac{\partial \theta}{\partial t} = -(\mathbf{u} \cdot \nabla) \theta + \mu \Delta \theta,$$

$$\frac{\partial P}{\partial t} = -(\mathbf{u} \cdot \nabla) P - \frac{R}{c_v} (\nabla \cdot \mathbf{u}) P,$$

Basis functions used:

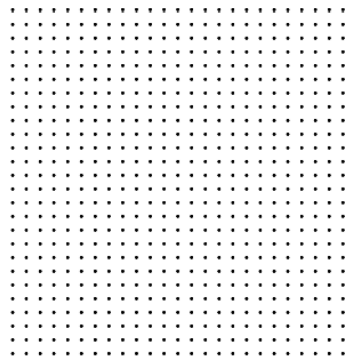
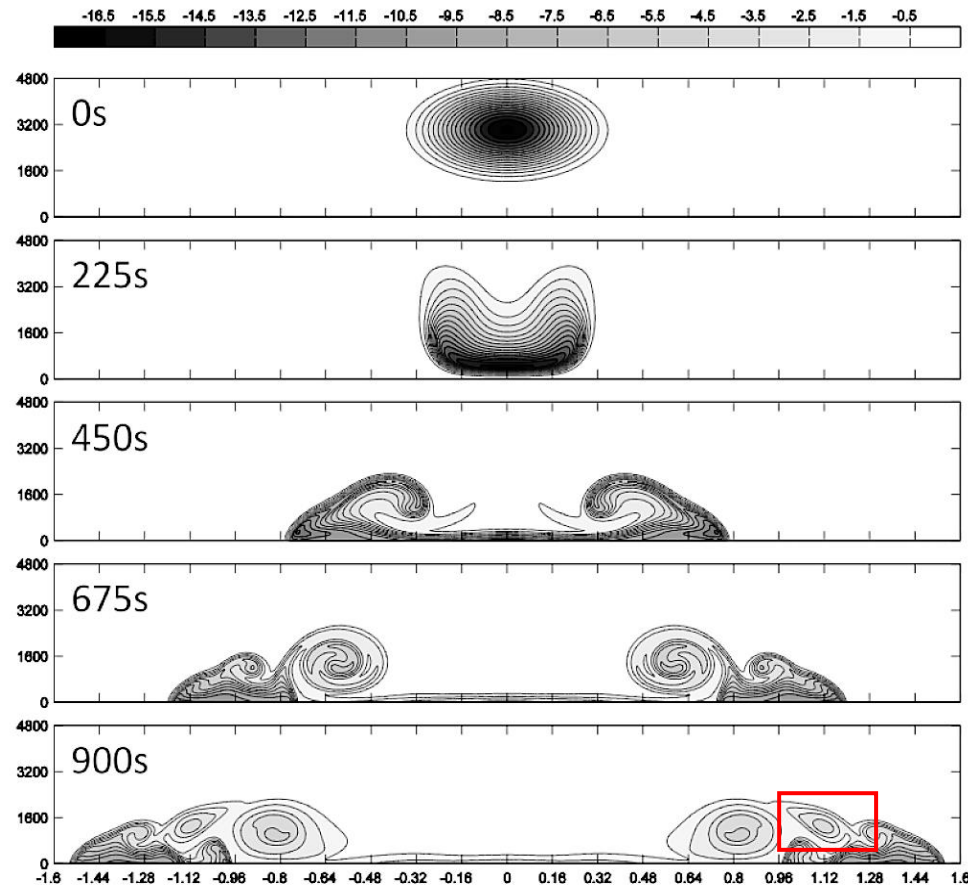
RBF  $r^5$   + Up to 4<sup>th</sup> degree polys.

Hyperviscosity use GA-based or PHS-based

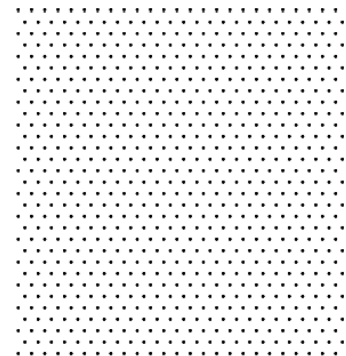
With RBF-FD, easy to explore the intrinsic capabilities of different layouts. Same Code.

Hexagonal have a long history, never became 'mainstream' due to implementation complexities.

