

Mesh-Free Applications for Static and Dynamically Changing Node Configurations

Natasha Flyer

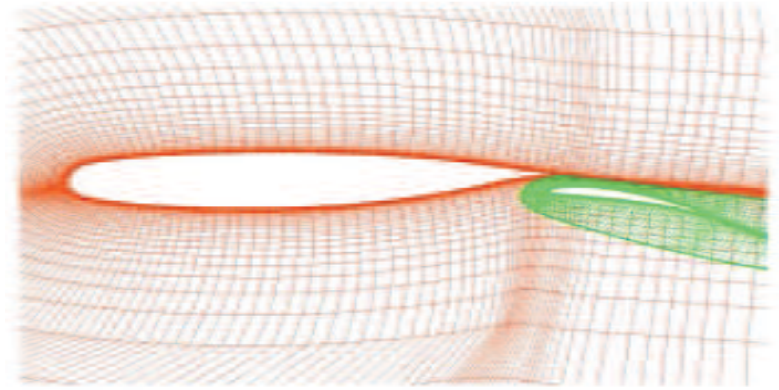
Computational Information Systems Lab
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Boulder, CO

Meshes vs. Mesh-free discretizations

Structured meshes:

FD, DG, FV, Spectral Elements

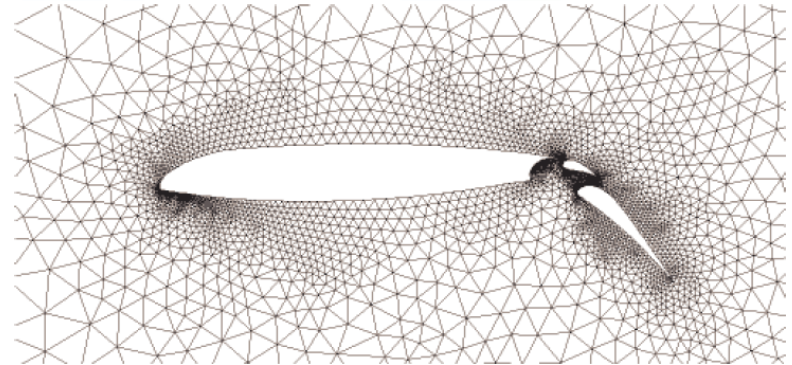
Requires domain decomposition /
curvilinear mappings



Unstructured meshes:

FEM, DG, FV, Spectral Elements

Improved geometric flexibility; requires
triangles, tetrahedral, etc.

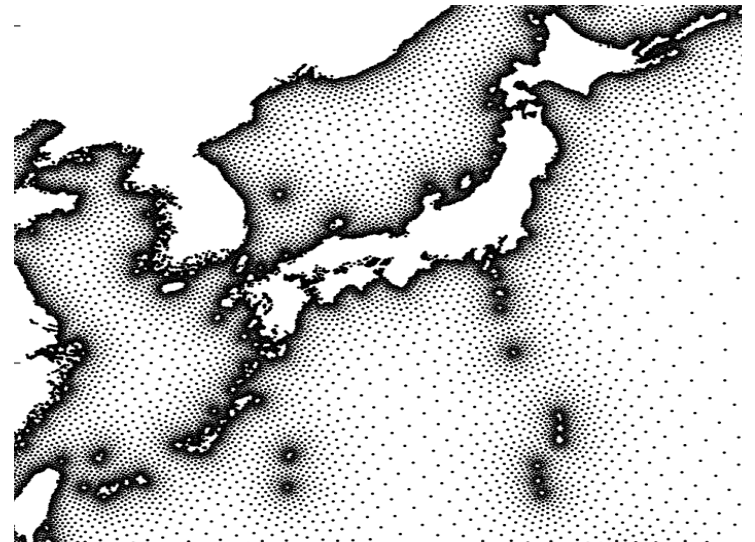


Mesh-free:

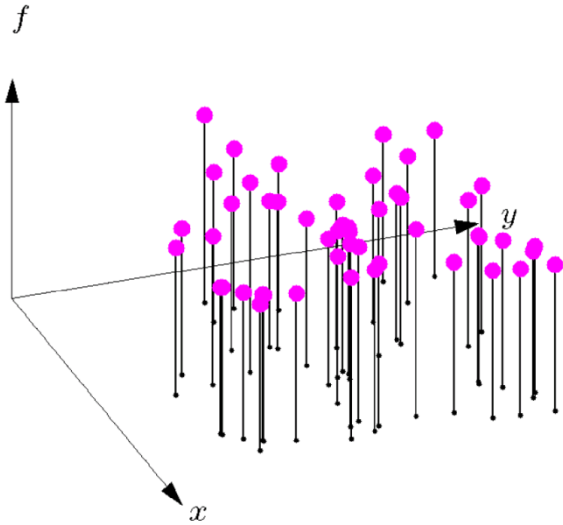
RBF-FD

(Radial basis Func.-generated Finite Differences)

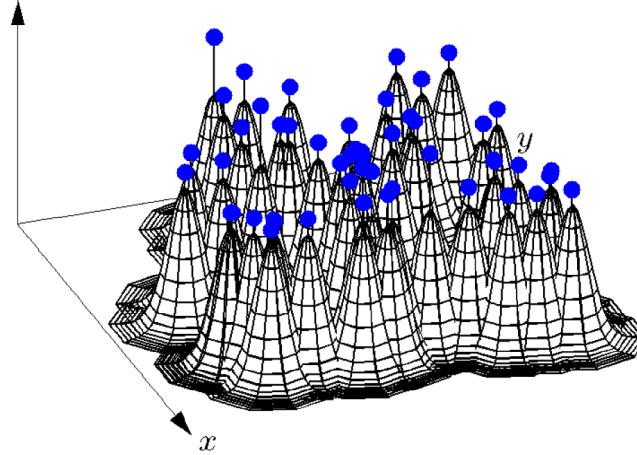
Total geometric flexibility;
needs node locations, but no connectivities,
e.g. no triangles or mappings



