On the Design of Dynamical Cores for Atmospheric General Circulation Models (GCMs): Numerical and Scientific Challenges

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Overview of the 2 tutorials

Today:

- Scientific and numerical aspects in the design process
- Review different forms of the equations and variables
- Computational meshes, staggering options, the characteristics and accuracy of numerical discretizations
- Suitable time-stepping schemes and vertical coordinates

Overview of the 2 tutorials

Tomorrow:

- Physical and computational challenges: conservation, positive tracer advection, stability
- Computational aspects, computer architectures, scalability, efficiency, how it determines science decisions
- Hidden features in the design process: subgrid-scale diffusive and filtering processes
- Evaluation techniques and selected dynamical core test cases

The pursuit of the 'perfect' dynamical core: Design aspects

Our scientific and numerical wish list:

- Accurate
- Stable
- Simple
- Computationally efficient
- Obeys physical constraints: conservation properties (which ones?), positive-definite tracer advection
- Truthful representation of the subgrid-scale

What is a dynamical core?

- Fluid dynamics component of every weather or climate model
- Based on equations of motion, they may be approximated
- Describes the resolved adiabatic motions on a computational grid
- Contains filters and diffusion processes, mostly for numerical purposes, physical justification may be weak
- Determines the choice of the prognostic (forecast) variables

Components of an Atmospheric General Circulation Model (AGCM)



Modular design of NASA's General Circulation Model GEOS-5



Does the dynamical core matter?

- Provocative: the fluid dynamics problem is solved, physics parameterizations matter most
- Let's take a look at 9 dynamical cores that participated in an intercomparison project during the 2008 NCAR Summer Colloquium



Mountain-triggered Rossby waves



0

-10

10

20

30

40

m/s

700 hPa zonal wind at day 15 (≈1°×1°L26)

Mountain-triggered Rossby waves



Details matter, diffusion matters

700 hPa zonal wind at day 25 (≈1°×1°L26)



The choice of the equations of motion

- The governing equations are the 3D Euler equations, but we never use them in their original form
- We make **simplifications** (e.g. the Earth is a perfect sphere) and use scaling arguments to simplify the dynamical core design
- The Euler equations contain 6 equations:
 - Three momentum equations
 - Continuity equation (mass conservation)
 - Thermodynamic equation
 - Ideal gas las
- 6 equations, 6 unknowns: u,v,w,T,p,ρ

Choice of the Equations: Common design decisions

- Deep or shallow atmosphere: is the distance 'r' to the center of the Earth represented as the constant radius 'a'?
- Hydrostatic or non-hydrostatic: is forecast equation for w maintained?
- Filtered equations? Anelastic, Boussinesq, pseudo-incompressible, unified
- Which prognostic variables are suitable?
- Which **coordinate system** is suitable:
 - Spherical coordinates
 - Local coordinates, Cartesian coordinates

Non-hydrostatic equations of motion (deep atmosphere, spherical coordinates)

$$\frac{Du}{Dt} - \frac{uv\tan(\phi)}{r} + \frac{uw}{r} = -\frac{1}{\rho r\cos\phi} \frac{\partial p}{\partial \lambda} + 2\Omega v \sin(\phi) - 2\Omega w \cos(\phi) + v \nabla^{2}(u)$$

$$\frac{Dv}{Dt} + \frac{u^{2}\tan(\phi)}{r} + \frac{vw}{r} = -\frac{1}{\rho r} \frac{\partial p}{\partial \phi} - 2\Omega u \sin(\phi) + v \nabla^{2}(v)$$

$$\frac{Dw}{Dt} - \frac{u^{2} + v^{2}}{r} = -\frac{1}{\rho} \frac{\partial p}{\partial r} - g + 2\Omega u \cos(\phi) + v \nabla^{2}(w) \quad \text{with variable } g = \frac{d\Phi}{dr} = G \frac{a^{2}}{r^{2}}$$

$$\frac{D\rho}{Dt} + \frac{\rho}{r\cos\phi} \left[\frac{\partial u}{\partial \lambda} + \frac{\partial (v\cos\phi)}{\partial \phi} \right] + \frac{\rho}{r^{2}} \frac{\partial (r^{2}w)}{\partial r} = 0$$

