



On new approaches and ideas useful in design of atmospheric models

J. Pudykiewicz

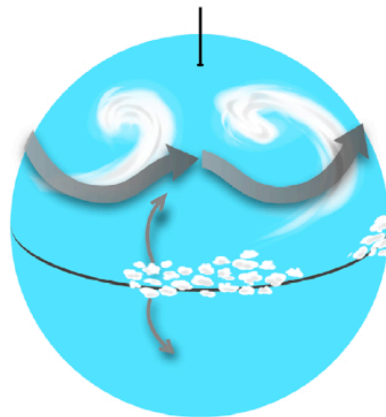
Atmosphere from the point of view of physics

rotating stratified fluid exhibiting both barotropic and baroclinic instabilities

complex energy transfer between the scales

*important interactions between waves and the mean flow;
turbulent and chaotic character*

*strong coupling between microphysical processes, fluid dynamics and
electromagnetic phenomena...*



1880

1900

1920

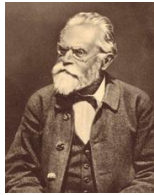
1940

1960

1980



**Empirical/
inductive**



v. Hann



V. Bjerknes



J. Bjerknes

*One hundred
years
of developments*



Exner



C. G. Rossby

**Theoretical/
deductive**



M. Margules



Charney

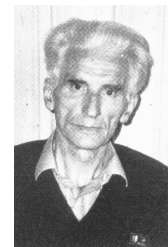


Richardson

Numerical



J. von Neumann



Fjortoft



G. Marchuk

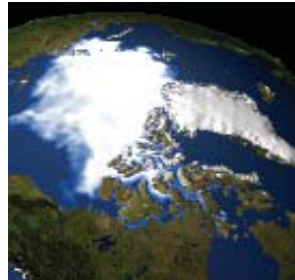
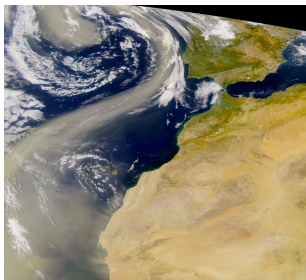


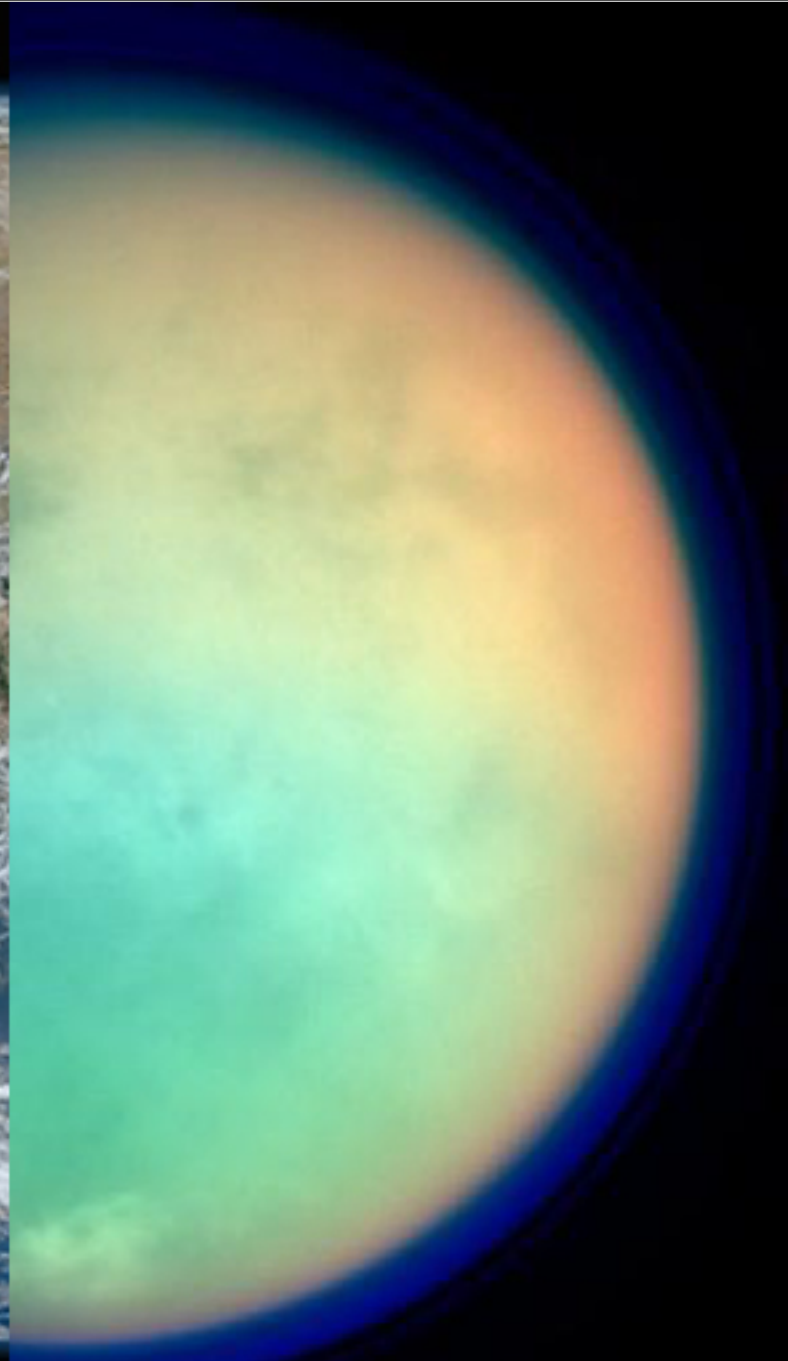
A. Robert

The developments in the dynamic meteorology, numerical methods and computer science led to the current highly developed models of the atmosphere

The ultimate weather prediction model is usually constructed upon the elastic system of equations and solved with semi - Lagrangian, semi-implicit method

Do we need to change this? Perhaps not, inevitable further mastering of the description of the unresolved scales should suffice to achieve the desired improvements of predictability...



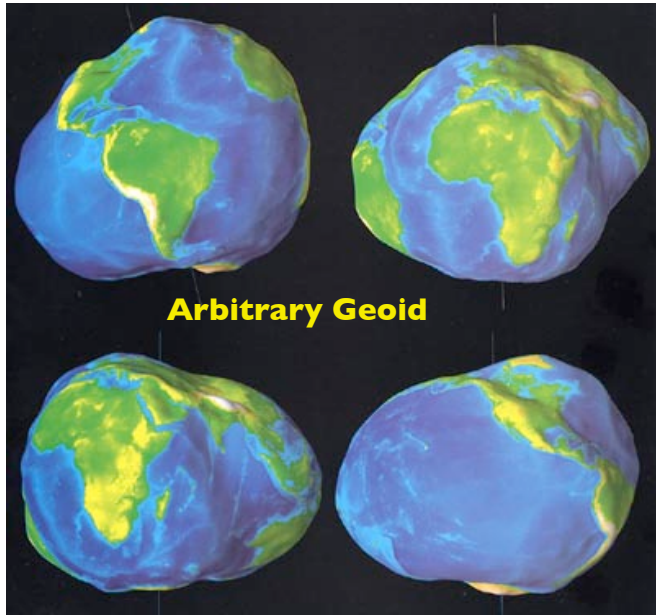


more
resolution?

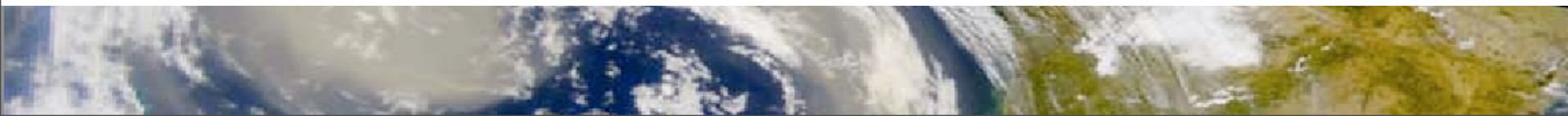
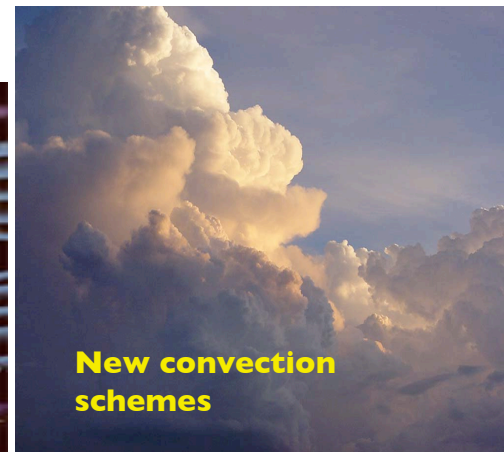
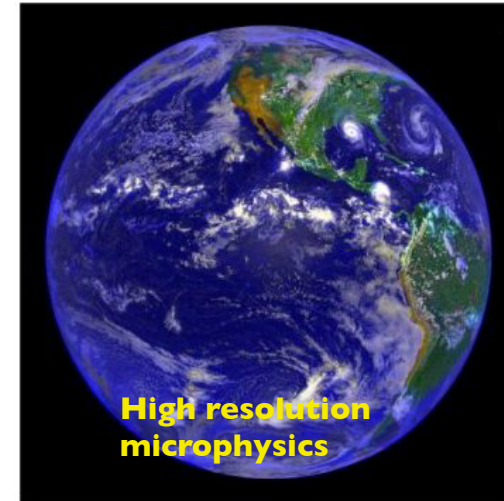
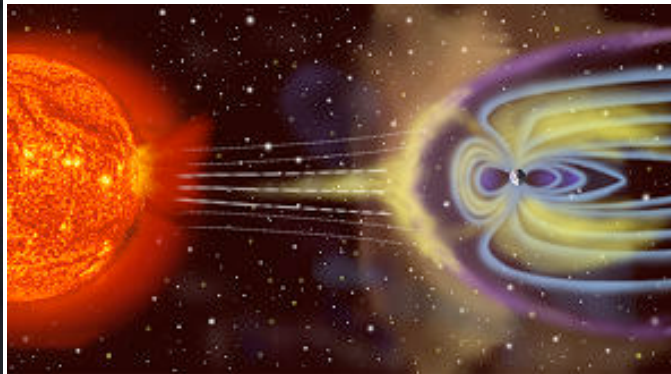
How clear is our understanding?