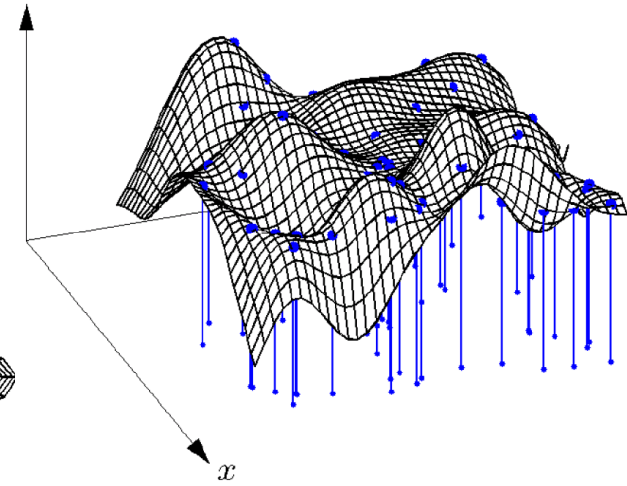
General RBF-FD Concept



Scattered data in 2D

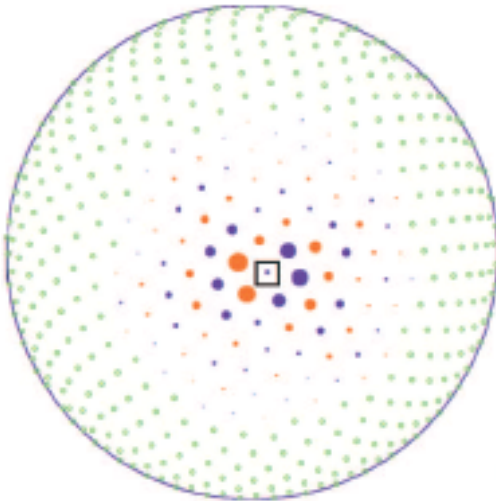


Collocate RBF, e.g. 2D Gaussians

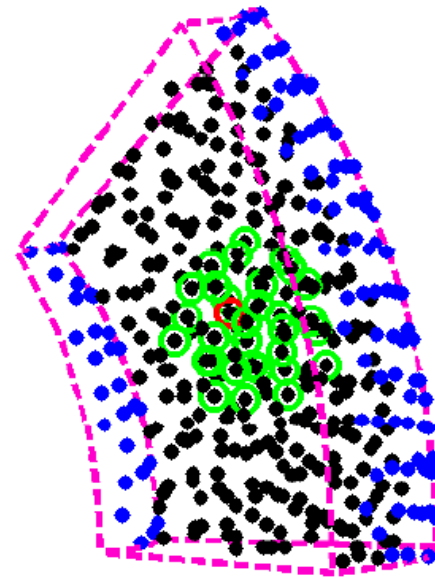


Find linear combination of RBF that fits all the data

On a sphere

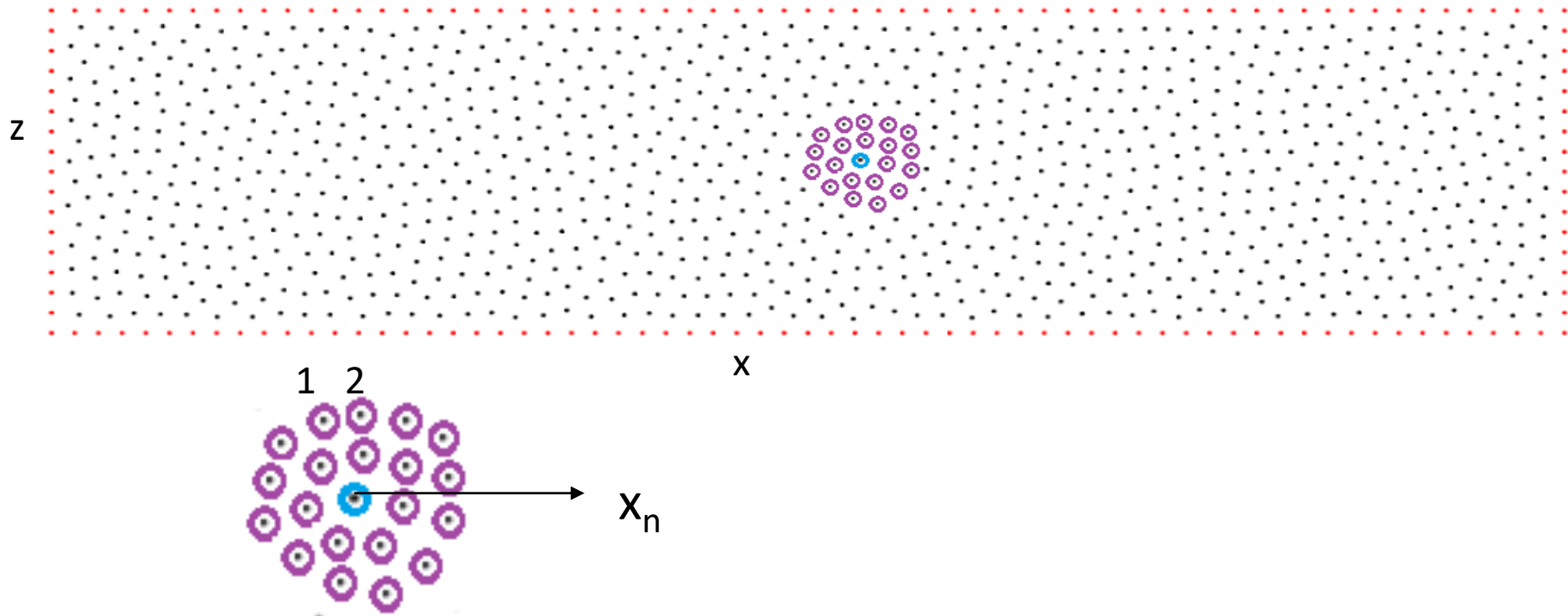


In a 3D



Simplicity of RBF-FD: Mesh-Free method

Ex.: Stencil of $n = 21$ nodes



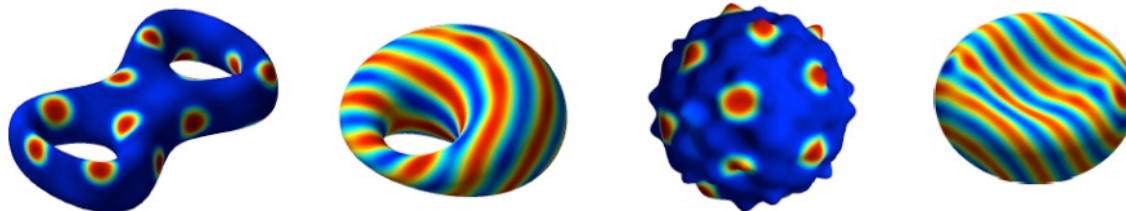
Get Gaussian matrix \mathbf{A} , use Gaussian elimination to solve for weights

$$\begin{bmatrix} \text{Exp}(-r_{11}^2) & \text{Exp}(-r_{12}^2) & \dots \text{Exp}(-r_{1n}^2) \\ \vdots & \ddots & \vdots \\ \text{Exp}(-r_{n1}^2) & \text{Exp}(-r_{n2}^2) & \dots \text{Exp}(-r_{nn}^2) \end{bmatrix} \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix} = \begin{bmatrix} d/dx[\text{Exp}(-r_1^2)]|_{\text{blue dot}} \\ \vdots \\ d/dx[\text{Exp}(-r_n^2)]|_{\text{blue dot}} \end{bmatrix}$$

Coding RBF-FD Method is FAST and EASY

```
IDX = knnsearch(xyz,xyz,'K',n); % n is stencil size
for k = 1:N                      % Loop over all points N in domain
    X = xyz(IDX(k,:),:);         % nodes in the kth stencil
    r2 = (X(:,1) - X(k,1)).^2 + ...
          (X(:,2) - X(k,2)).^2 + ...
          (X(:,3) - X(k,3)).^2 ; % Distance matrix
    A = exp(-r2);                % RBF-FD matrix
    RHS_dx = -2*(X(:,1) - X(k,1)).* exp(-r2); % derivative of GA w.r.t x,y,z
    RHS_dy = -2*(X(:,2) - X(k,2)).* exp(-r2);
    RHS_dz = -2*(X(:,3) - X(k,3)).* exp(-r2);
    Dx(k,:) = A\RHS_dx;          % Differentiation matrices (DM)
    Dy(k,:) = A\RHS_dy;
    Dz(k,:) = A\RHS_dz;
end
```

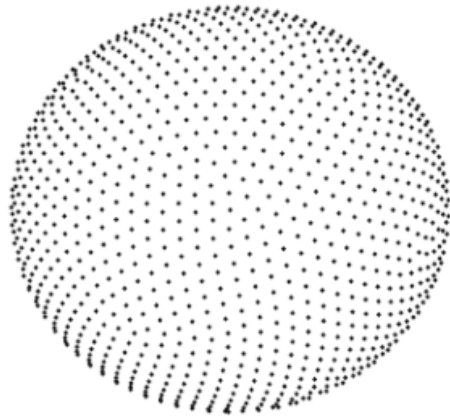
Have DMs for any geometry and point distribution in 3D space