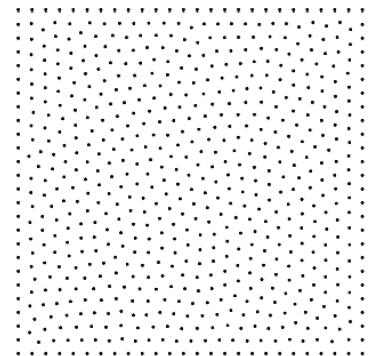
Accurate time evolution of Temperature



Cartesian



Hexagonal

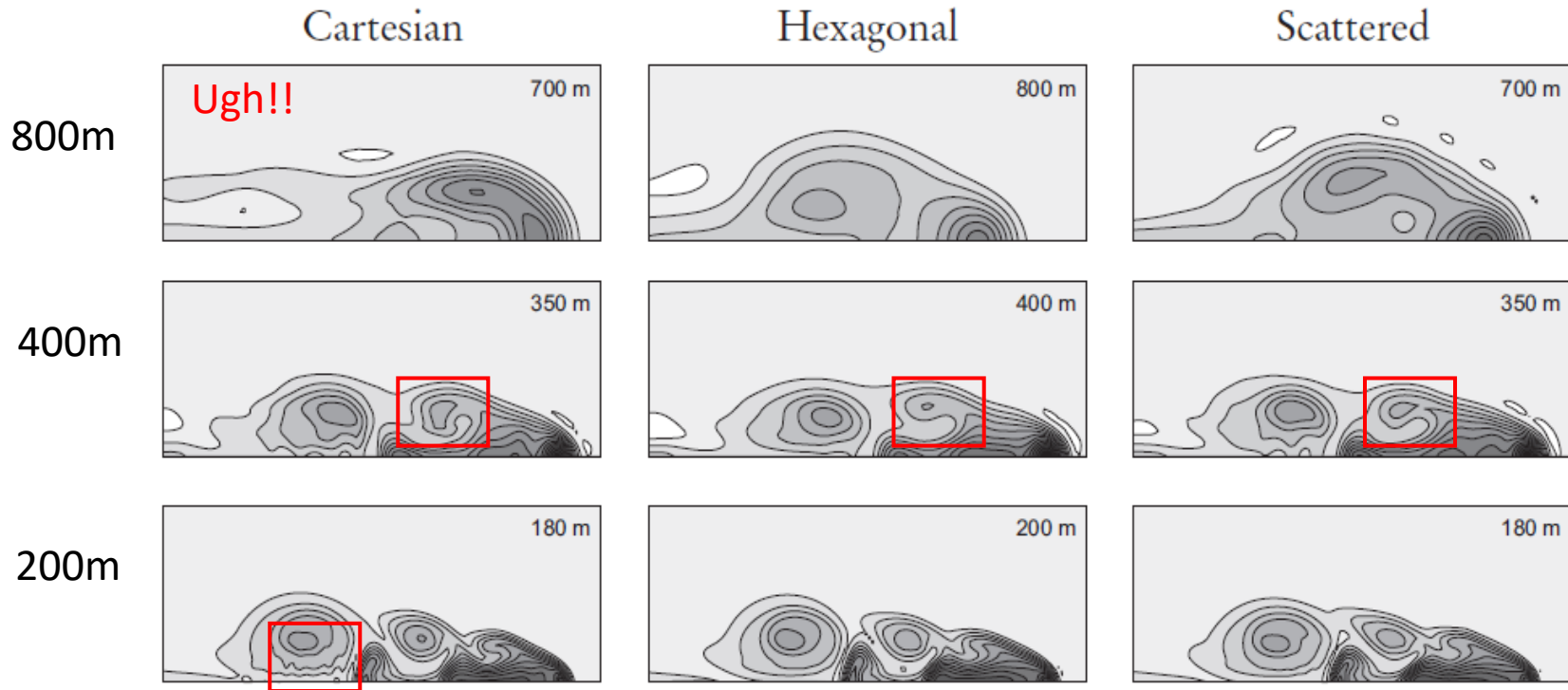


Scattered



# Comparisons on different node layouts: change 1 line of code

Only showing half of domain due to symmetry



## Comparison:

Cartesian: Most unphysical artifacts ('wiggles'), 1<sup>st</sup> rotor not formed at 800m

Hexagonal: Excellent results; now easy to implement opposed to past

Scattered: Little performance penalty but one gains greatly geometric flexibility

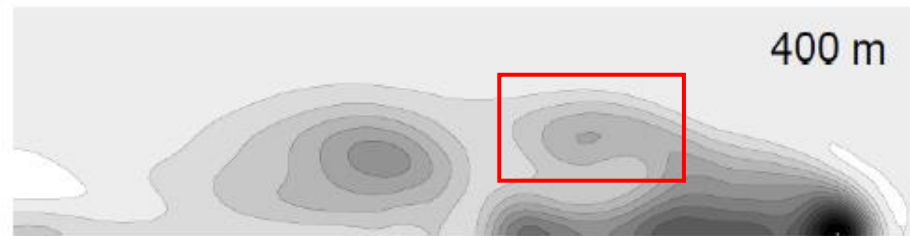
# Comparisons to other numerical methods

At high resolutions, 100m and under, most methods perform well.

Key issue: Data-based initialization of weather prediction models > 500m

Below: Comparisons from the literature, at 400m resolution?