New Trends...

- *removal of shallow atmosphere approximation*
- *arbitrary geoid shape*
- *flexible spatial discretization*
- *new time integration schemes*
- *redefinition of the concept of parameterizations*
- *chemical and aerosol processes fully coupled with dynamics*
- *new forcings (electrodynamics, space weather data...)*



Solar-Terrestrial system and
revision of the Bjerknes-
Richardson paradigm
or the end of a hydraulic analogy



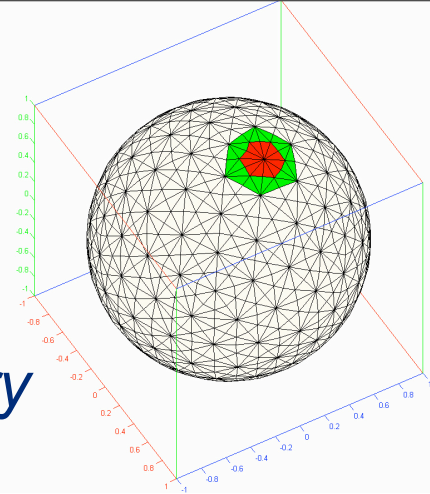
Flexible space discretization

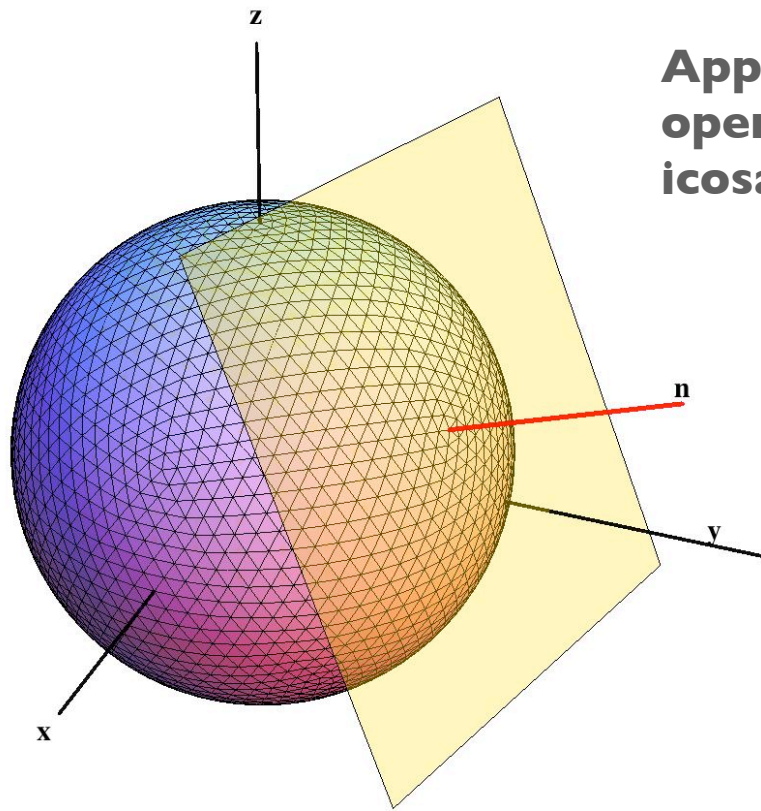
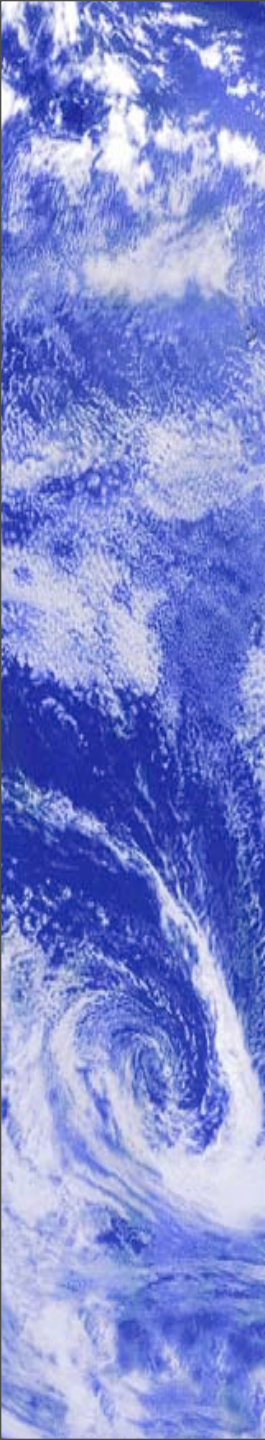
Formulation of the model on an arbitrary differentiable manifold

Strong theoretical background is provided by discrete differential geometry

Manifolds are represented via triangulation

Optimum triangulation could be established by the the minimization of the truncation error of the Laplace-Beltrami operator





Approximation of differential operators could be performed on icosahedral-geodesic grid

Poincare - Stokes-theorem:

$$\int_{\Omega} \mathbf{d}\omega = \int_{\partial\Omega} \omega$$



Semi-discretization...

$$\frac{dU}{dt} = F(U)$$

$$U(t) = \{u_1(t), \dots, u_m(t)\}^T$$

U {velocity, thermodynamic variables,
concentrations of chemical species...}

Multiple time scales, homoclinic and heteroclinic points

Chaotic behaviour of the system...

Unified approach to multiple problems in atmospheric models

$$\frac{dU}{dt} = F(U)$$

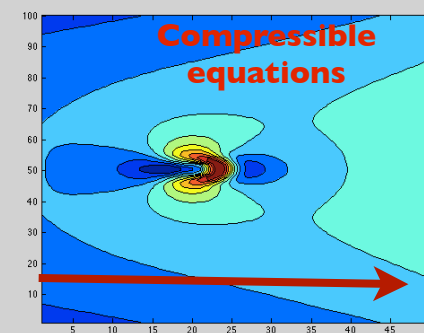
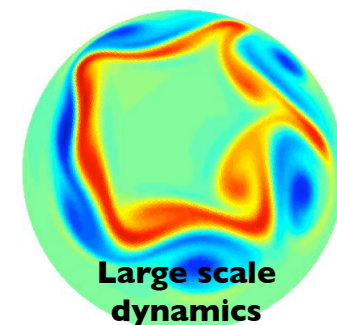
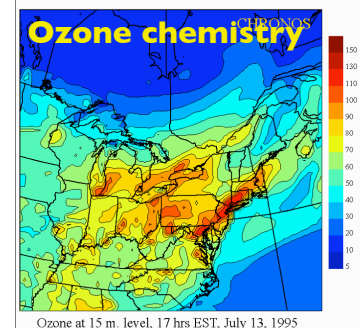
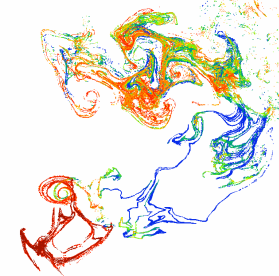
$$U(t) = \{u_1(t), \dots, u_m(t)\}^T$$

One of the first ideas was to separate the linear (stiff) and nonlinear terms

$$\frac{dU}{dt} = \mathcal{L}U + N(U(t))$$

Use the implicit scheme for the linear term and the explicit for the nonlinear part

Chaotic advection 600



It is much better to use the dynamic linearization

$$\frac{dU}{dt} = F(U)$$

$$U(t) = \{u_1(t), \dots, u_m(t)\}^T$$

$$\frac{dU}{dt}(t) = F_n + A_n(U(t) - U_n) + R(U(t))$$

$$R(U(t)) = F(U(t)) - F(U_n) - \frac{dF}{dU}(U_n)(U(t) - U_n)$$

$$A_n = \frac{dF}{dU}(U_n) \quad U_n = U(t_n) \quad F_n = F(U_n)$$

Jacobi operator

The use of the integrating factor $e^{-A_n t}$ yields

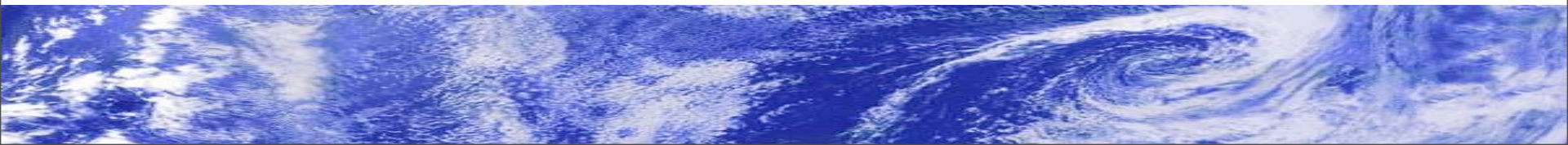
$$\frac{d}{dt}(e^{-A_n t} U(t)) = e^{-A_n t} (F_n - A_n U_n) + e^{-A_n t} R(U(t))$$