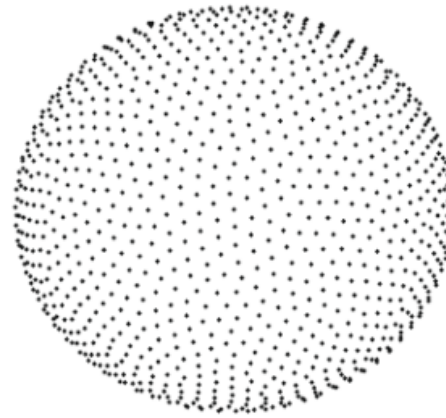
Translating Static Node Refinement (Flyer and Lehto, 2010, *JCP*)

$$\partial h / \partial t = \vec{U}(\omega(lat.), long., t) \cdot \nabla h$$

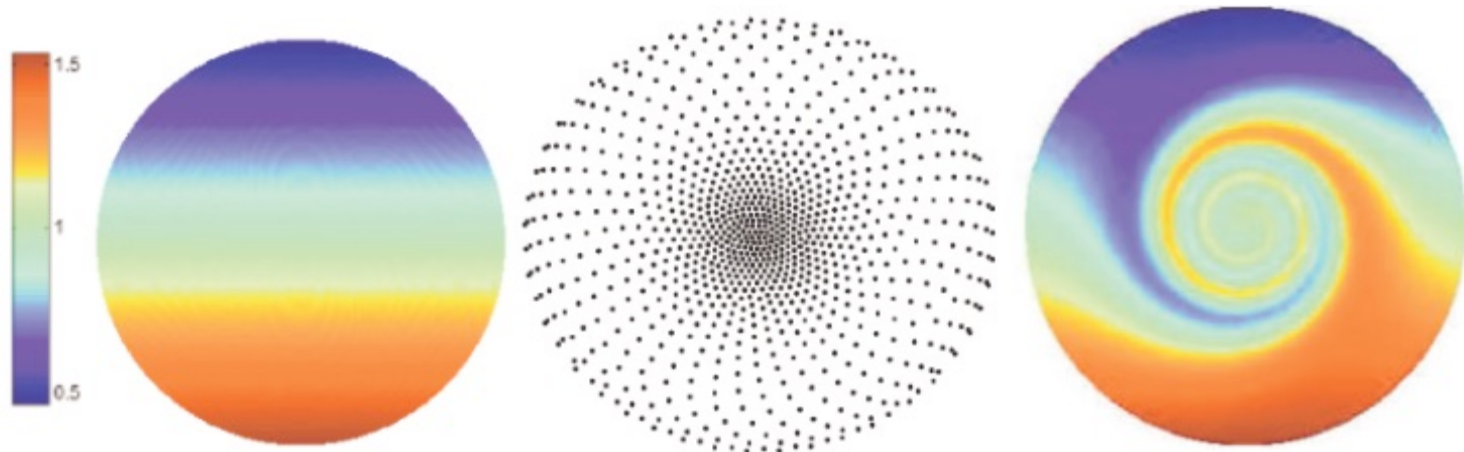
<http://web.maths.unsw.edu.au/~rsw/Sphere/>

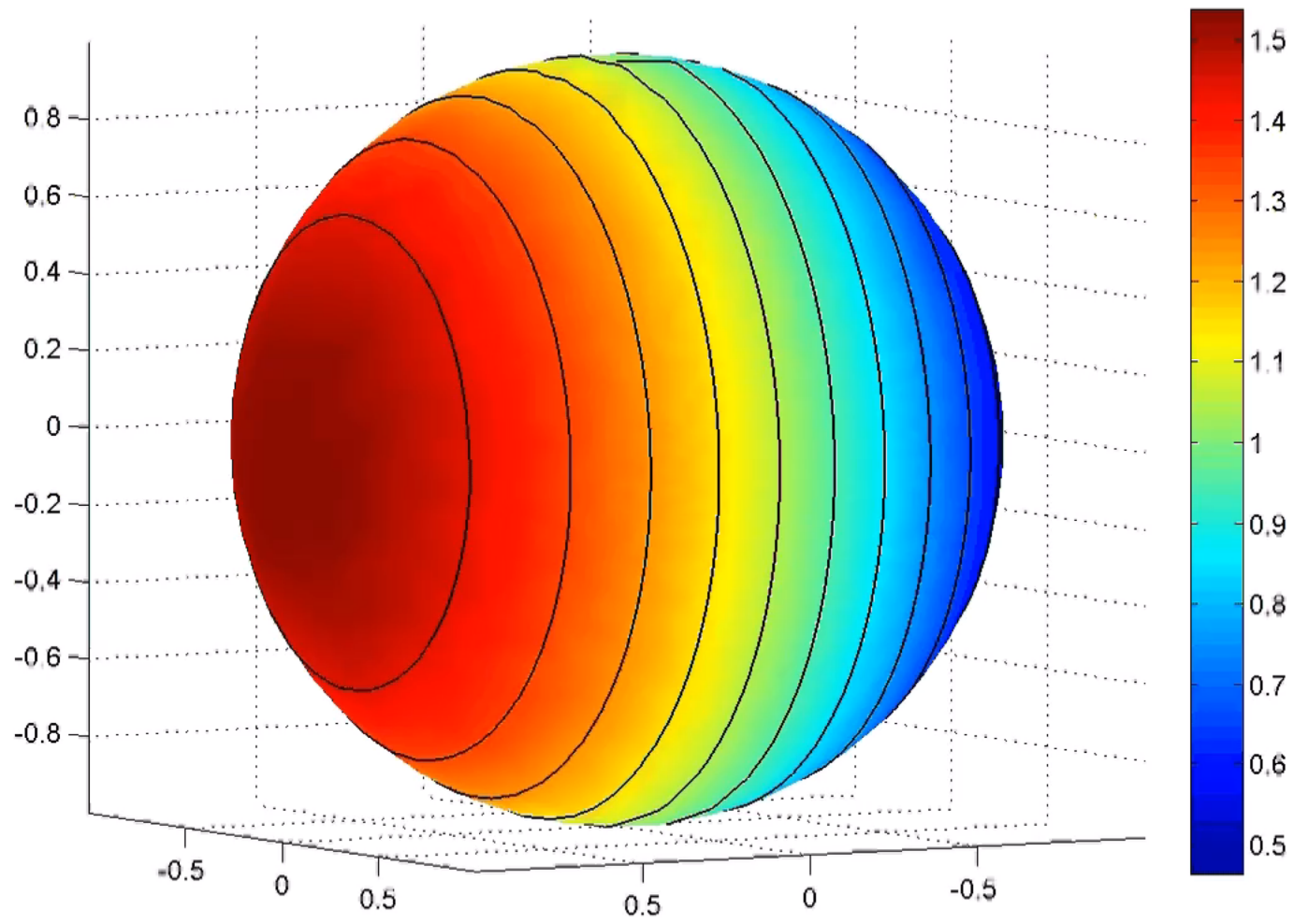


(a) ME Nodes



(b) MD Nodes





Comparison between ME, MD, Refined, and other methods

Method	N	Δt (mins.)	RMS error
RBF, refined	900	60	5e-3
RBF, ME	3136	60	4e-3
RBF, MD	3136	60	5e-3
DG	9600	6	7e-3
FV	38,400	30	2e-3
FV, AMR	2500 to 165K	Variable	2e-3

How we Cluster Nodes and Variable Shape Parameter

Fine features in flow are formed where ω is large

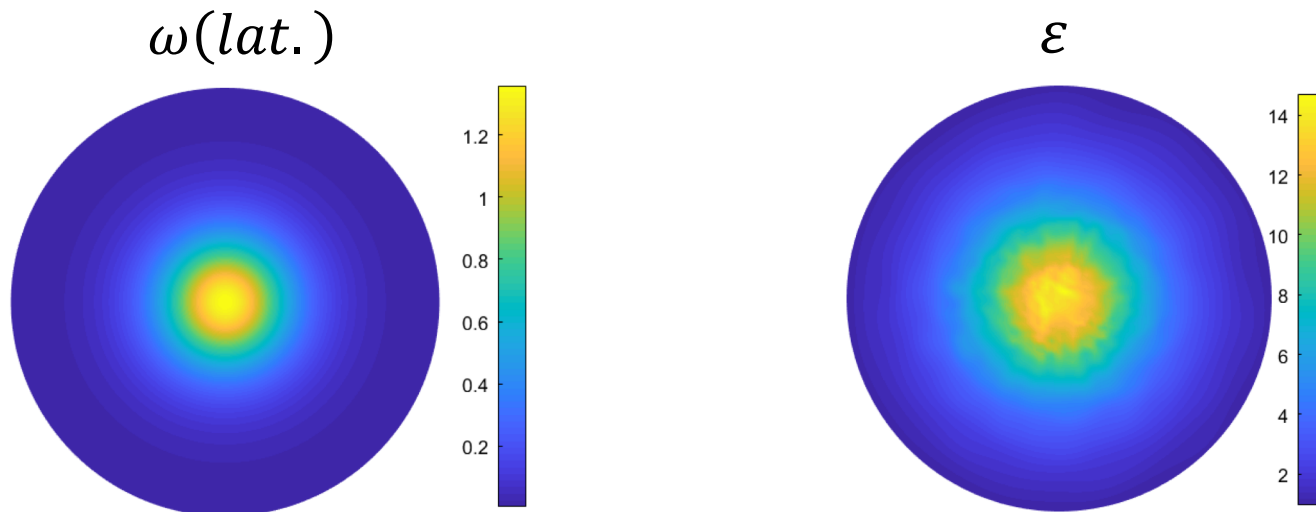
→ use ω to assign charge distribution for node repel

$$\omega(lat.) \propto \text{sech}^2(lat.) \tanh(lat.) \longrightarrow q(lat.) = [0.1 + \textcolor{red}{c} * \omega(lat.)]^{-1}$$