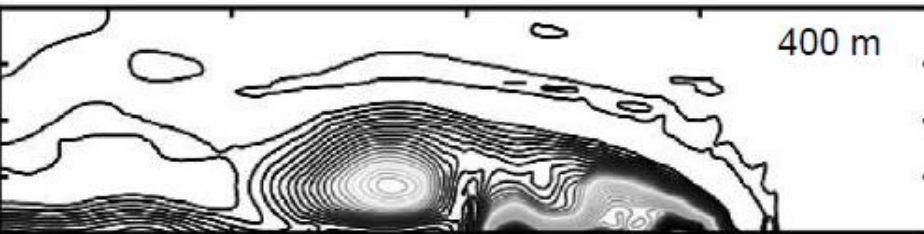
RBF – FD 37 node stencil



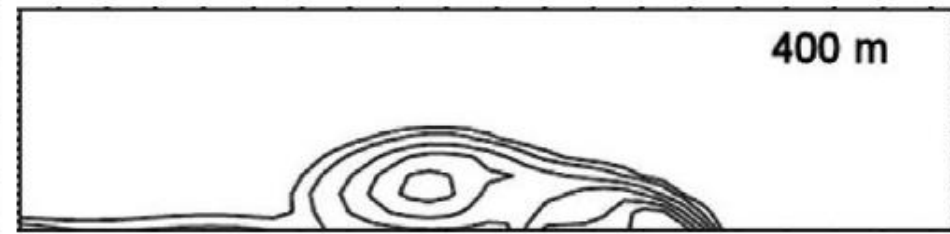
5<sup>th</sup> – order upwind advection



8<sup>th</sup> – order DG and SE



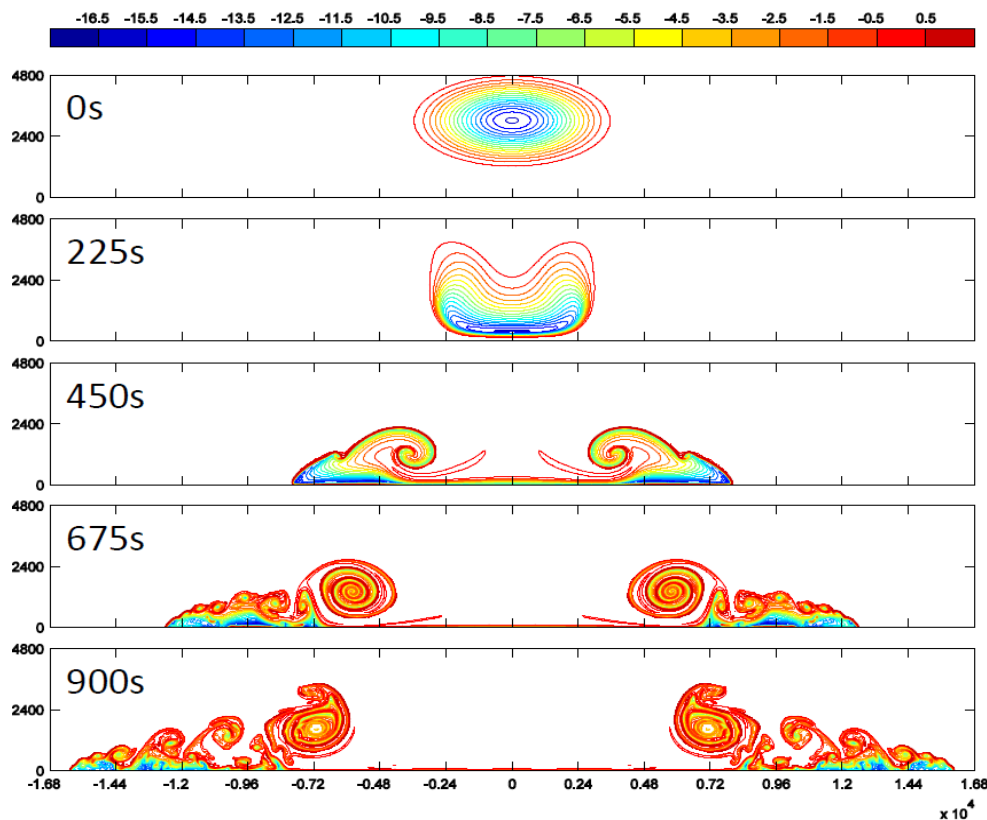
4<sup>th</sup> – order finite volume



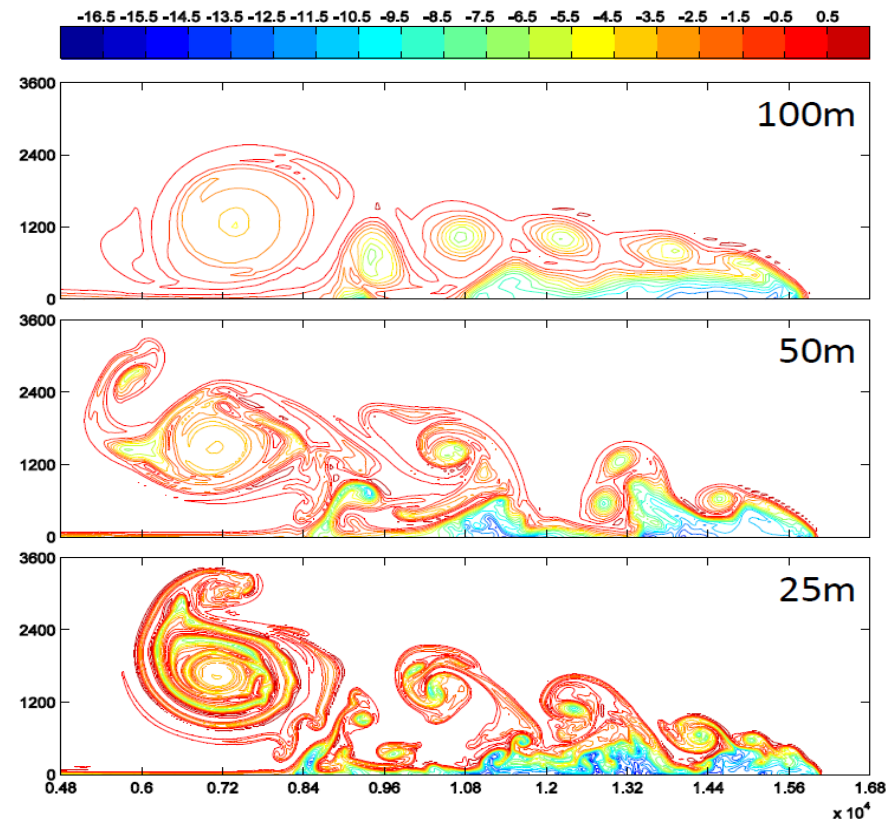
At this coarse resolution, only the RBF-FD calculations