And integrating over $[t_n, t_n + h_n]$

followed by multiplication by $e^{A_n(t_n + h_n)}$

$$U(t_n + h_n) = U_n + (e^{A_n h_n} - I) A_n^{-1} F_n +$$

$$\int_{t_n}^{t_n + h_n} e^{A_n(t_n + h_n - t)} R(U(t)) dt$$

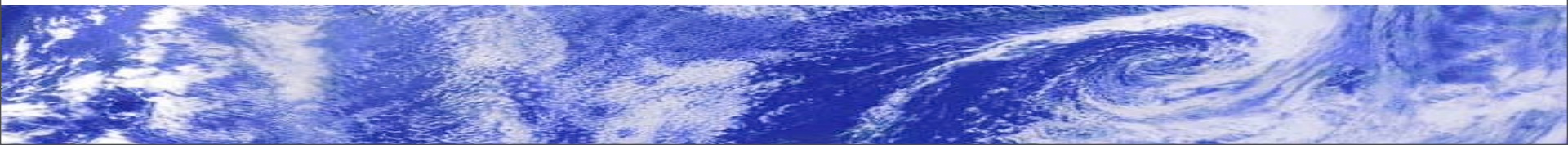


Depending on the quadrature used we will obtain different versions of the exponential integration schemes. Their common property is that the desired solution could be expressed as weighted sum of “phi functions”

$$\phi_0 = e^A \quad \phi_1 = \frac{e^A - I}{A} \quad \phi_2 = \frac{e^A - (I + A)}{A^2}$$

$$\phi_l(A) = A\phi_{l+1} + \frac{1}{l!}$$

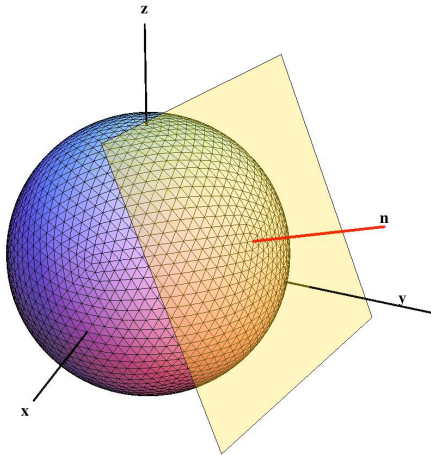
“phi functions” are evaluated using the software available in the literature



Advection equation

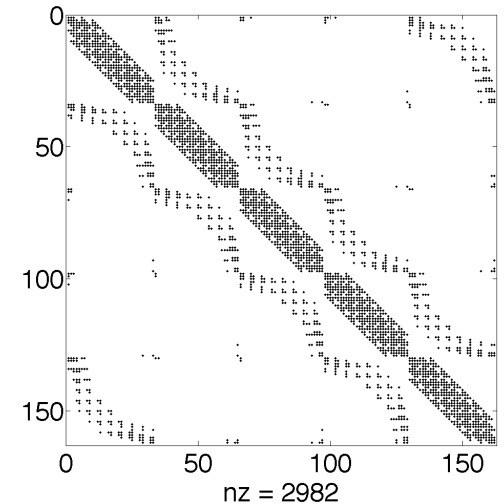
The simplest case for evaluation of the exponential propagation methods:

$$\frac{\partial U}{\partial t} + \nabla U \mathbf{v} = 0.$$



$$\frac{dU}{dt} = F(U)$$

$$U(t) = \{u_1(t), \dots, u_m(t)\}^T$$



Jacobi operator structure is determined by the connectivity matrix

The scheme selected for the integration

$$U_{n+1} = U_n + \phi_1(A_n h) h F_n + \frac{2}{3} \phi_2(A_n h) h R_{n-1}$$

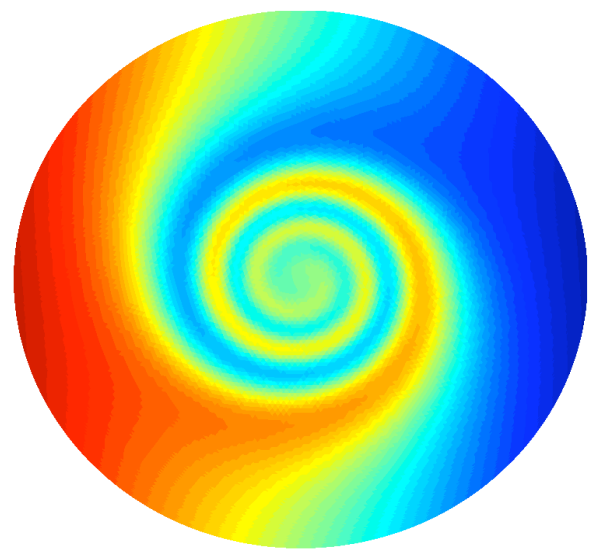
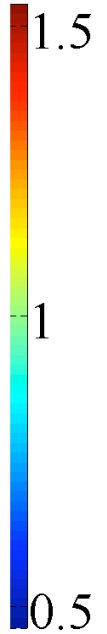
$$\phi_1(A_n h) = \frac{e^{A_n h} - I}{A_n h}$$

EPI3

$$\phi_2(A_n h) = \frac{e^{A_n h} - (I + A_n h)}{A_n^2 h^2}$$

$$R_{n-1} = F(U_{n-1}) - F(U_n) - A_n(U_{n-1} - U_n)$$

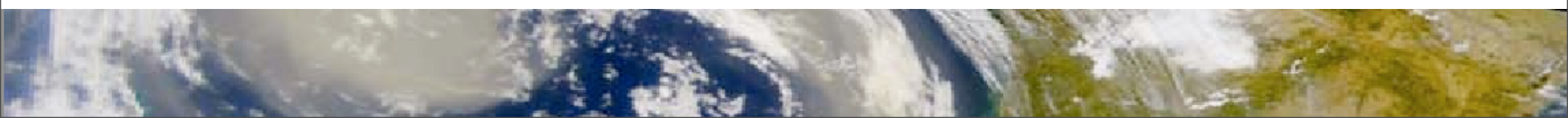
(f) Numerical solution 12 days



Deformation flow



Advection algorithm is mass conserving and stable for very large Courant numbers (much larger than SL methods)



Reaction-diffusion waves and pattern formation

When reacting species can diffuse on the sphere the system can be described by a coupled set of reaction-diffusion equations

$$\frac{\partial \phi}{\partial t} = f(\phi, \psi) + D_1 \nabla^2 \phi$$

$$\frac{\partial \psi}{\partial t} = g(\phi, \psi) + D_2 \nabla^2 \psi$$

One of the possible solutions of these system are so called reaction-diffusion waves first suggested by A. Turing in his seminal paper on morphogenesis (the concept is used also as an explanation for self-organization processes in physical, biological and economical systems).

In geophysics reaction-diffusion waves can be used to explain the distribution of chemical species in the environment, distribution of ice in the polar caps, surface properties important in the simulation of climatic system. The results could be obtained without an excessive calculations....

The method is often illustrated with so called bruselator system

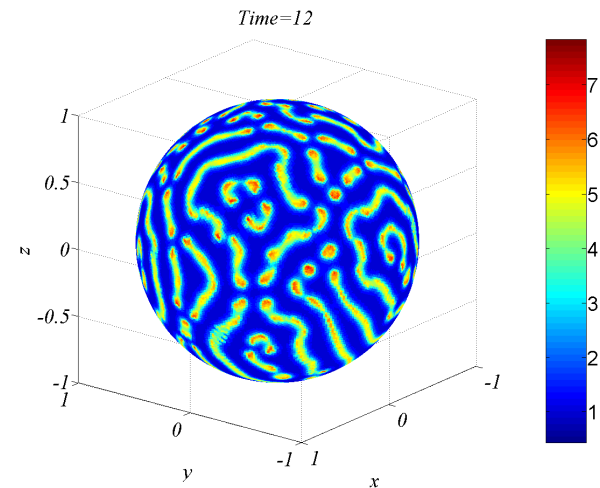
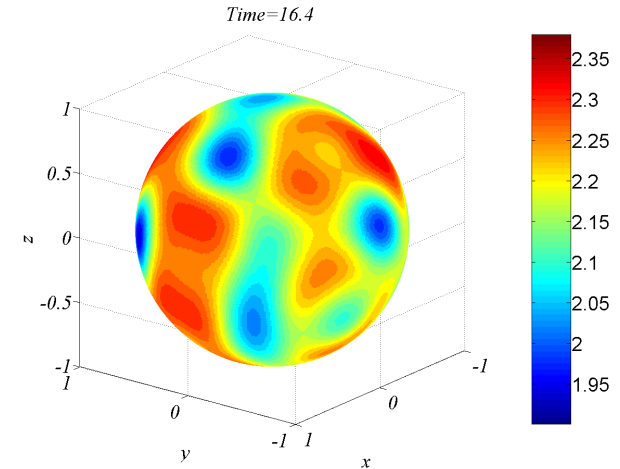
Oscillations (reactions-diffusion waves)