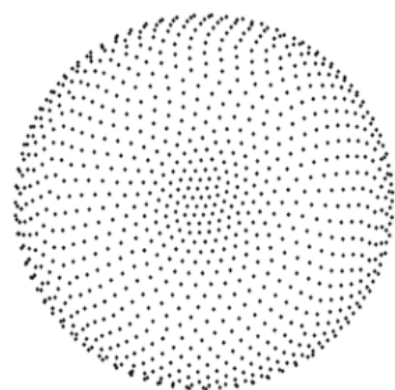
When clustering nodes, the shape parameter of the Gaussian $\text{Exp}[-(\textcolor{red}{\epsilon})^2]$ must scale over the domain to avoid ill-conditioned matrices and Runge phenomena

Rule-of-Thumb: $\textcolor{blue}{\epsilon} \propto \text{Inverse of Euclidean distance to nearest neighbor}$

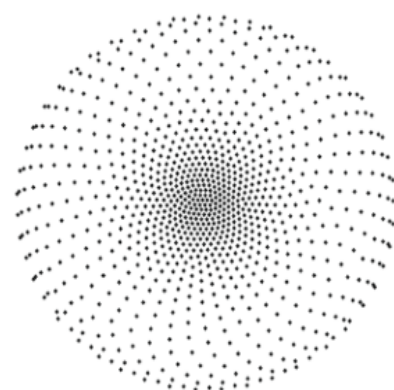
$$\omega(lat.) \longrightarrow q(lat.) \longrightarrow \text{Node distribution} \longrightarrow \epsilon$$



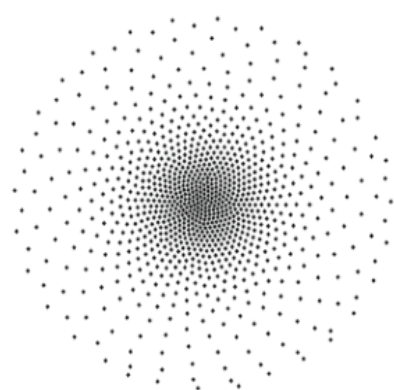
Effect of Clustering on Error



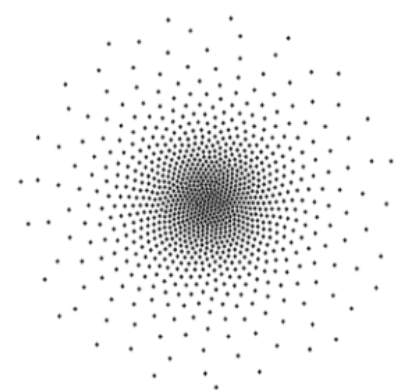
$c = 0.1$



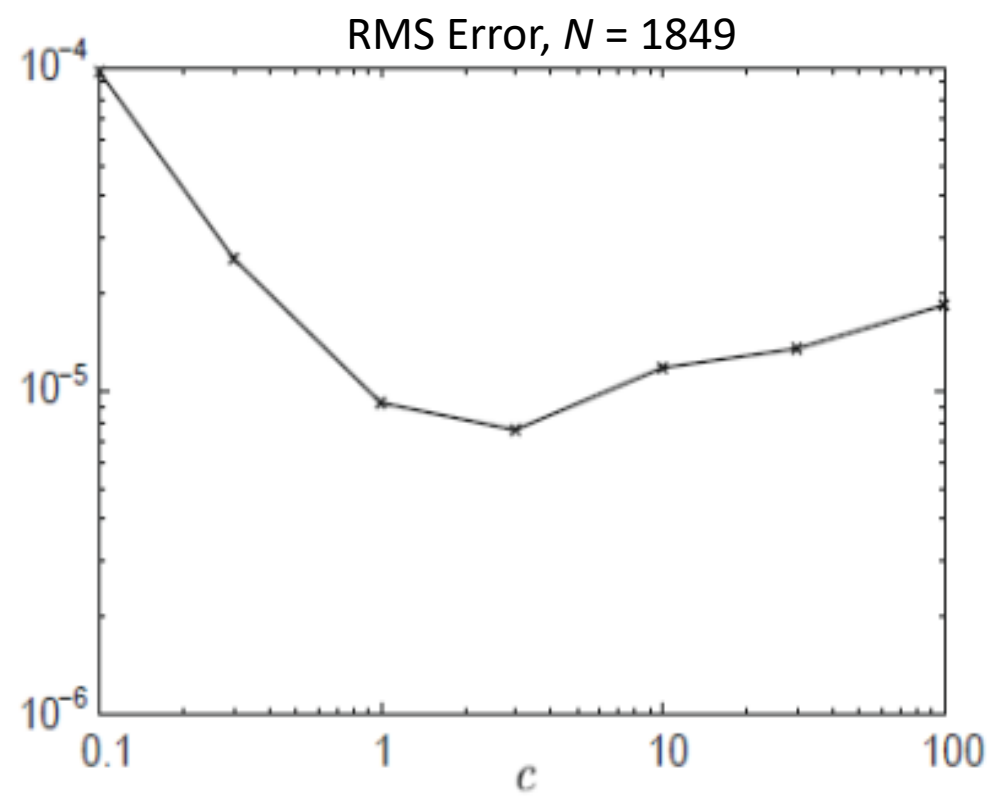
$c = 1$



$c = 10$



$c = 100$



Dynamic Node Refinement: Simple Tropical Cyclogenesis

(in collaboration with Erik Lehto)

Barotropic Vorticity Equation

$$\frac{\partial \zeta}{\partial t} + u \frac{\partial \zeta}{\partial x} + v \frac{\partial \zeta}{\partial y} + \beta v = 0 \quad \nabla^2 \psi = \zeta \quad \begin{array}{l} \psi = \text{streamfunction} \\ \zeta = \text{relative vorticity} \end{array}$$

$$u = -\frac{\partial \psi}{\partial y} \quad v = \frac{\partial \psi}{\partial x}$$

- We do not know apriori where fine features will occur
- Need a good Monitor Function for node adaptation
- Generally, takes the form $M = M(\nabla f)$, f is a physical feature of the flow