$$c_{v} \frac{DT}{Dt} + p \frac{D}{Dt} (\frac{1}{\rho}) = J \quad \text{Only approximation:}$$

$$p = \rho RT \quad \text{Earth is a perfect sphere}$$

Quasi-hydrostatic equations of motion (deep atmosphere, spherical coordinates)

form

 $=G\frac{a^2}{r^2}$

$$\frac{Du}{Dt} - \frac{uv\tan(\phi)}{r} + \frac{uw}{r} = -\frac{1}{\rho r\cos\phi} \frac{\partial p}{\partial \lambda} + 2\Omega v \sin(\phi) - 2\Omega w \cos(\phi) + v\nabla^{2}(u)$$

$$\frac{Dv}{Dt} + \frac{u^{2}\tan(\phi)}{r} + \frac{vw}{r} = -\frac{1}{\rho r} \frac{\partial p}{\partial \phi} - 2\Omega u \sin(\phi) + v\nabla^{2}(v)$$

$$\frac{Dw}{Dt} - \frac{u^{2} + v^{2}}{r} = -\frac{1}{\rho} \frac{\partial p}{\partial r} - g + 2\Omega u \cos(\phi) + v\nabla^{2}(w) \quad \text{with variable } g = \frac{d\Phi}{dr} = G\frac{a^{2}}{r^{2}}$$

$$\frac{D\rho}{Dt} + \frac{\rho}{r\cos\phi} \left[\frac{\partial u}{\partial \lambda} + \frac{\partial(v\cos\phi)}{\partial \phi} \right] + \frac{\rho}{r^{2}} \frac{\partial(r^{2}w)}{\partial r} = 0$$

$$c_{v} \frac{DT}{Dt} + p\frac{D}{Dt} (\frac{1}{\rho}) = J$$

$$p = \rho RT$$

$$\frac{D()}{Dt} = \frac{\partial()}{\partial t} + \frac{u}{r\cos\phi} \frac{\partial()}{\partial \lambda} + \frac{v}{r} \frac{\partial()}{\partial \phi} + w \frac{\partial()}{\partial r}$$

$$\frac{\partial(v)}{\partial t} = \frac{\partial(v)}{\partial t} + \frac{u}{r\cos\phi} \frac{\partial(v)}{\partial t} + \frac{v}{r} \frac{\partial(v)}{\partial \phi} + w \frac{\partial(v)}{\partial r}$$

$$\frac{\partial(v)}{\partial t} = \frac{\partial(v)}{\partial t} + \frac{u}{r\cos\phi} \frac{\partial(v)}{\partial t} + \frac{v}{r} \frac{\partial(v)}{\partial \phi} + w \frac{\partial(v)}{\partial r}$$

Shallow atmosphere approximation

- Approximate distance *r* = *a*+*z* to the center of the Earth with the constant mean radius of the Earth *a*
- Replace r by a and ∂/∂r by ∂/∂z, where z is height above mean sea level
- Omit all the metric terms not involving tan φ
- Omit those Coriolis terms that vary as the cosine of the latitude
- Neglect all variations of the gravity g (constant)
- Neglect the vertical component of the diffusion
- All is necessary to guarantee energy and absolute momentum conservation on a shallow Earth

Non-hydrostatic equations of motion (shallow atmosphere)

$$\frac{Du}{Dt} - \frac{uv\tan(\phi)}{a} + \frac{uw}{r} = -\frac{1}{\rho \arccos \phi} \frac{\partial p}{\partial \lambda} + 2\Omega v \sin(\phi) - 2\Omega w \cos(\phi) + v \nabla^{2}(u)$$

$$\frac{Dv}{Dt} + \frac{u^{2}\tan(\phi)}{a} + \frac{vw}{r} = -\frac{1}{\rho a} \frac{\partial p}{\partial \phi} - 2\Omega u \sin(\phi) + v \nabla^{2}(v)$$

$$\frac{Dw}{Dt} - \frac{u^{2} + v^{2}}{r} = -\frac{1}{\rho \partial p} \frac{\partial p}{\partial z} - g + 2\Omega u \cos(\phi) + v \nabla^{2}(w) \quad \text{with variable } g = \frac{d\Phi}{dr} = O \frac{e^{2}}{r^{2}}$$