Complex eigenvalues of the Jacobian

Pattern formation

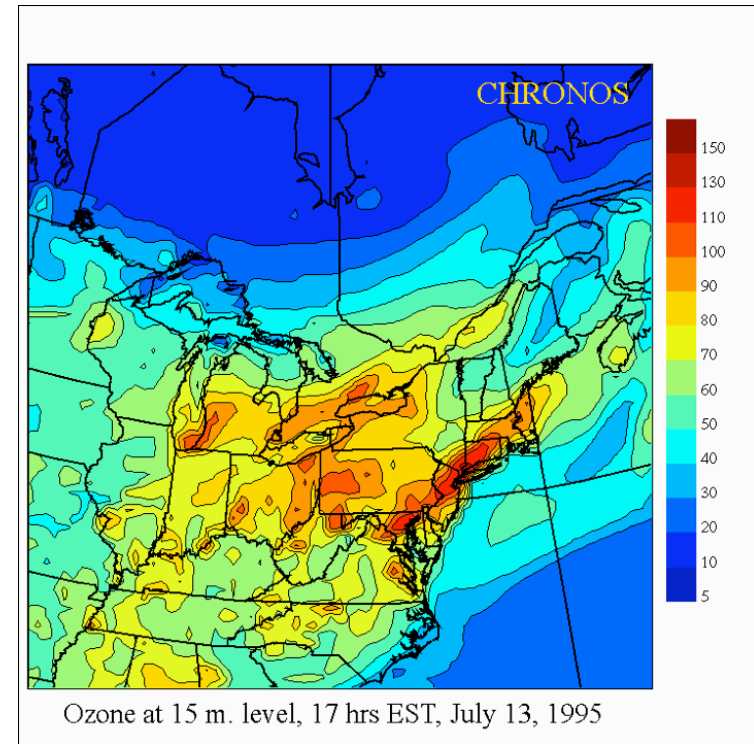
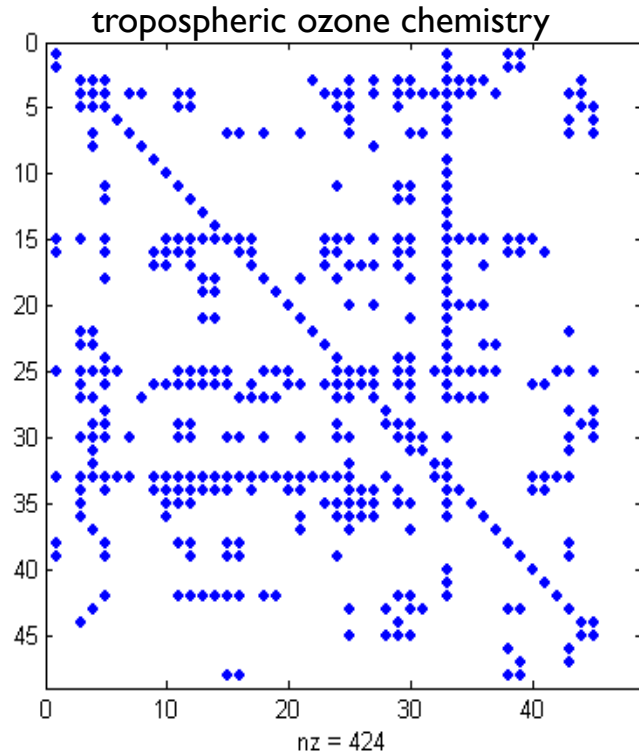
Diffusion combined with reactive terms leads to the evident sharpening of the patterns

negative eigenvalues of the Jacobian



Chemical Reactions of an Ozone and NO_x System

$$\partial_t \varphi^k = -\nabla \mathbf{u} \varphi^k + \nabla \mathbf{K} \nabla \varphi^k + F_c^k(\varphi^1, \dots, \varphi^{N_s})$$



The same method of solution could be used for an arbitrary system of chemical reactions after evaluation of the Jacobi operator (single routine for transport, mixing and chemistry)

Shallow water system

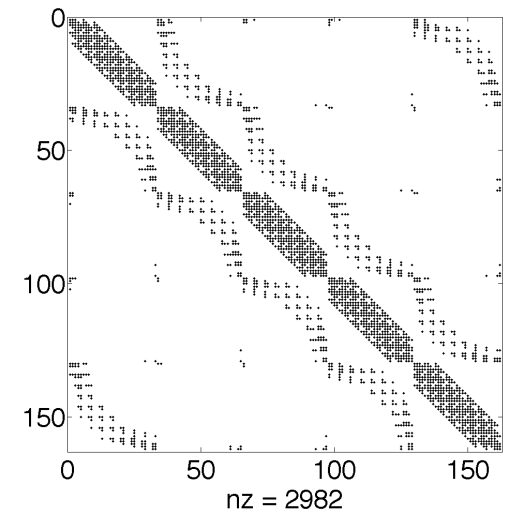
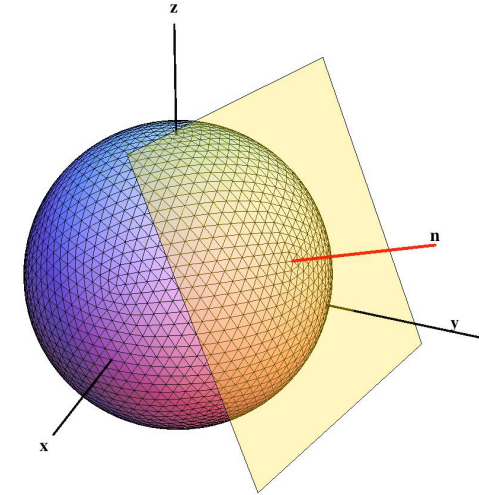
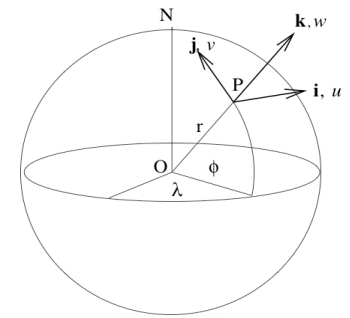
The shallow water equations in vector invariant form

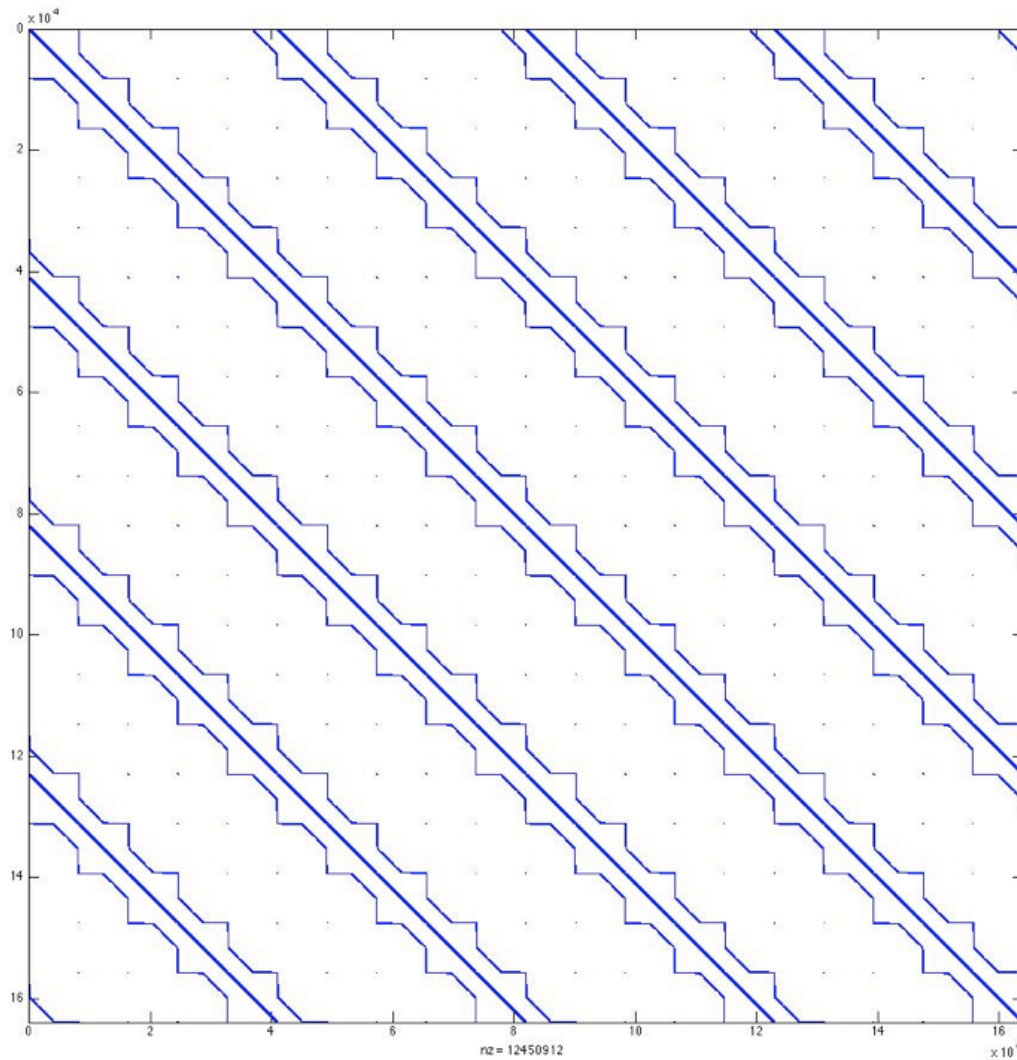
$$\frac{\partial \mathbf{v}}{\partial t} + (\zeta + f)(\mathbf{k} \times \mathbf{v}) = -\nabla(gh + \frac{\mathbf{v}\mathbf{v}}{2})$$

$$\frac{\partial h^*}{\partial t} + \nabla(h^* \mathbf{v}) = 0,$$

Semidiscrete autonomous system

$$\begin{cases} \frac{d}{dt}\{u^x\} = -\mathbf{W}^x - \mathbf{G}\mathbf{S}_x\{f\} \\ \frac{d}{dt}\{u^y\} = -\mathbf{W}^y - \mathbf{G}\mathbf{S}_y\{f\} \\ \frac{d}{dt}\{u^z\} = -\mathbf{W}^z - \mathbf{G}\mathbf{S}_z\{f\} \\ \frac{d}{dt}\{h^*\} = -\mathbf{D}_x\{u^x h^*\} - \mathbf{D}_y\{u^y h^*\} - \mathbf{D}_z\{u^z h^*\}, \end{cases}$$





nz = 12450912
sparse matrix (4n)x(4n)