shows the beginning of second rotor (does it on Cartesian, hexagonal, and scattered node sets) and can perform at 800m.

# Same test problem, but with no physical viscosity

25m resolution (RBF-FD, hex nodes)



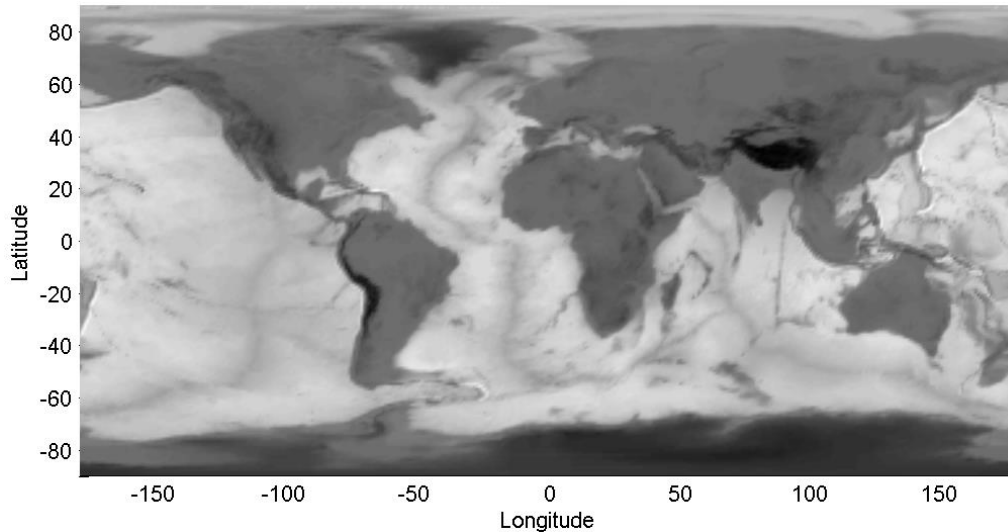
Details when using different resolutions



# Distributing variable node density on sphere

(Fornberg and Flyer, 2015)