$$\frac{D\rho}{Dt} + \frac{\rho}{2 \cos \phi} \left[\frac{\partial u}{\partial \lambda} + \frac{\partial (v \cos \phi)}{\partial \phi} \right] + \frac{\rho}{r^{2}} \frac{\partial (e^{-w})}{\partial z} = 0$$

$$c_{v} \frac{DT}{Dt} + p \frac{D}{Dt} (\frac{1}{\rho}) = J$$

$$p = \rho RT \quad \text{Omit terms}$$

$$\frac{D(\cdot)}{Dt} = \frac{\partial (\cdot)}{\partial t} + \frac{u}{2 \cos \phi} \frac{\partial (\cdot)}{\partial \lambda} + \frac{v}{2} \frac{\partial (\cdot)}{\partial \phi} + w \frac{\partial (\cdot)}{\partial z}$$

Non-hydrostatic equations of motion (shallow atmosphere)

 $\frac{Du}{Dt} - \frac{u \operatorname{vtan}(\phi)}{a} = -\frac{1}{\rho \operatorname{a} \cos \phi} \frac{\partial p}{\partial \lambda} + 2\Omega \operatorname{v} \sin(\phi) + v \nabla^2(u)$ $\frac{Dv}{Dt} + \frac{u^2 \tan(\phi)}{a} = -\frac{1}{\rho a} \frac{\partial p}{\partial \phi} - 2\Omega u \sin(\phi) + v \nabla^2(v)$ $\frac{Dw}{Dt} = -\frac{1}{\rho} \frac{\partial p}{\partial z} - g \qquad \text{with constant g}$ $\frac{D\rho}{Dt} + \frac{\rho}{a\cos\phi} \left[\frac{\partial u}{\partial\lambda} + \frac{\partial(v\cos\phi)}{\partial\phi} \right] + \frac{\partial w}{\partial z} = 0$ $c_v \frac{DT}{Dt} + p \frac{D}{Dt} (\frac{1}{\rho}) = J$ $p = \rho RT$ $\frac{D()}{Dt} = \frac{\partial()}{\partial t} + \frac{u}{a\cos\phi}\frac{\partial()}{\partial\lambda} + \frac{v}{a}\frac{\partial()}{\partial\phi} + w\frac{\partial()}{\partial z}$

Hydrostatic equations of motion (shallow atmosphere): **Primitive Equations**

 $\frac{Du}{Dt} - \frac{u \operatorname{vtan}(\phi)}{a} = -\frac{1}{\rho \operatorname{a} \cos \phi} \frac{\partial p}{\partial \lambda} + 2\Omega \operatorname{v} \sin(\phi) + v \nabla^2(u)$ $\frac{Dv}{Dt} + \frac{u^2 \tan(\phi)}{a} = -\frac{1}{\rho a} \frac{\partial p}{\partial \phi} - 2\Omega u \sin(\phi) + v \nabla^2(v)$ $\frac{Dw}{Dt} = -\frac{1}{\rho} \frac{\partial p}{\partial z} - g \qquad \text{with constant g}$ $\frac{D\rho}{Dt} + \frac{\rho}{a\cos\phi} \left[\frac{\partial u}{\partial\lambda} + \frac{\partial(v\cos\phi)}{\partial\phi} \right] + \frac{\partial w}{\partial z} = 0$ $c_v \frac{DT}{Dt} + p \frac{D}{Dt} (\frac{1}{\rho}) = J$ $p = \rho RT$ $\frac{D()}{Dt} = \frac{\partial()}{\partial t} + \frac{u}{a\cos\phi}\frac{\partial()}{\partial\lambda} + \frac{v}{a}\frac{\partial()}{\partial\phi} + w\frac{\partial()}{\partial z}$

Design choices

- The hydrostatic shallow atmosphere equations are called **Primitive Equations (PE)**
- **PE most popular choice** in today's GCMs
- Vertically propagating sound waves are removed
- But note: acoustic modes can also be removed by selecting filtered equation sets
- Filtered equations are sometimes used for special purposes like cloud models, meso-scale models
- Word of caution: filtered equations are not a good choice for global GCMs