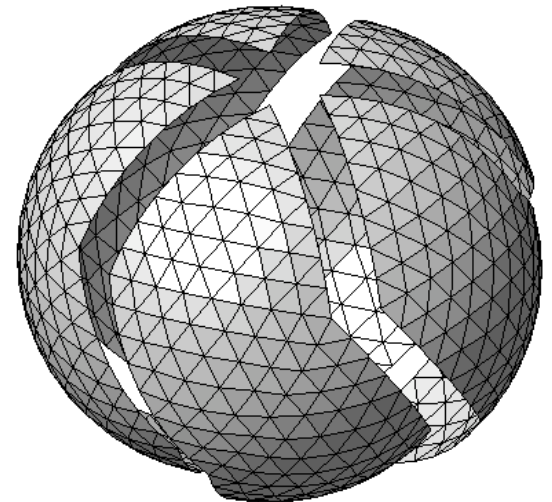
Jacobian for the shallow water system

The Jacobi operator:

$$\frac{\partial(F_x, F_y, F_z, G)^T}{\partial(u_x, u_y, u_z, h)}$$

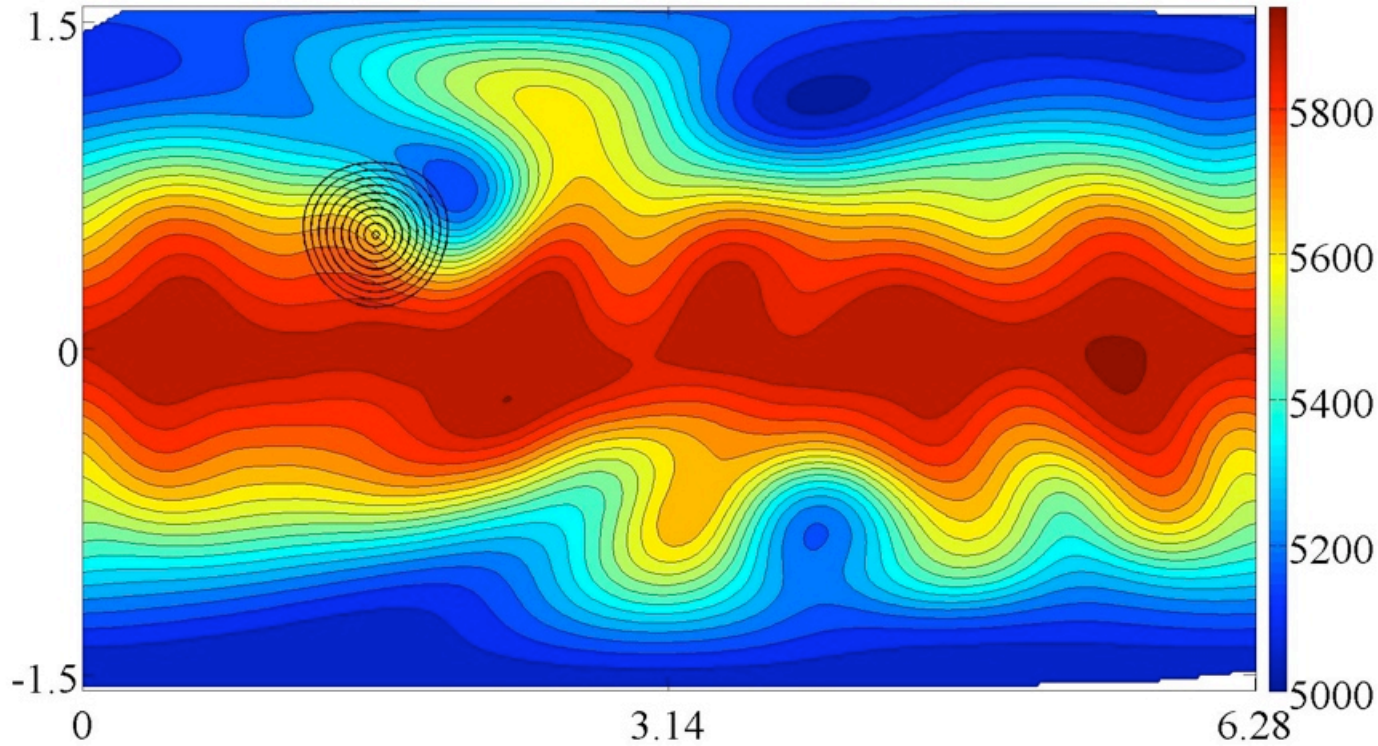
$$F_x^i = -(AGt_\gamma^{ij} u_\gamma^j + f^i) \cdot (N_y^i u_z^i - N_z^i u_y^i) - GS_x^{ij} \left(\frac{1}{2}(u_x^j{}^2 + u_y^j{}^2 + u_z^j{}^2) + g(h^j + h_s^j) \right)$$

$$G^i = -(D_x^{ij} h^j u_x^j + D_y^{ij} h^j u_y^j + D_z^{ij} h^j u_z^j)$$

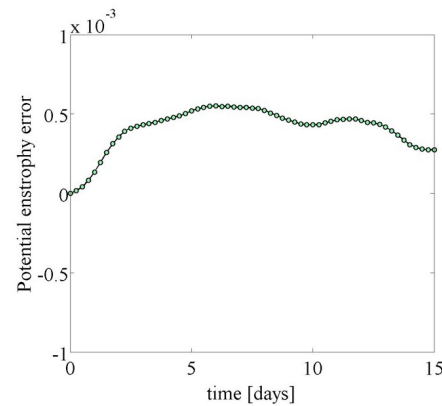
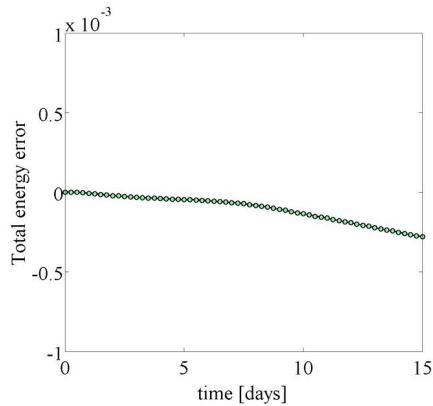


Zonal flow perturbed by mountain, grid number 6, dt=7200 sec.

Time=15 days



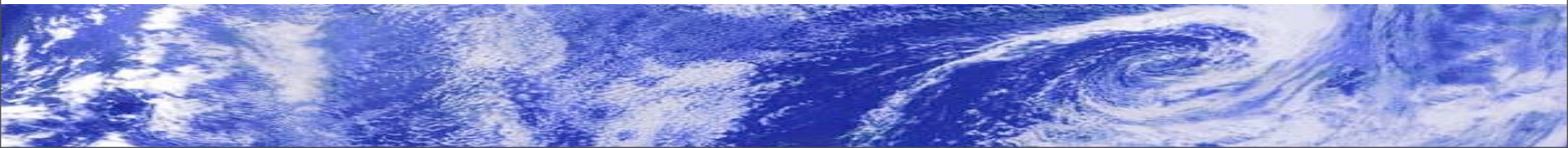
EPI3



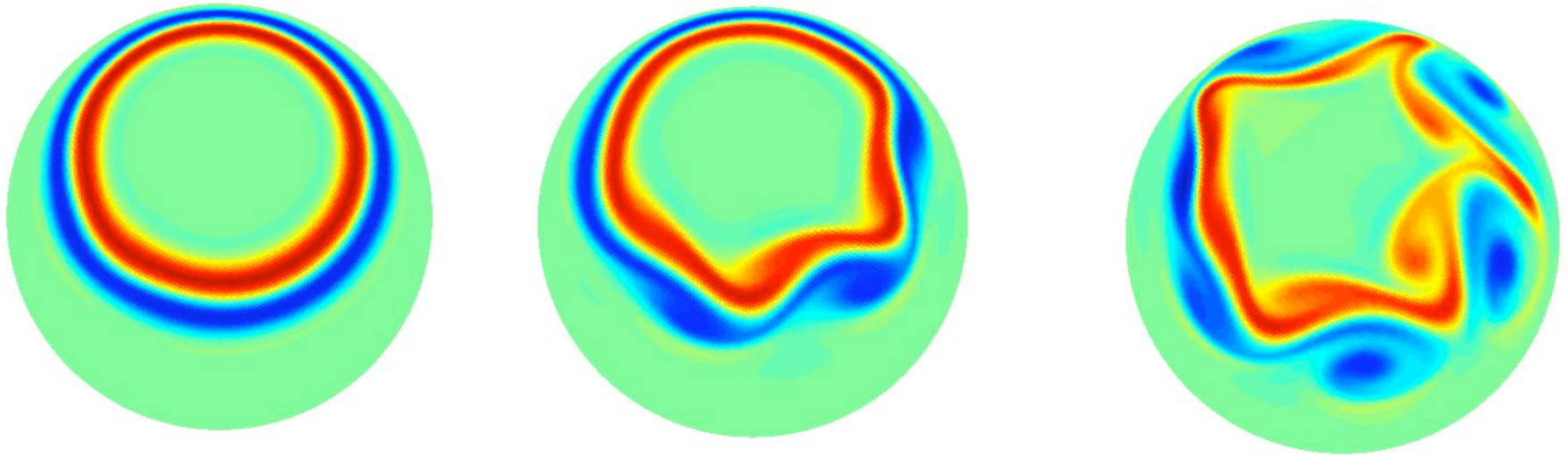
Advantages of the exponential integration method

The exponential methods are able to resolve high frequencies to the required level of tolerance without severe time step restriction of the explicit numerical schemes

The good resolution of the high frequencies with exponential methods is in contrast to usual implicit integration used with large time steps, which either damp high frequencies or map them to one and the same frequency (or nearly so) in the discretization



Unstable jet, grid number 6, $dt=7200$ sec



The main conclusion from the research performed with barotropic model was that although it is accurate the overall efficiency is not satisfactory...