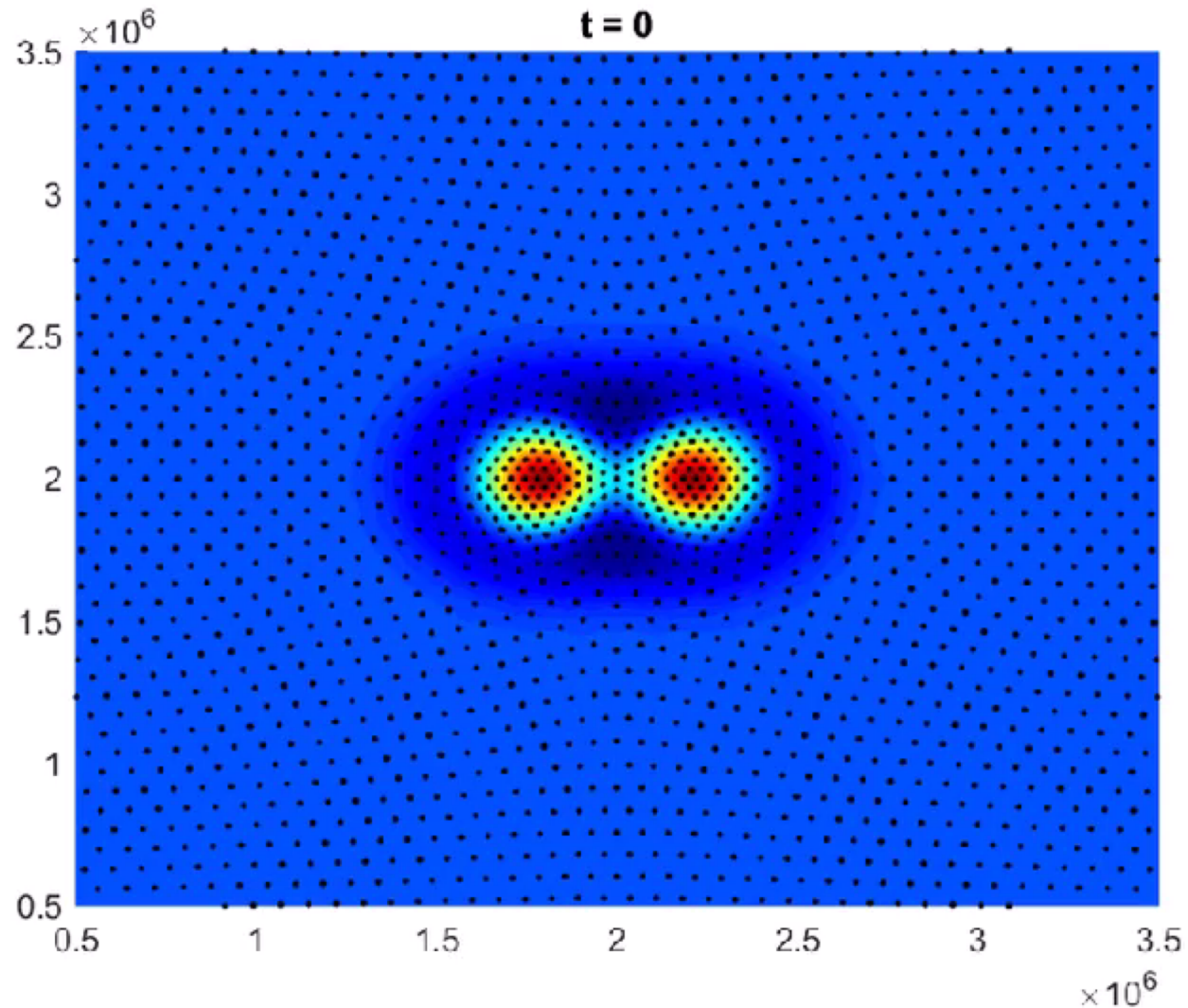
$$M(x, y, t) = \sqrt{1 + \frac{|\nabla \zeta|^2}{\alpha}} \quad \alpha \text{ is a scaling parameter}$$

Steps in implementation

1. M at a given time is approximated with RBFs
2. Assign charge distribution according $1/M$ and repel
3. Evaluate solution at new pts. via RBF interpolation
4. Calculate spatially variable ε and recalculate RBF differentiation matrices

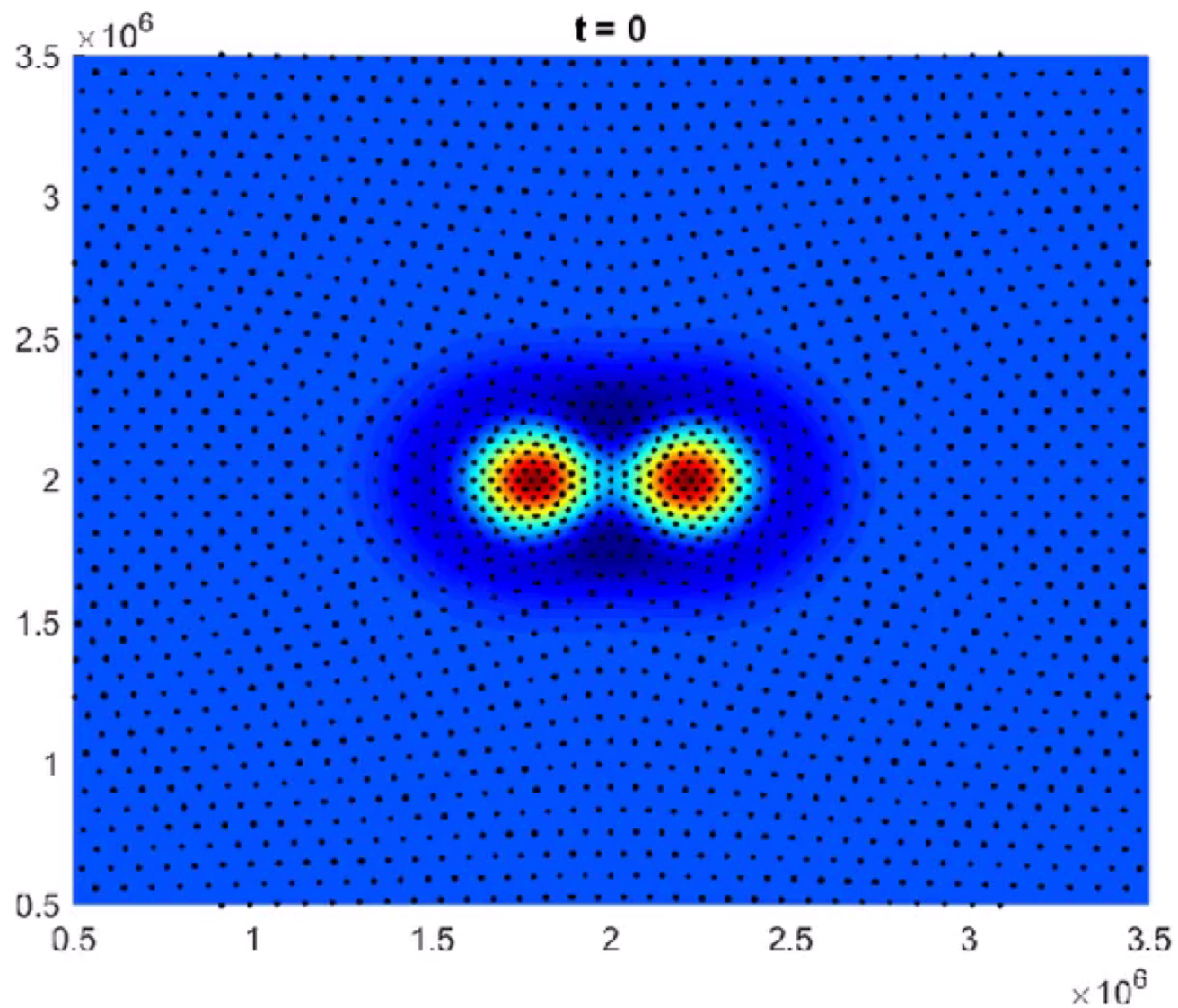
3500km x
3500km

Separation
Dist: 400km



3500km x
3500km

Separation
Dist: 410km



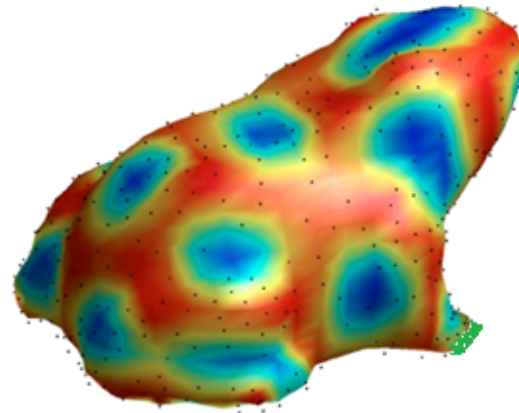
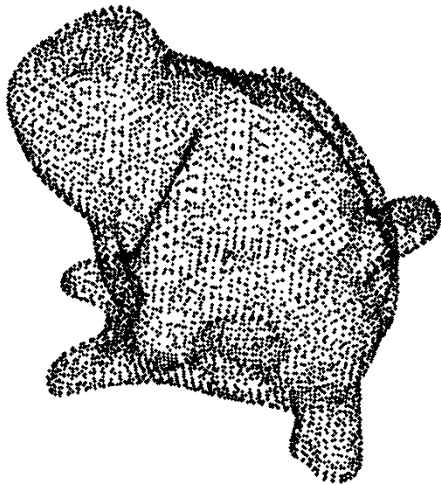
```
fname=['movie\mov' num2str(n) 'img' num2str(i)];
```