Filtered equations: Getting rid of sound waves

- Vertically propagating sound waves are a nuisance in weather and climate models (not important)
- They propagate at high speed and require small time steps in numerical schemes (stability constraints)
- The hydrostatic approximation filters vertically propagating sound waves
- If non-hydrostatic equations need to be used, a filtered equation set might be a choice (just be careful and know the limitations: e.g. shallow flows, static stability requirements)

Filtered equations: Boussinesq

- Boussinesq approximation: set the density to a constant unless it is multiplied with the gravity term
- Continuity equation becomes:

$$\frac{\partial \rho}{\partial t} + \frac{1}{a\cos\phi} \left[\frac{\partial(\rho u)}{\partial\lambda} + \frac{\partial(\rho v\cos\phi)}{\partial\phi} \right] + \frac{\partial(\rho w)}{\partial z} = 0$$
$$\Leftrightarrow \nabla \bullet \vec{v} = 0$$

- Flow is non-divergent
- Limited to shallow flows in the boundary layer

Filtered equations: Anelastic

- Anelastic approximation: the density varies according to a prescribed vertical profile $\overline{\rho}(z)$
- Continuity equation becomes:

$$\frac{\partial \overline{\rho}}{\partial t} + \frac{1}{a\cos\phi} \left[\frac{\partial (\overline{\rho}u)}{\partial \lambda} + \frac{\partial (\overline{\rho}v\cos\phi)}{\partial \phi} \right] + \frac{\partial (\overline{\rho}w)}{\partial z} = 0$$
$$\Rightarrow \nabla \cdot (\overline{\rho}(z) \ \vec{v}) = 0$$

- Justification from scale analysis: density variations in the vertical direction are bigger than horizontal variations
- Problem: Specifying a generic profile is difficult, sometimes $\overline{\rho}(z)$ assumes isentropic conditions

Filtered equations: Pseudo-incompressible

- Neglects the influence of perturbation p' on p'
- The continuity equation includes the steady reference fields $\overline{\theta}(x,y,z)$ and $\overline{\rho}(x,y,z)$ that need to obey the equation of state

$$\overline{p} = p_0 \left(\frac{R_d}{p_0} \overline{\rho} \ \overline{\theta}\right)^{\frac{c_p}{c_0}}$$

• Continuity equation becomes:

$$\nabla \bullet \left(\overline{\rho} \ \overline{\theta} \ \vec{v} \right) = 0$$

- Less severe restriction, wider application range
- Not used in GCMs

Durran (JAS, 1989)

More design decisions: The form of the equations

- Lagrangian versus Eulerian form
- Advective form versus flux form
- Model variables
- Vertical coordinate transformations

Lagrangian versus Eulerian framework

 Lagrangian form: The variations are observed following a moving particle, requires the total derivative, e.g. the continuity equation is:

$$\frac{D\rho}{Dt} + \frac{\rho}{a\cos\phi} \left[\frac{\partial u}{\partial\lambda} + \frac{\partial(v\cos\phi)}{\partial\phi} \right] + \frac{\partial w}{\partial z} = 0$$

 Eulerian form: The variations are observed at a fixed location and snapshot in time, requires partial derivatives, e.g. the continuity equation is:

$$\frac{\partial \rho}{\partial t} + \frac{1}{a \cos \phi} \left[\frac{\partial (\rho u)}{\partial \lambda} + \frac{\partial (\rho v \cos \phi)}{\partial \phi} \right] + \frac{\partial (\rho w)}{\partial z} = 0$$

Advective form versus the flux form

Consider a tracer advection equation for tracer q:

$$\frac{Dq}{Dt} = 0$$
$$\Leftrightarrow \frac{\partial q}{\partial t} + \vec{v} \bullet \nabla q = 0$$

- This is the so-called advective form
- The flux form can be formed by incorporating the continuity equation:

$$\frac{\partial(\rho q)}{\partial t} + \nabla \bullet (\rho q \vec{v}) = 0$$

 The flux form has great advantages concerning mass conservation, especially in finite-volume models

Mass conservation in flux form

• The continuity equation in the Eulerian framework is also an equation in flux form

$$\frac{\partial \rho}{\partial t} + \nabla \bullet \left(\rho \vec{v} \right) = 0$$