Clancy C., Pudykiewicz J. (2013) On the use of exponential time integration methods in atmospheric models, Tellus A, vol. 65

Speed-up factor

$$\eta = \frac{T_{explicit}}{T_{exponential}} = \frac{4 T_d \left(\frac{\tau_s}{\Delta t}\right)}{T_d^e \left(\frac{\tau_s}{\Delta t_e}\right)} = 4 \frac{T_d}{T_d^e} \frac{\Delta t_e}{\Delta t}$$

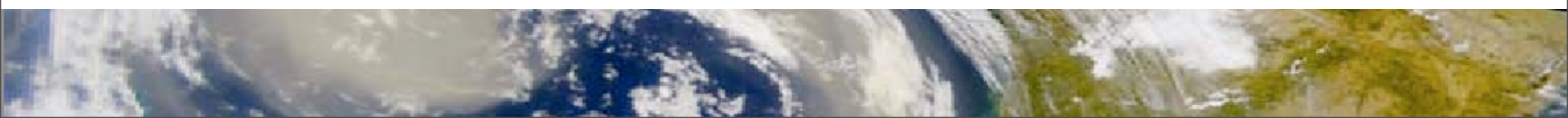
Time for evaluation of the:

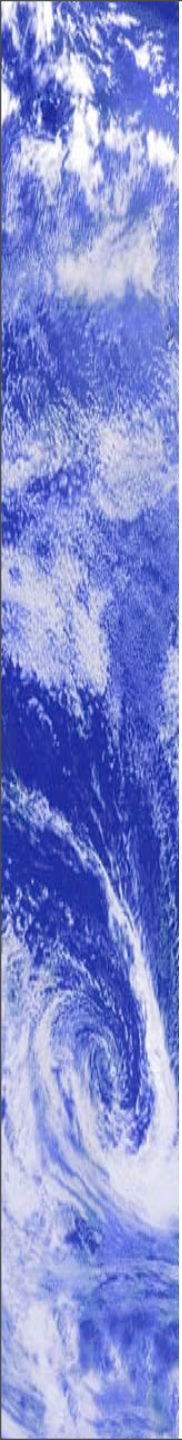
T_d - Right hand side

T_d^e " - phi functions" (EPI3)

Time steps

Δt - Runge-Kutta scheme, Δt_e - EPI3 scheme




$$\eta = 4\alpha\beta, \quad \alpha = \frac{T_d}{T_d^e}, \quad \beta = \frac{\Delta t_e}{\Delta t}$$

Original runs from October 2012:

$$T_d = 0.3, \quad T_d^e = 20, \quad \Delta t = 240, \quad \Delta t_e = 7200, \quad \eta = 1.8$$

Optimized version of the solver with phipm:

$$T_d = 0.3, \quad T_d^e = 8, \quad \Delta t = 240, \quad \Delta t_e = 7200, \quad \eta = 4.5$$

New version with phipmv will probably permit $\eta = 7$

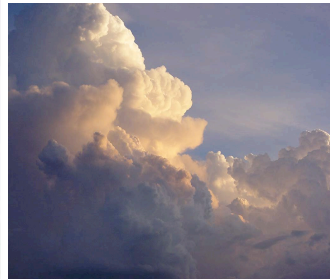
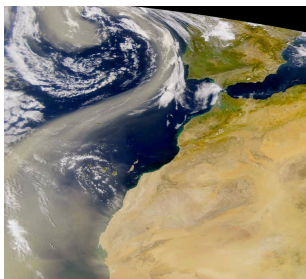
Icosahedral grid number 6 with 41000 nodes (MATLAB 2007); these numbers are smaller with new editions of MATLAB and much smaller when the code is converted to Fortran. Big gains are achieved with the fast sparse matrix vector multiplication procedures...Further work in multilevel model

Euler equations

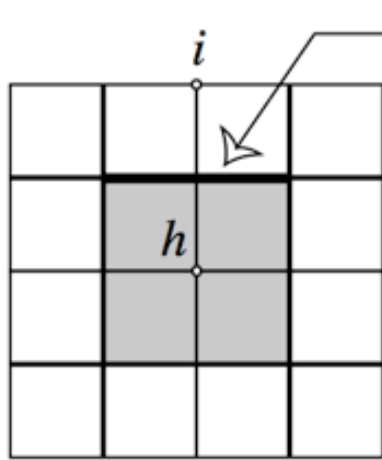
$$\frac{\partial \rho}{\partial t} + \nabla \rho \mathbf{v} = 0$$

$$\frac{\partial \rho \mathbf{v}}{\partial t} + \nabla \rho \mathbf{v} \otimes \mathbf{v} = -\nabla p - \rho g \mathbf{k} + 2\rho \boldsymbol{\Omega} \times \mathbf{v}$$

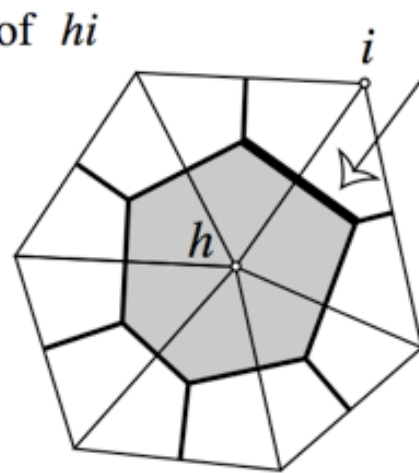
$$\frac{\partial \rho e}{\partial t} + \nabla h \rho \mathbf{v} = \mathbf{v} \nabla p + q_{heat}$$



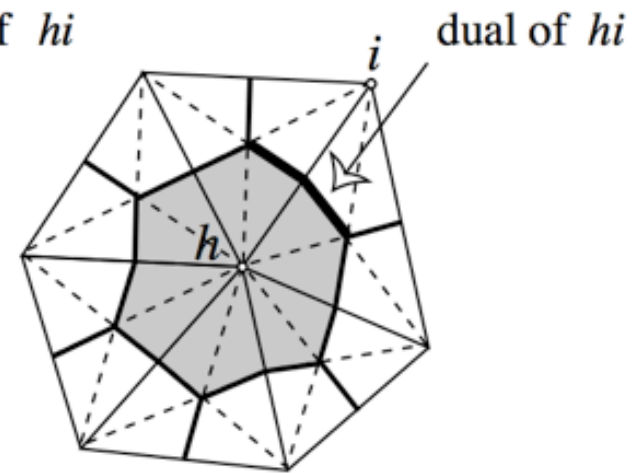
When generalizing the method to the compressible system in 3-D we consider multiple grid options



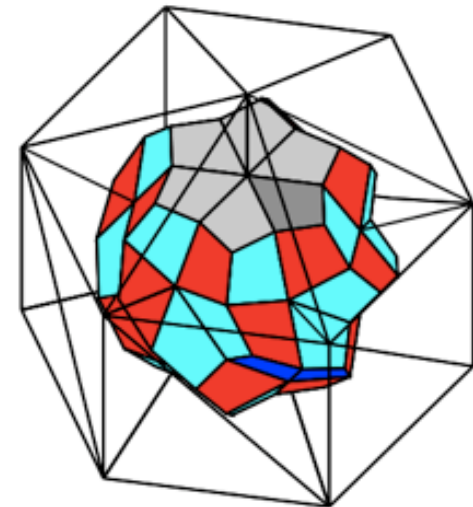
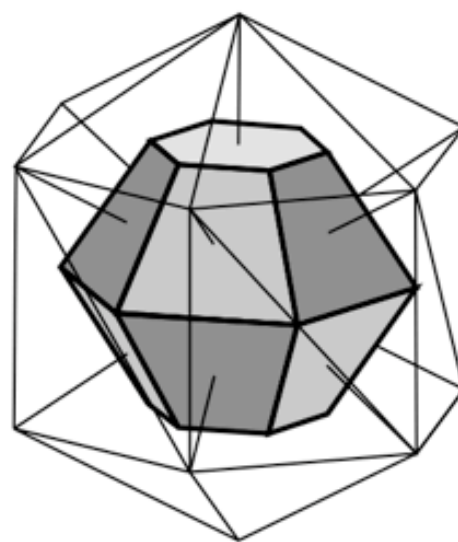
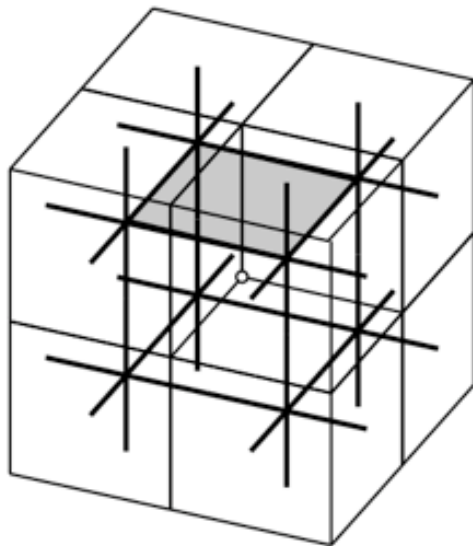
a) cartesian dual



b) Voronoi dual



c) barycentric dual



Equations used are from Robert (1993)