Coupled Reaction-diffusion equations over irregular surfaces (Piret, JCP 2012)

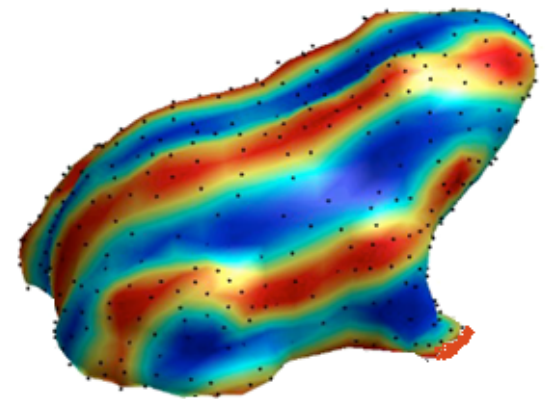
The *Brusselator equations* (Alan Turning) model pattern formation in nature,
Solved by RBF over the surface of a frog

Snapshots from a computed time evolution for two different parameter regimes

Tabasara rain frog



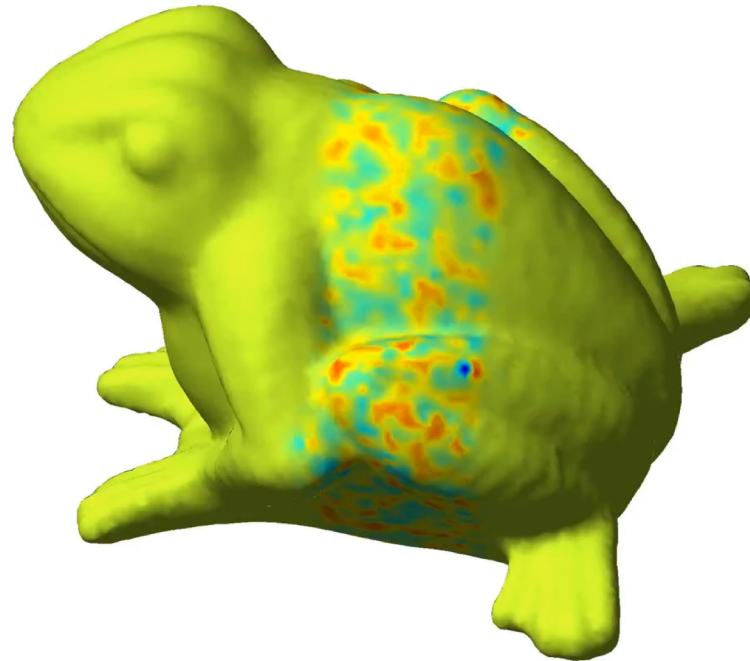
Poison dart frog



- RBF Node layout
- AIM@Shape Online Repository



Movie Courtesy of Grady Wright



Node generation algorithms

Iterative-Type Schemes

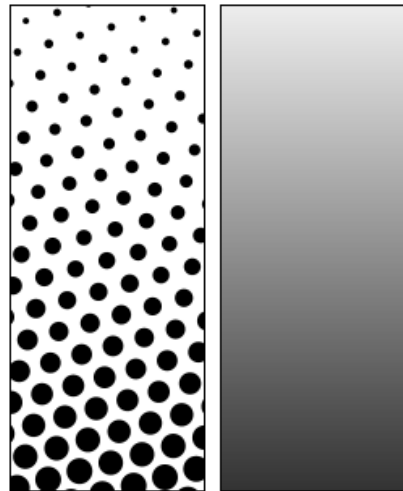
For a given number of nodes, the quality of the distribution depends on how soon the iteration is stopped.

- Min. energy distribution
- Voronoi relaxation
- Delaunay triangulations
 - DistMesh (Persson-Strang)
 - Gmsh (Geuzaine-Remacle)

Advancing-Front Type Schemes

For a given number of nodes, Start at a boundary and advance forward until the domain is filled.

- Dithering for half-tone images



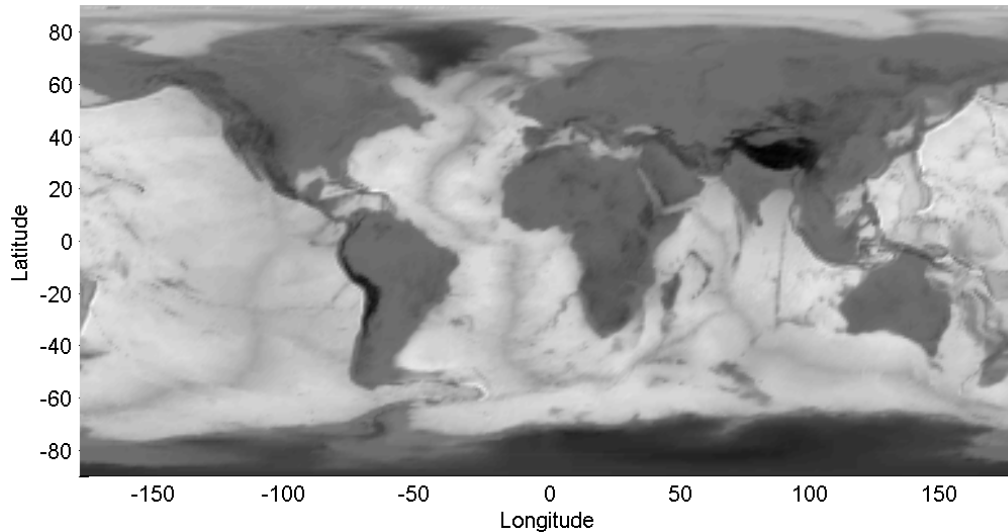
Notice nice hex pattern. Instead of changing width of dots, we change density of dots. We start at bottom boundary and march upward until domain is filled

Half-tone image Human eye

Distributing variable node density on sphere

(Fornberg and Flyer, 2015)

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

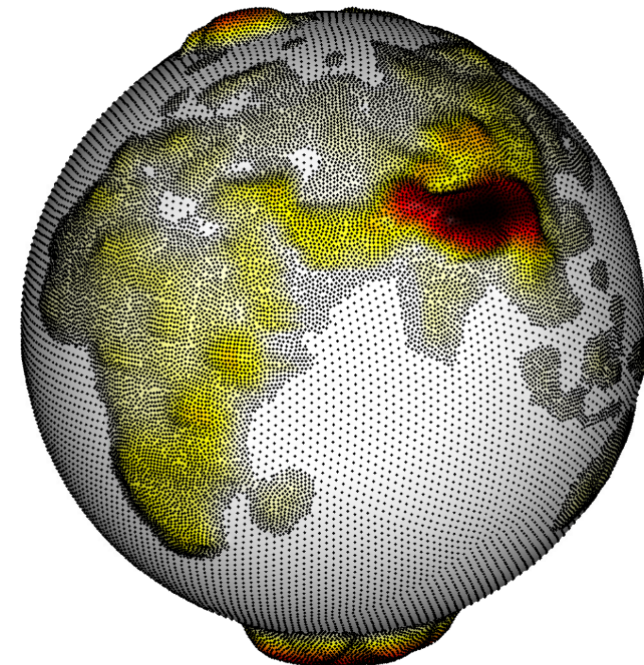
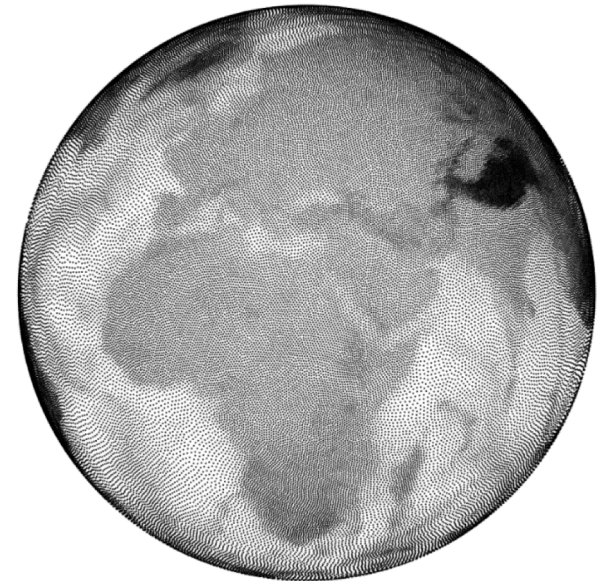