• For simplicity, let us assume the equation is 1D:

$$\frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x} (\rho u) = 0$$

• The finite-difference discretization (numerical scheme) for this PDE may be (n time index):

$$\frac{\rho_i^{n+1} - \rho_i^n}{\Delta t} + \frac{1}{\Delta x} \left(\left(\rho u \right)_{i+1/2}^n - \left(\rho u \right)_{i-1/2}^n \right) = 0$$

Mass-conserving by design

Mass conservation in flux form

• Rewrite the equation with numerical fluxes F in the x direction:

$$\rho_{i}^{n+1} = \rho_{i}^{n} - \frac{\Delta t}{\Delta x} \left(F_{i+1/2}^{n} - F_{i-1/2}^{n} \right)$$

 The density variations at the next time step n+1 are determined by the balance of incoming and outgoing fluxes at the grid interfaces with indices i+1/2, i-1/2



Choice of the model variables

- We can choose (within limits) the model variables
- Hydrostatic models lose the ability to forecast the vertical velocity (becomes diagnostic)
- The choices are also determined by the numerical schemes (e.g. vertical coordinate system)
- A common set is u, v, T, p_s, ρ
- Another common set is ζ, δ, Τ, p_s, ρ where ζ, δ are the relative vorticity and horizontal divergence
- The thermodynamic variable is sometimes the potential temperature θ instead of T
- Advantage: built-in conservation $\frac{\partial(\rho\theta)}{\partial t} + \nabla \cdot (\rho\theta \vec{v}) = 0$

Choice of the vertical coordinate

 First decision to make: Orography-intersecting model levels or orography-following coordinate?

Most common choice: orography following, e.g.

- Pressure-based, so-called σ-coordiante: σ = (p-p_t)/(p_s-p_t)
 with p_t (p at the model top), p_s is surface pressure
- Hybrid σ -p coordinate called η -coordinate:

 $\eta = A p_0 + B p_s$ with prescribed coefficients A and B (dependent on vertical position), constant $p_0=1000$ hPa, used in many GCMs

Hybrid (η) vertical coordinates



Floating Lagrangian vertical coordinate

- 2D transport calculations, let layers expand
- Layers are material surfaces, no vertical advection
- Periodic re-mapping of the Lagrangian layers onto reference grid



Choice of the vertical coordinate

Other choices might be:

- Height based coordinates
- Floating Lagrangian coordinate (Lin (2004))
- Isentropic based (hybrid θ -p)
- Shaved cells, step coordinate
- **Requirement:** the new vertical coordinate needs to be **monotonic**
- Whatever we choose it requires a coordinate transformation and modifies the equations of motion, e.g. see the example on the next slide for a pressure-based coordinate s=p

Vertical coordinate transformations



Vertical coordinate transformations

- New vertical coordinates introduce new vertical velocities
- Example: in a pressure-based system the vertical velocity becomes

$$\omega = \frac{dp}{dt} = \dot{p}$$

- In a hybrid σ -p (η) system the vertical velocity becomes $\dot{\eta} = \frac{d\eta}{dt}$
- The new vertical velocity enters the equations of motion, e.g. $\frac{D(\cdot)}{Dt} = \frac{\partial(\cdot)}{\partial t} + \frac{u}{a\cos\phi}\frac{\partial(\cdot)}{\partial\lambda} + \frac{v}{a}\frac{\partial(\cdot)}{\partial\phi} + \frac{\dot{\eta}\frac{\partial(\cdot)}{\partial\eta}}{\partial\eta}$

The pursuit of the 'perfect' model grid

- How to distribute grid points over the sphere: yet to be solved
- Possible design criteria:
 - Highly uniform coverage
 - Orthogonal
 - Structured versus unstructured
 - Adaptive mesh

Platonic solids - Regular grid structures



• Platonic solids can be enclosed in a sphere

Computational grids (horizontal)



Latitude-Longitude Grid



- Popular choice in the past
- Meridians converge: requires polar filters or/and small time steps
- Orthogonal

Adaptive Mesh Refinements (AMR)



AMR on a cubed-sphere mesh

AMR on a latitudelongitude grid

St-Cyr, Jablonowski, et.al (MWR, 2008)

Example of an AMR simulation

Merging vortices



Other non-uniform (nested) grids



Model ICON

Icosahedral grid with nested highresolution regions

under development at the German Weather Service (DWD) and MPI, Hamburg, Germany

Source: DWD

Why do we might want AMR grids in GCMs?