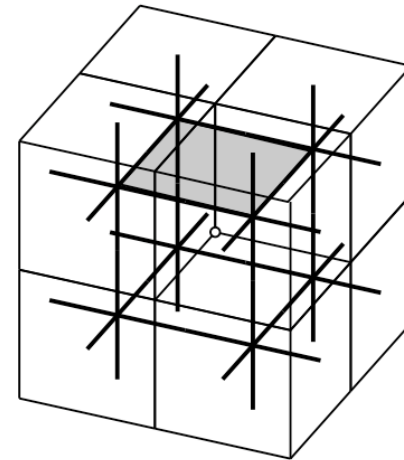
$$\frac{d\mathbf{C}}{dt} = \mathbf{B}$$

$$\mathbf{C} = (u, v, w, q', T')$$

$$B_1 = -RT \frac{\partial q'}{\partial x}$$

$$B_2 = -RT \frac{\partial q'}{\partial y}$$

$$B_3 = -RT \frac{\partial q'}{\partial z} + \frac{gT'}{T^*}$$

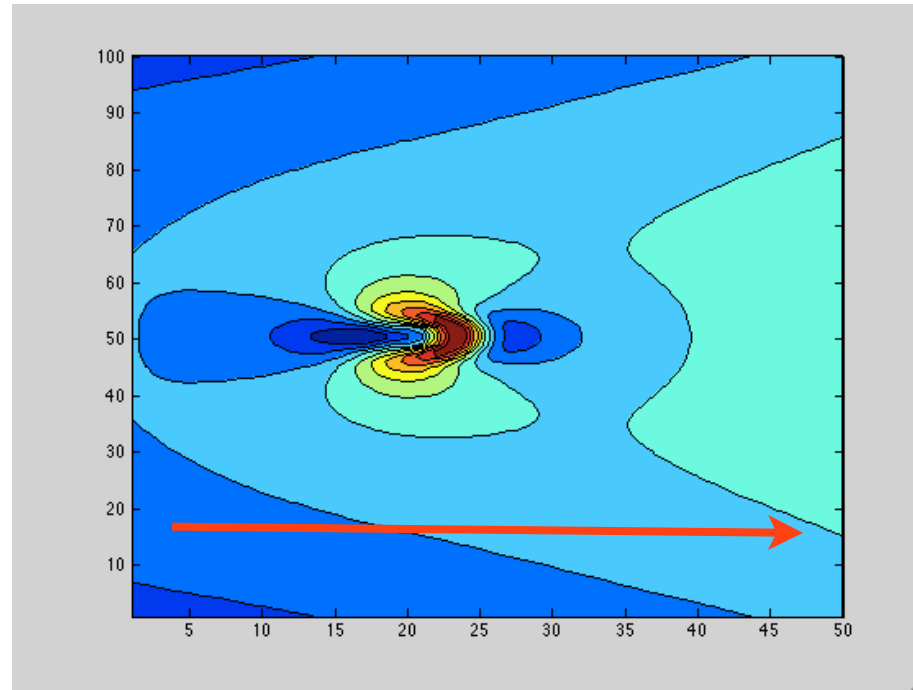
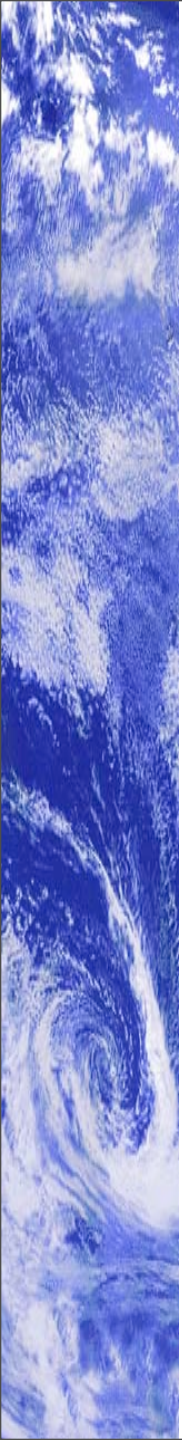


Cartesian grid is used...

$$B_4 = -\frac{D}{1 - \kappa} + \frac{g w}{RT^*}$$

$$B_5 = -\frac{\kappa T}{1 - \kappa} D$$

Notation is standard and it is same as in Robert (1993)



$$U_{n+1} = U_n + \phi_1(A_n h) h F_n$$

EPI2

Courant number 150

Boundary conditions enforced by relaxation

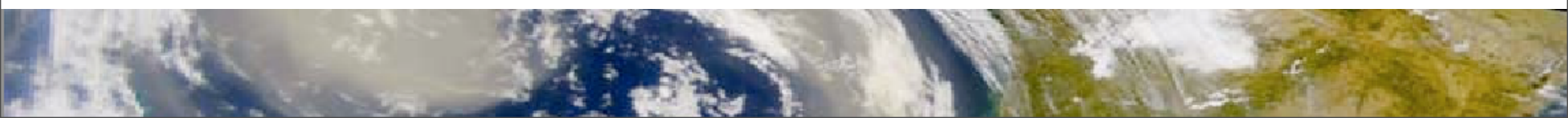
Finite differences or finite volumes

Major findings

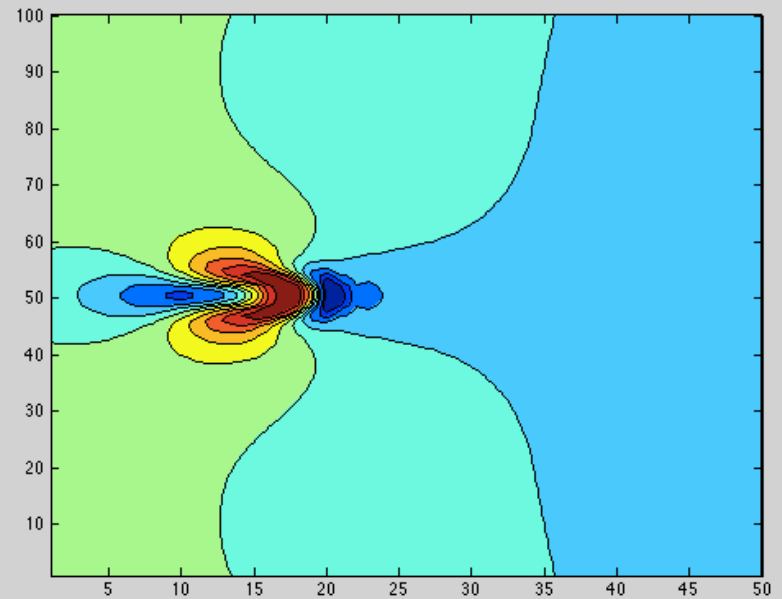
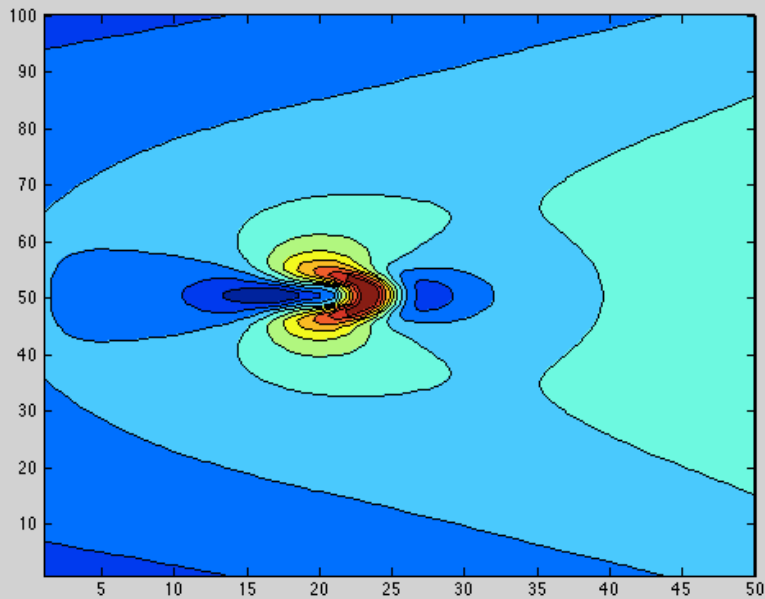
The scheme is free of noise even when the low order spatial discretization is combined with the exponential integration scheme

The method works very well on A-grid

Efficiency ratio with respect to the explicit solver is the same as in the barotropic model



The system is very sensitive to the relaxation procedure used to impose the boundary conditions



$$U_{n+1} = U_n + \phi_1(A_n h) h F_n$$

EPI2

relaxation applied at the top whereas on the left side was
no relaxation for w at the top