Below: Gray scale rendering of the file topo.mat in Matlab's Mapping toolbox

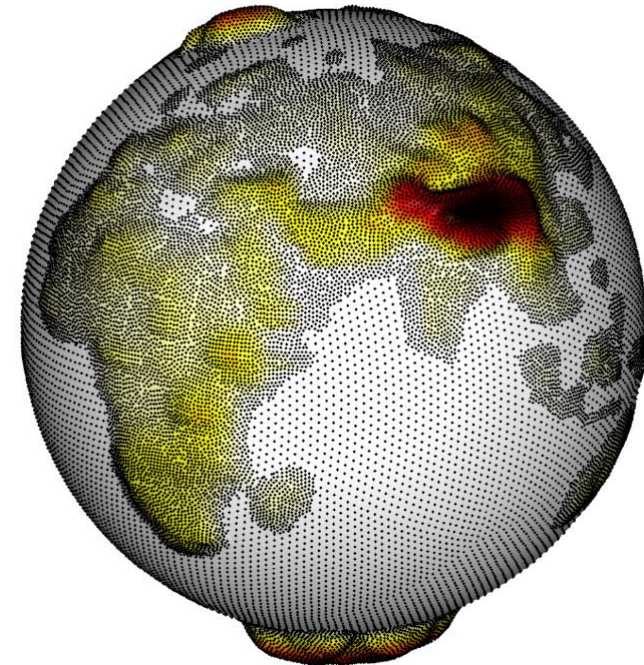
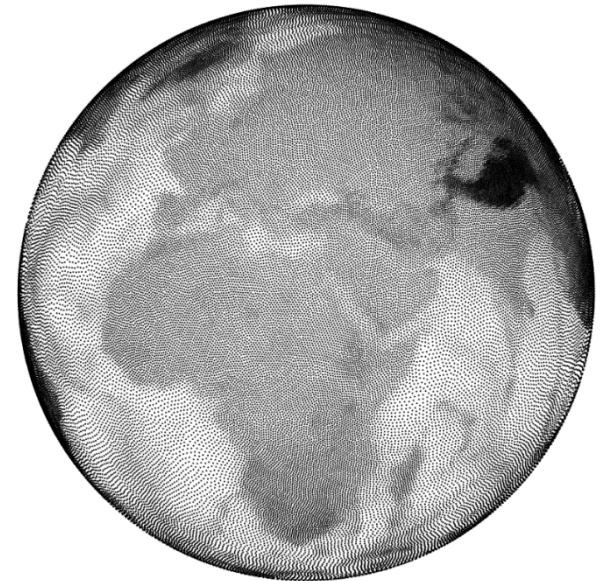


Top right:

$N = 105,419$  nodes rendering of the topo map above  
Computational speed in MATLAB still around  