Top right: Advancing Front Algorithm

$N = 105,419$ 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) :

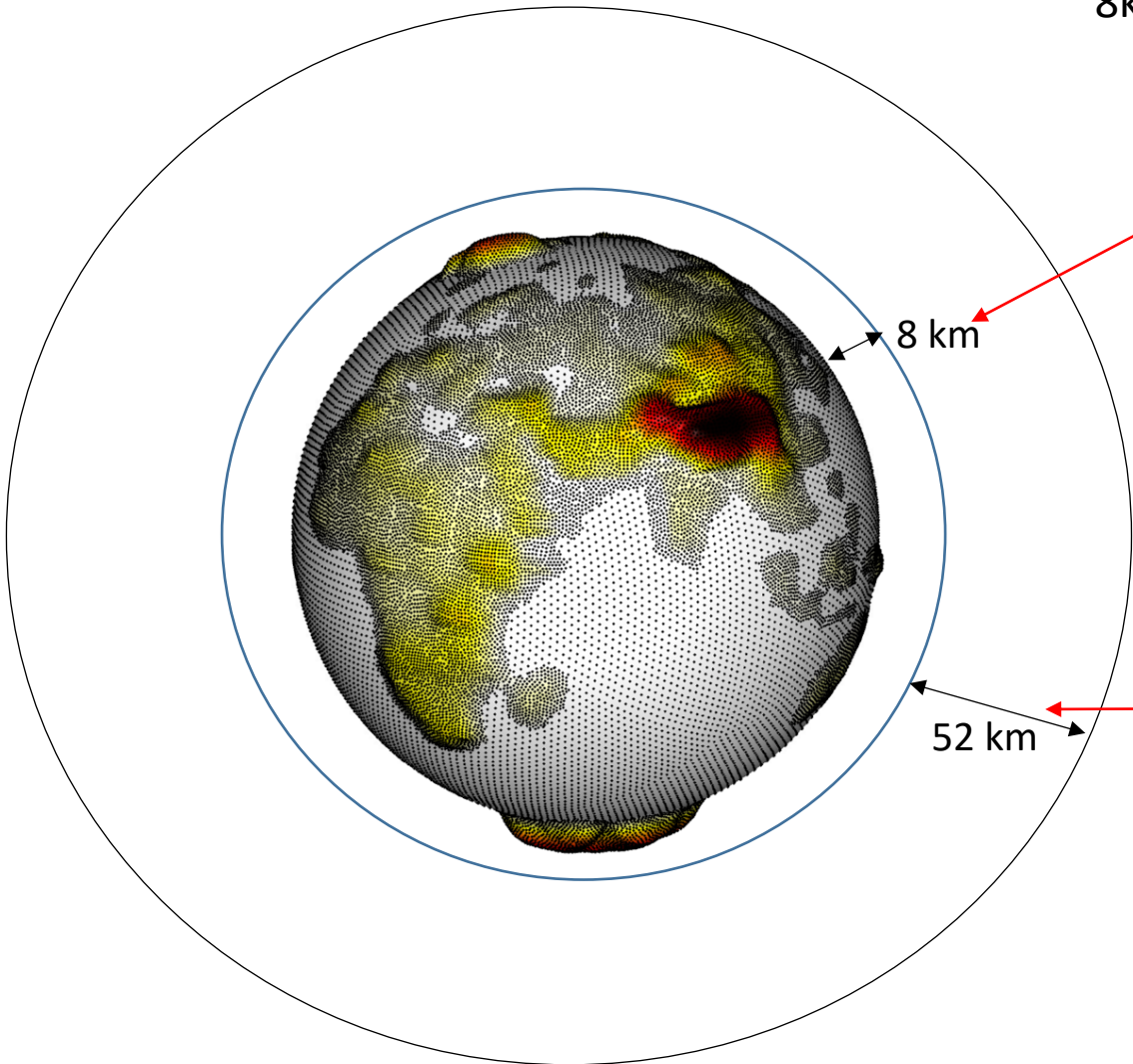
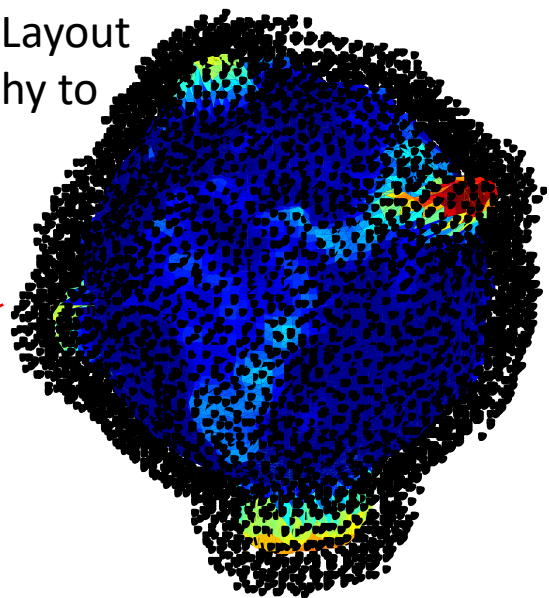
Take elevation physically taken into account



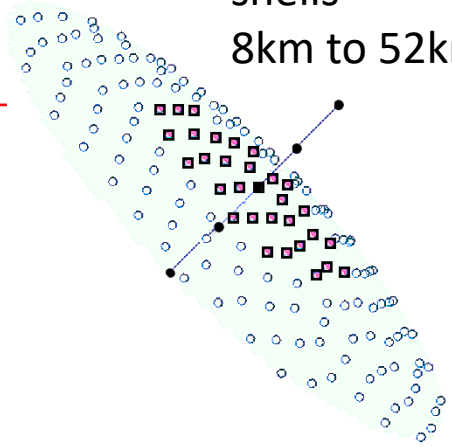
3D Elliptic PDE: Modeling Electrical Currents in the Atmosphere

$$\nabla_{3D} \cdot \underbrace{(\sigma(x,y,z))}_{\text{Conductivity}} \underbrace{\nabla_{3D} U}_{\text{Elec. potential}} = \underbrace{S(x,y,z)}_{\text{Thunderstorm sources (NASA measured data)}}$$