Conclusions

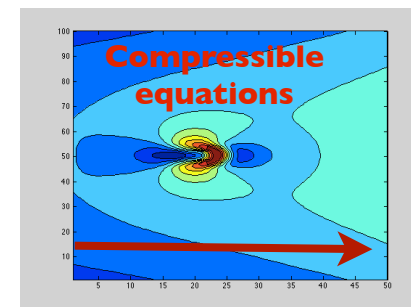
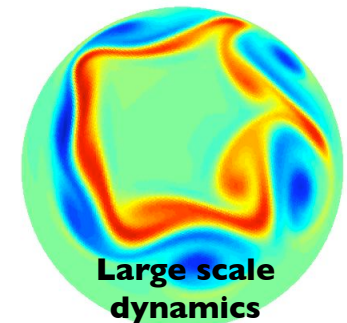
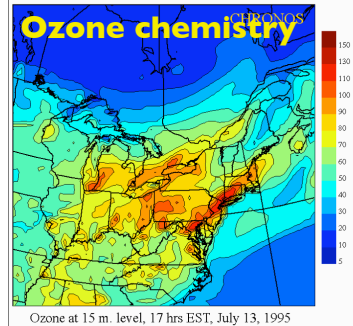
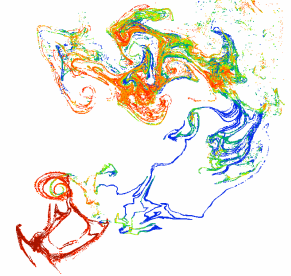
Combining the **Poincare-Stokes** theorem (on arbitrary polyhedral grid) with **exponential** integrators leads to the effective technique for solution of multiple problems including:

- advection
- chemical kinetics and transport
- shallow water equations on the sphere
- multilayer hydrostatic models on the sphere
- Euler equations

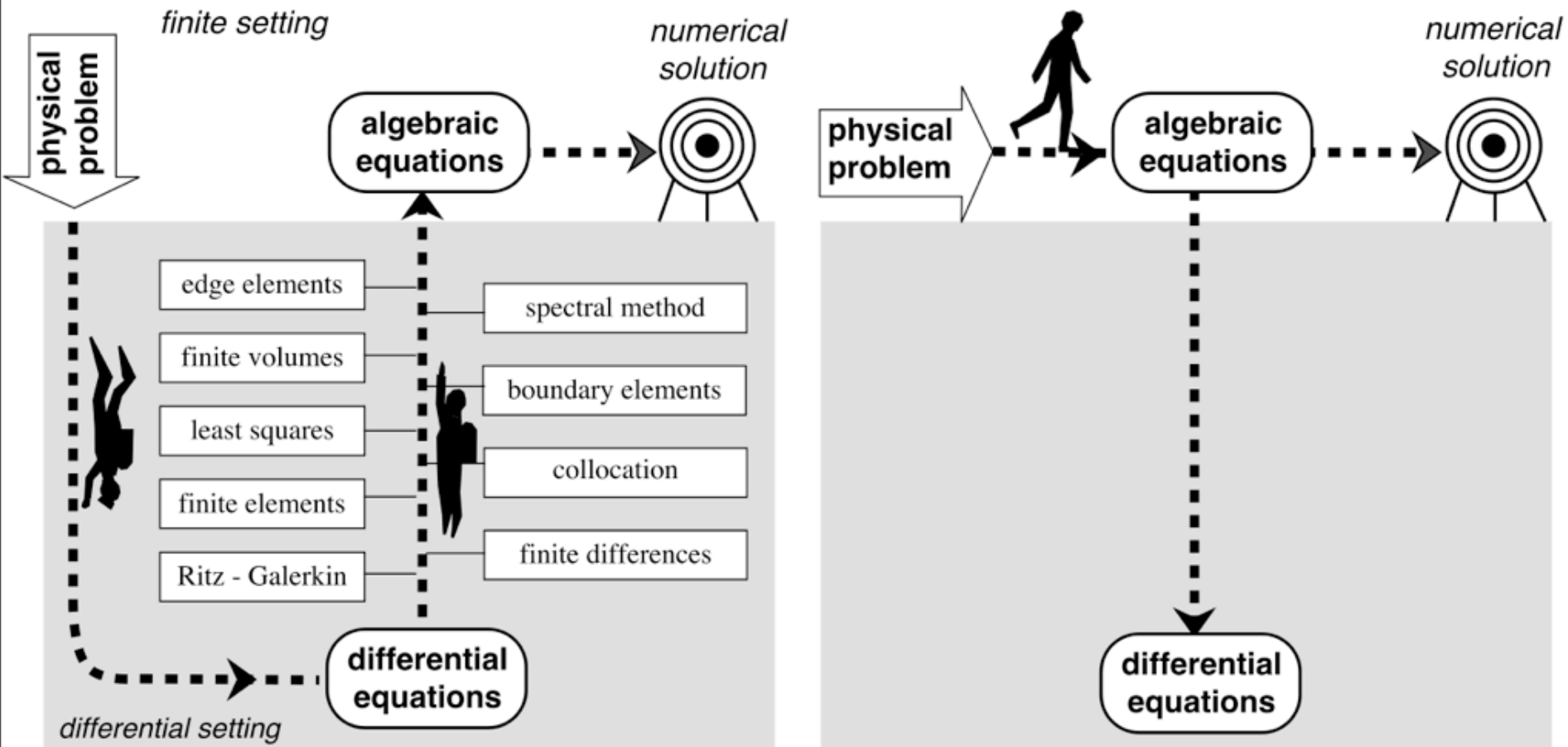
Theoretically all these problem are solved with one universal method

All codes have very good potential for parallel processing

Chaotic advection 600



After extensive use of Poincare Stokes theorem to discretize continuous problems one can ask: Are we really using partial differential equations in our models?



from E. Tonti (2014) Why starting from differential equations for computational physics?

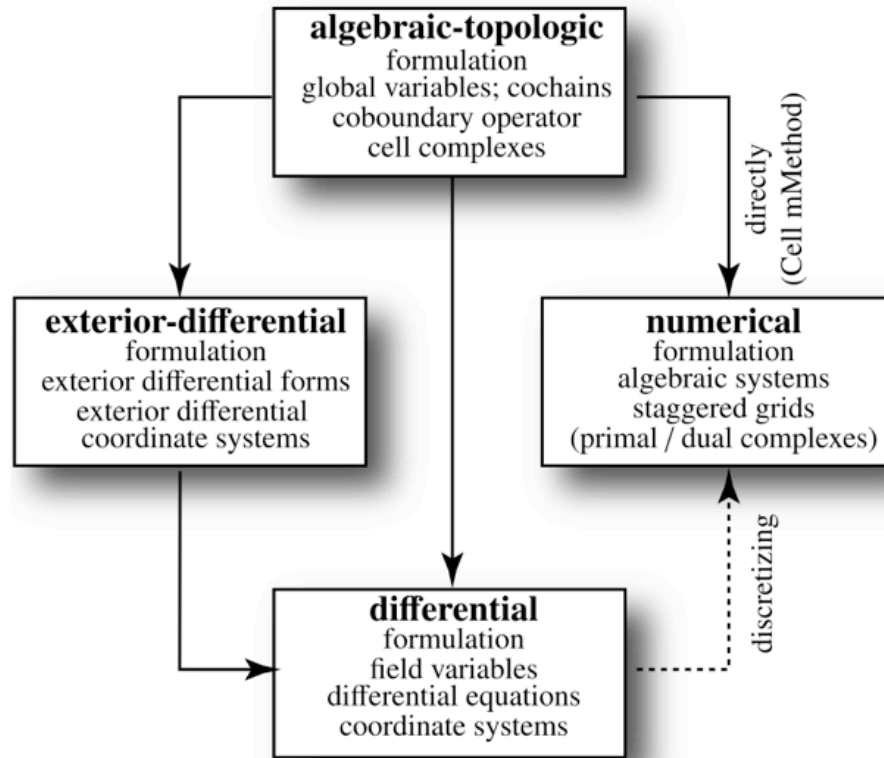
The issue of discretization in more general context



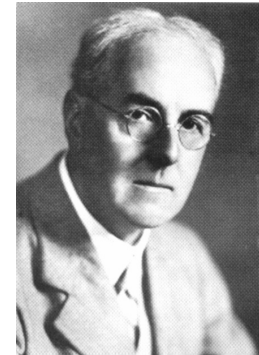
Poincare

from E. Tonti (2014) Why starting from differential equations for computational physics?

(I have selected names of scientist who I believe are most representative for the problem)



Cartan



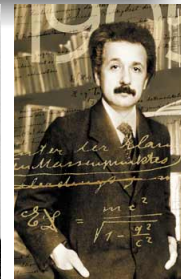
Richardson



Lagrange



Riemann



Einstein



Levi Civitta

A historical illustration of a forecast factory. The scene is set within a large, circular room with a curved wall. The wall is covered in a large-scale map of the world, showing continents like North America, South America, Europe, and Africa, and oceans like the Pacific and Indian. Several horizontal film strips are stretched across the room, each showing a different view of the globe. In the center, a tall, dark, cylindrical structure rises from the floor, topped with a platform where several figures are standing. At the base of this structure, a large group of people, mostly men in period clothing, are seated at desks, looking towards the center. The overall atmosphere is one of a busy, early 20th-century office or laboratory.

**Thank you for
your attention**

**Richardson's forecast factory,
drawing by F. Shuiten**