11,000 nodes per second.

Next step in modeling (Bayona et al. 2015) :

Take elevation physically taken into account





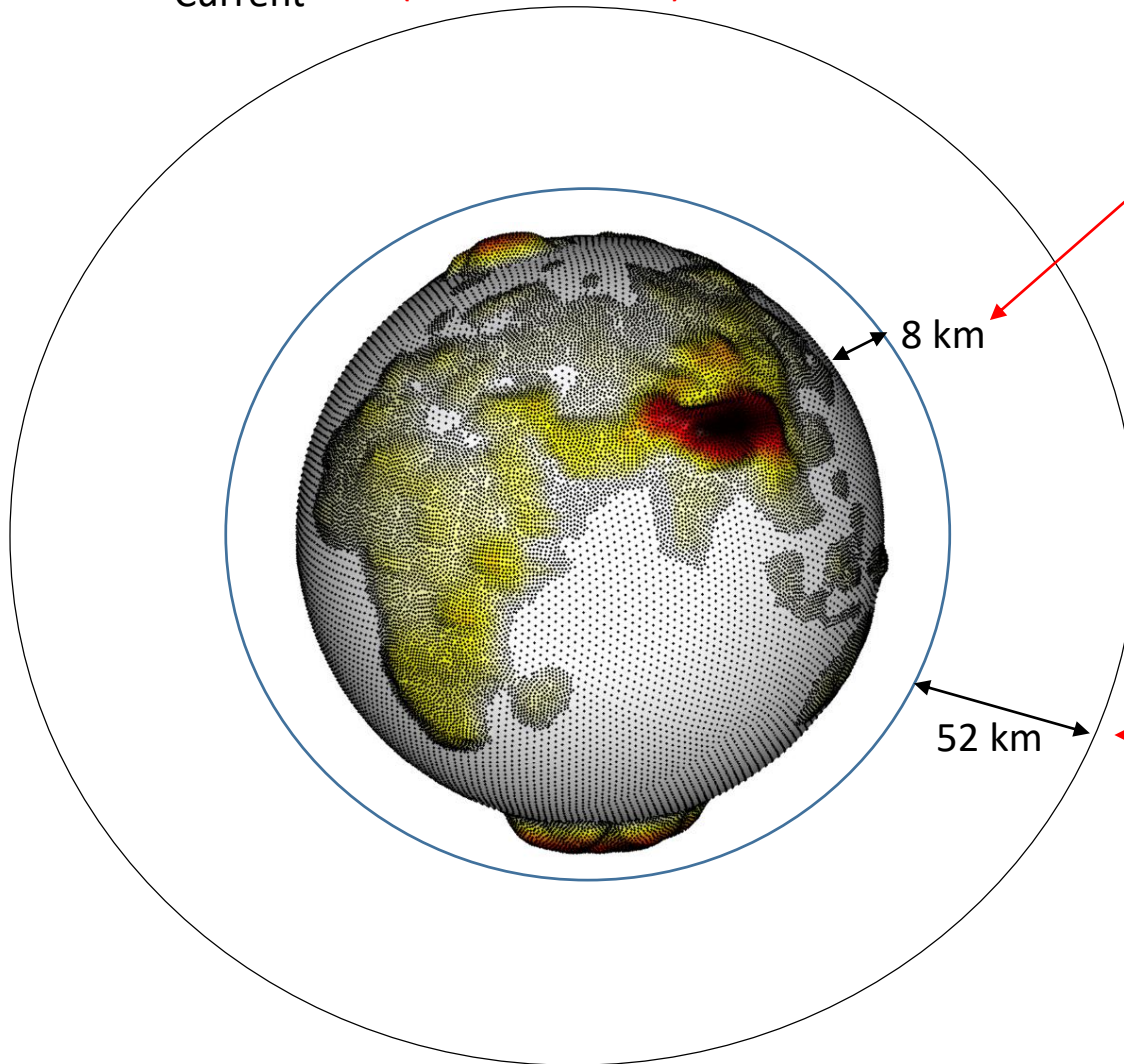
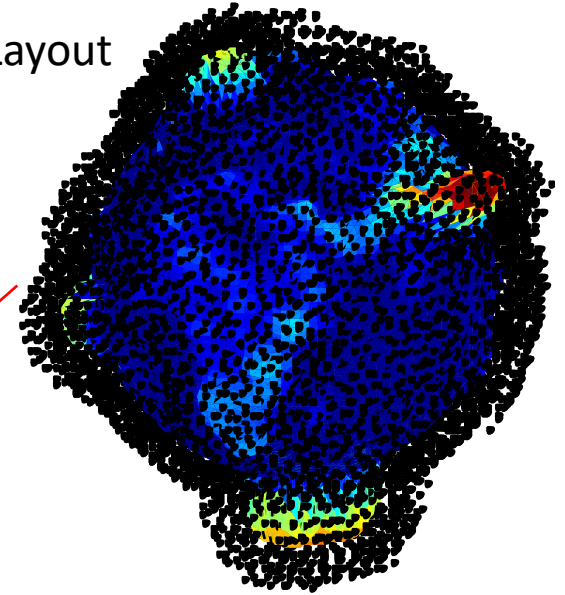
# 3D Elliptic PDE: Modeling Electrical Currents in the Atmosphere