GEOS-5 Modeled Clouds at 7 km Global Resolution for Aug 17, 2009 21z through Aug 26, 2009 21z

NASA/GSFC:

Dr. Bill Putman Dr. Max Suarez Greg Shirah The pursuit of the 'perfect' positions of variables in the discrete system

- Having decided on the basic distribution of grid points, a choice has to be made as to how to arrange the different prognostic variables on the grid
- Most obvious choice of representing all variables at the same point has disadvantages
- There are many choices, called:
 A, B, C, D, E, Z or ZM grid (the first five are based on a classification by Akio Arakawa (UCLA))

Example: Grid staggerings

- Many choices how to place scalars and vector winds in the computational grid
- Examples are



- Staggerings determine properties of the numerical schemes: dispersion and diffusion properties
- Additional staggering options in the vertical



Vertical grid staggerings: Charney-Phillips



The pursuit of the 'perfect' numerical scheme

- We want: high order of accuracy, but computationally cheap method
- We need: discretizations in space and time
- Many space discretization philosophies:
 - Finite difference methods (FD)
 - Finite volume methods (FV)
 - Finite (spectral) element methods (FE, SE)
 - Spectral methods
- Sometimes different spatial methods are used in the horizontal and vertical directions

The pursuit of the 'perfect' numerical scheme

- Phase errors and damping should be small (often a compromise)
- Explicit scheme is 'easy' to program, but it will only be conditionally stable and so the choice of time step is limited
- Implicit schemes are absolutely stable; however at every time step a system of simultaneous equations has to be solved
- More than two time levels: additional computational modes and possibly separation of the solution at odd and even time steps. Higher storage (memory) requirements.

Discretizations in time: explicit versus implicit

 The earlier example used a so-called explicit time stepping scheme with two time levels n+1 and n

$$\rho_{i}^{n+1} = \rho_{i}^{n} - \frac{\Delta t}{\Delta x} \left(F_{i+1/2}^{n} - F_{i-1/2}^{n} \right)$$

• How about rewriting this equation as an implicit scheme $o^{n+1} = o^n \frac{\Delta t}{(E^{n+1} - E^{n+1})}$

$$\rho_i^{n+1} = \rho_i^n - \frac{\Delta t}{\Delta x} \left(F_{i+1/2}^{n+1} - F_{i-1/2}^{n+1} \right)$$

- It uses the (unknown) fluxes at the future time step, requires sophisticated numerical methods to solve (more expensive, e.g. iterative methods)
- Big advantage: increased numerical stability, allows longer time steps

Multi-level time discretizations

- We can increase the order of accuracy of a time stepping scheme by using multiple time levels, e.g. three time levels n+1, n, n-1
- Picture an equation like (describes Rayleigh friction)

$$\frac{\partial \mathbf{u}}{\partial \mathbf{t}} = -\tau \mathbf{u}$$

• Possible time discretizations are:

$$\frac{u^{n+1} - u^n}{\Delta t} = -\tau u^n \quad \text{or} \quad \frac{u^{n+1} - u^n}{\Delta t} = -\tau u^{n+1} \quad (\text{explicit / implicit})$$
$$\frac{u^{n+1} - u^{n-1}}{2\Delta t} = -\tau u^n \quad (3 \text{ - step method, called Leapfrog method})$$

Multi-level time discretizations

- Leapfrog is a popular choice in today's GCMs
- Uses 3 time levels and computes the forcing (here friction) at the center time with index n, forcing is applied over a time span of 2Δt

$$u^{n+1} = u^{n-1} - 2\Delta t \tau u^n$$

- 2nd-order accurate
- Unfortunately, the Leapfrog method can be numerically unstable (has a computational mode, separates odd and even time steps)
- But it then can be stabilized by applying a time filter that mimics diffusion in time (Asselin filter)