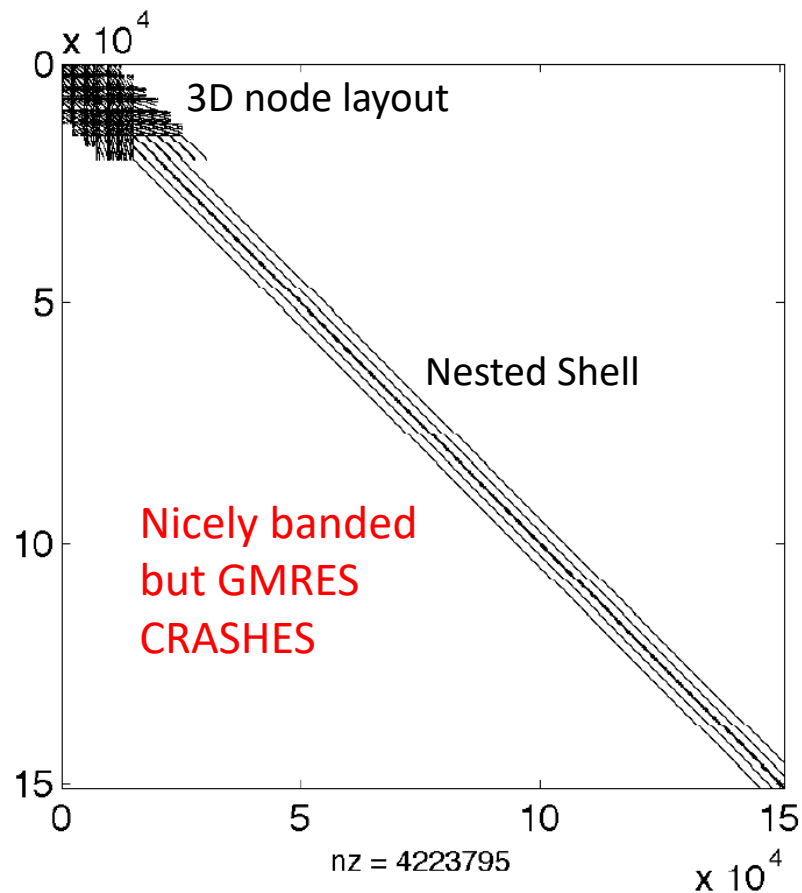
3D Node Layout
Topography to 8km



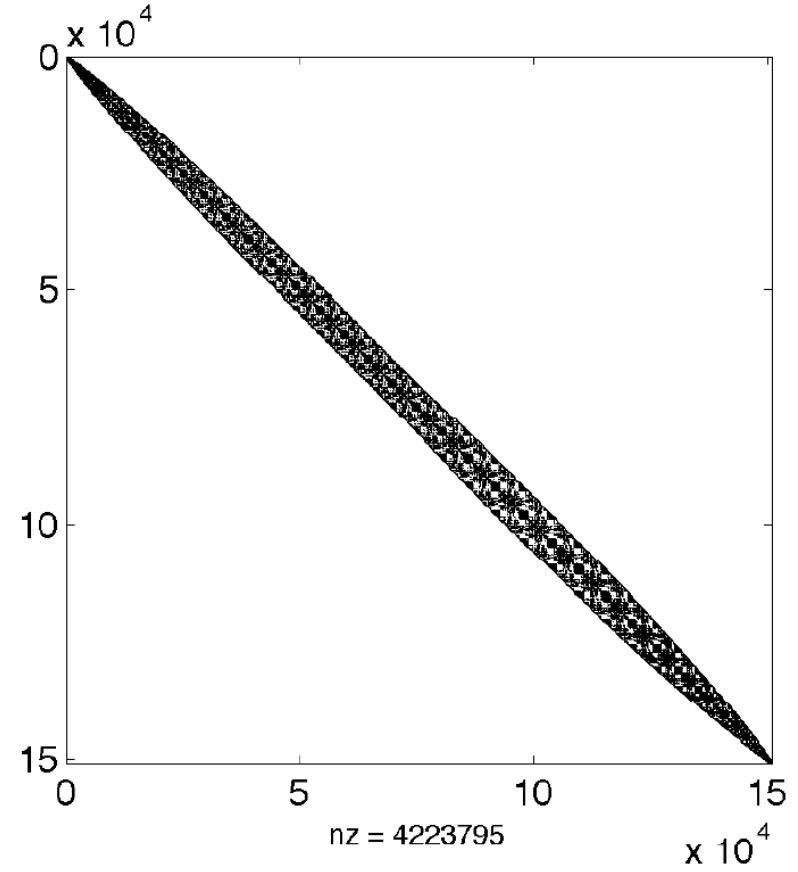
Nested spherical
shells
8km to 52km



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.

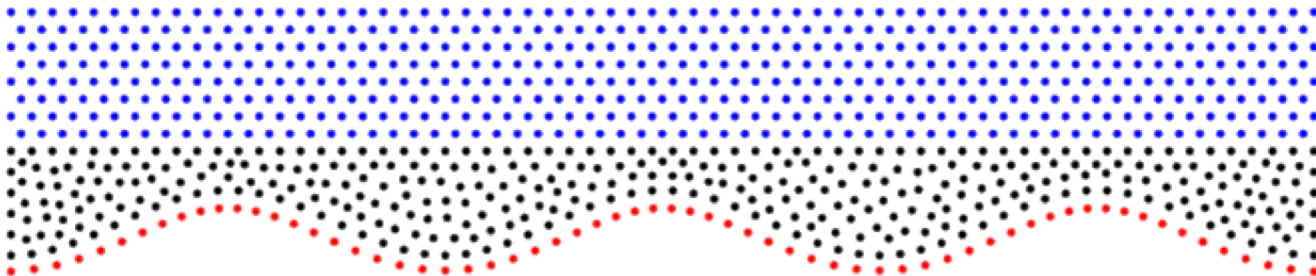
2D Compressible Navier-Stokes with Topography using RBF-FD

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

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

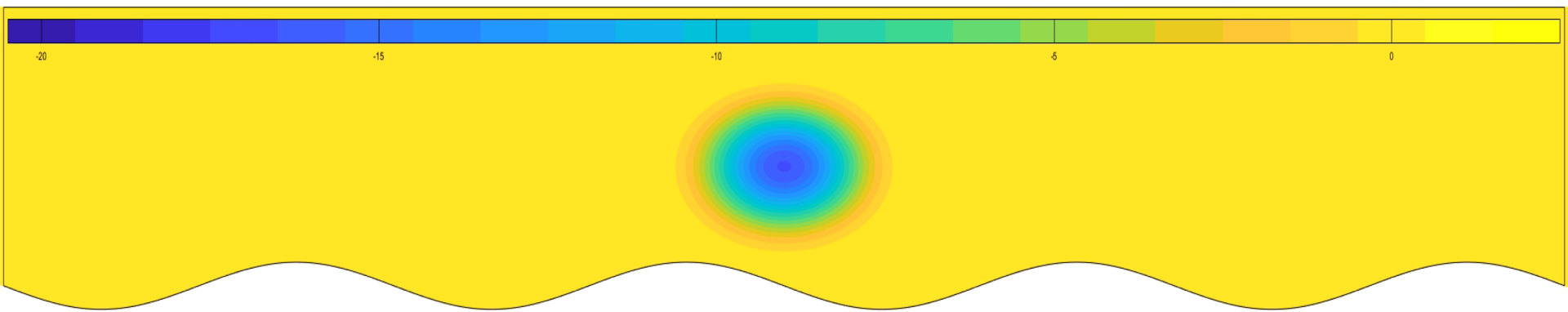
$$\frac{\partial P}{\partial t} = - (\mathbf{u} \cdot \nabla) P - \frac{R}{c_v} (\nabla \cdot \mathbf{u}) P, \quad \text{mass}$$

Schematic node layout




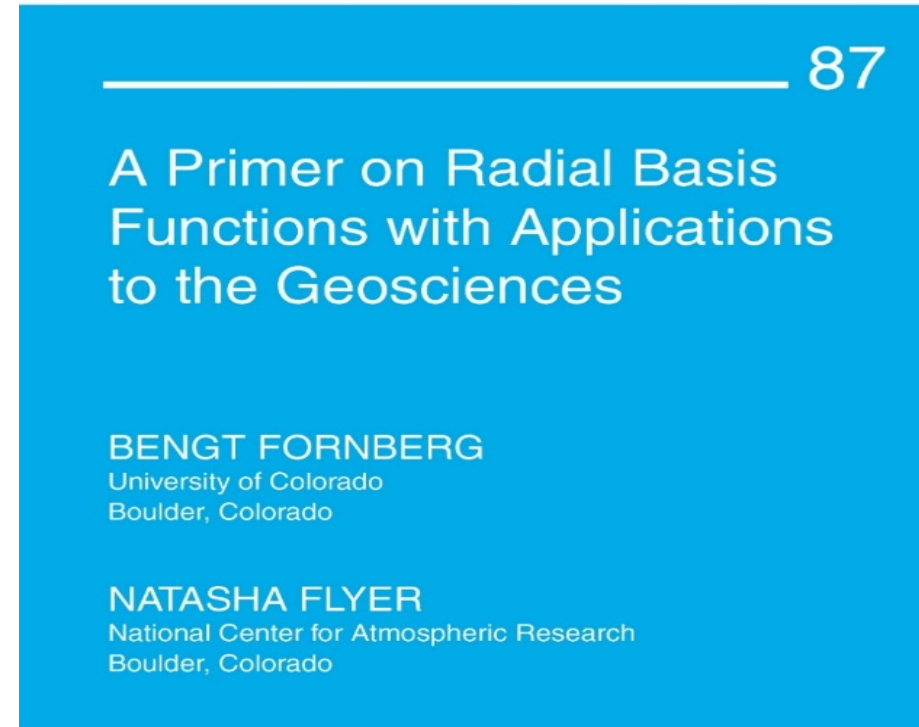
Movie Courtesy of Gregory A. Barnett

Simulation of a cold downdraught in a dry atmosphere at 300K



Recent Review Material for RBF

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. 



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