$$\nabla_{3D} \cdot \underbrace{\vec{J}(\vec{x})}_{\text{Electric Current}} = \underbrace{F(\vec{x})}_{\text{Thunderstorms (measured data)}}$$

Electric  
Current

Thunderstorms  
(measured data)

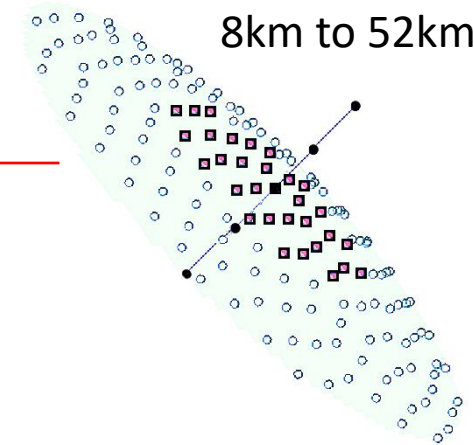
3D Node Layout  
to 8km



8 km

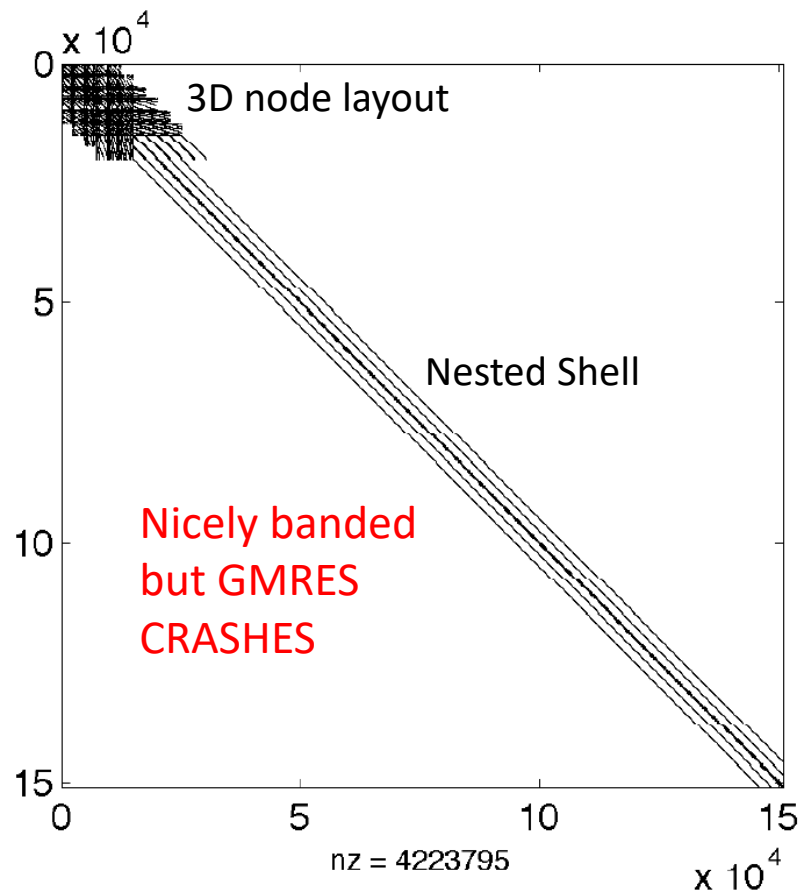
52 km

Nested shells  
8km to 52km

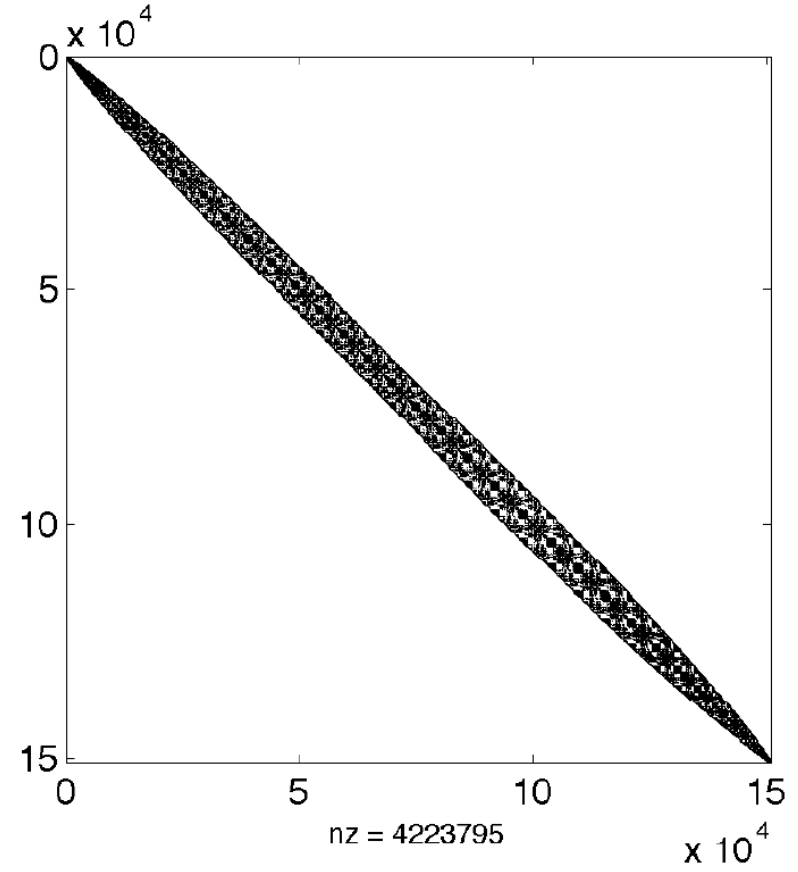




# Sparsity pattern of 3D elliptic operator (99.998% zeros)



Before any node reordering



After using reverse Cuthill- McKee

Result: Testing with data, 4.2M nodes

100 km. lat. – long. By 600m vertical, 31 mins on laptop using GMRES

[GitHub Open Source Code:](#) Bayona et al. , A 3-D RBF-FD solver for modelling the atmospheric Global Electric Circuit with topography (GEC-RBFFD v1.0), Geosci. Model Dev. 2015.

# Tracer Transport in 3D Spherical Shell

$$\partial q / \partial t + \mathbf{v}(x,y,z,t) \cdot \nabla_{3D} q = 0$$

Specs:

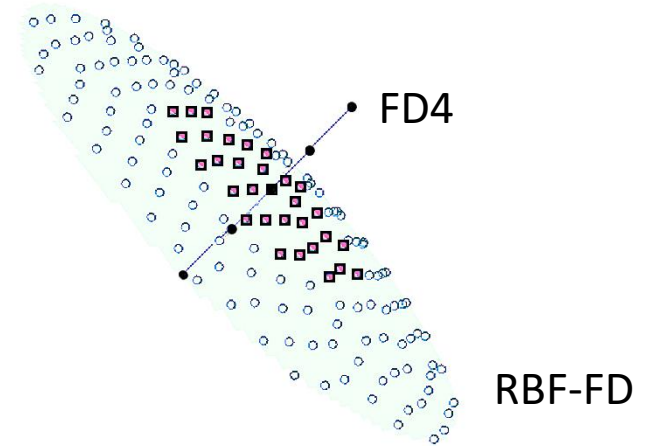
Nodes: Icosahedral on nested spheres

RBF:  $r^3$  with up to 5<sup>th</sup>-order polynomials on sphere

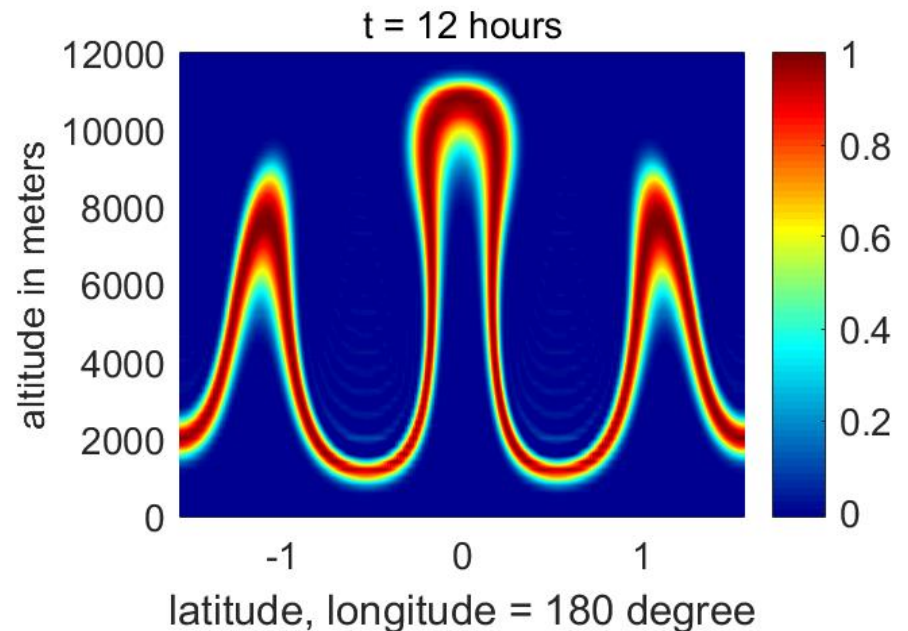
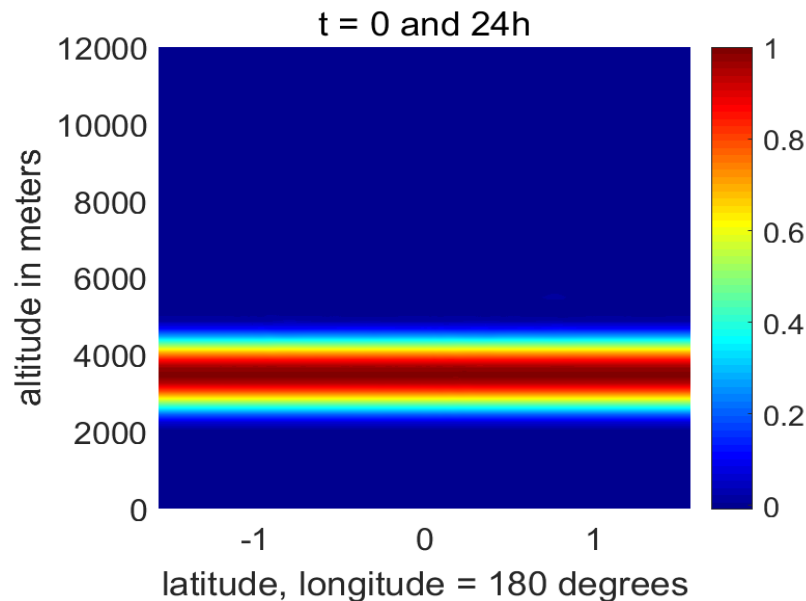
FD4: In vertical

Stencil:  $n = 55$

No Hyperviscosity Needed!



Concentration of tracer  $q$  plotted



# Comparison to other community models based on finite volume

Numbers represent error in  $L_2$

	2° by 300m (N = 360K)	1° by 200m (N = 2.45M)	1/2° by 100m (N = 19.6M)
CAM-FV	0.20	0.05	0.02
Mcore(FV)	0.17	0.05	0.01
RBF-FD	0.03	0.003	0.0005

FV is used for its conservation properties, but sacrifice is accuracy and convergence.


Comment: The need for hyperviscosity depends on how long it takes for the spurious eignmodes that are close to machine rounding to grow.

# Conclusions

## Established:

- RBF-FD latches onto the physics at much coarser resolutions than other numerical methods, giving higher accuracy and convergence
- RBF-FD have shown strong linear scaling on on the latest HPC platforms
- Startup cost for modeling with RBF-FD is cheap due to their algorithmic simplicity

## Some recent review material

1. N. Flyer, G.B. Wright, and B. Fornberg, 2014.  
*Radial basis function-generated finite differences:  
A mesh-free method for computational geosciences*,  
Handbook of Geomathematics, Springer-Verlag
2. B. Fornberg and N. Flyer, 2015  
*Solving PDEs with Radial Basis Functions*,  
Acta Numerica.
3. B. Fornberg and N. Flyer, 2015   
*A Primer on Radial Basis Functions with  
Applications to the Geosciences*, SIAM Press.