Time-splitting

- Sometimes is equations are separated into two parts (e.g. horizontal and vertical part) and solved independently (in a time-split fashion)
- Example: Consider the tracer conservation equation

$$\frac{\partial q}{\partial t} + \vec{v} \bullet \nabla q = 0$$

- Can be split like $q^* = q^{n-1} + 2\Delta t \left(\vec{v}_h \cdot \nabla_h q \right)^n$ (using Leapfrog) $q^{n+1} = q^* + 2\Delta t \left(w \frac{\partial q}{\partial z} \right)^n$
- Allows to use different techniques for the horizontal (h) and vertical advection, introduces a splitting error

The flavor of FV spatial discretizations

- Finite volume discretization are based on an integrated version of the equations of motion
- Consider 2D example with h (=height of shallow water system), this is a conservation law:

$$\frac{\partial h}{\partial t} + \nabla \bullet \left(h \vec{v} \right) = 0$$

- Conservation equation can be integrated over spatial domain Ω with "volume" (here an area) A_Ω and time t

$$\int_{t_n}^{t_{n+1}} \int_{\Omega} \left(\frac{\partial}{\partial t} h \right) d\Omega \, dt + \int_{\Omega} \int_{t_n}^{t_{n+1}} \nabla \cdot (h \, \vec{v}) \, dt \, d\Omega = 0$$

The flavor of FV spatial discretizations

Integration over spatial domain Ω with area/volume
 A_Ω and time t, can be rearranged:

$$\int_{t_n}^{t_{n+1}} \left(\frac{\partial}{\partial t}\bar{h}\right) dt + \frac{\Delta t}{A_\Omega} \int_{\Omega} \nabla \cdot \vec{F} \, d\Omega = 0$$

- Overbar denotes the spatial mean, F denotes time averaged fluxed across the interface of a volume
- Apply Gauss' divergence theorem to second term:

$$\int_{t_n}^{t_{n+1}} \left(\frac{\partial}{\partial t}\bar{h}\right) dt + \frac{\Delta t}{A_\Omega} \oint_{\partial\Omega} \vec{F} \cdot d\vec{n} = 0$$

Introduces line integral, \vec{n} is the line segment vector normal to the boundary

The flavor of FV spatial discretizations

Leads to discretized forecast equation:

$$ar{h}^{n+1} = ar{h}^n - rac{\Delta t}{A_\Omega} \sum_{i=1}^4 l_i \, ec{F_i} \cdot ec{n_i}$$

- Where I_i denotes the length of a line segment, \vec{n}_i is the unit vector normal to the boundary 'i'
- The future time step is determined by the sum of all fluxes across the boundaries of a finite volume
- Order of accuracy determined by the fluxes *F*, rely on subgrid-scale representations (constant, linear, quadratic, cubic) of transported variable (here h)
- Idea: express subgrid structure of h with polynomials

The flavor of spectral transform models

- Spectral transform methods on latitude-longitude grids have been very popular in the past
- Some GCM (ECHAM, ECMWF's weather model IFS) still use it
- Idea:
 - Use a model formulation with vorticity and divergence as prognostic variables
 - Use Fourier and Legendre transformation to transform/ represent the flow in spectral space
 - Solve the linear parts of the Eqs. in spectral space (exact)
 - Solve the nonlinear parts in grid point space
- Highly accurate, but suffers from Gibb's ringing, nonlocal discretization

The flavor of spectral transform models

Triangular wave number range m, n



Spectral representations of variable q:

$$q(\lambda,\varphi,t) = \sum_{m=-M}^{M} \sum_{n=|m|}^{N(m)} q_n^m(t) Y_n^m(\lambda,\varphi)$$

Spherical harmonics

$$Y_n^m(\lambda,\varphi) = P_n^m(\sin\varphi)\exp(im\lambda)$$
associated Legendre Fourier mod

functions

des

Structure of the global basis functions:



The flavor of spectral transform models



- Triangular truncation T... with N(m)=M is unique, provides uniform spatial resolution over the entire surface of the sphere, e.g. M=42, 85, 170
- Eliminates pole problem in latitude-longitude grids
- Allows reduced grids with fewer points towards poles

Summary

Today we reviewed:

- Scientific and numerical aspects in the design process
- Different forms of the equations and variables
- Computational meshes, staggering options
- Characteristics (and accuracy) of numerical discretizations
- Suitable time-stepping schemes and vertical coordinates