Unstructured-mesh based non-oscillatoryforward-in-time integration for atmospheric flows

J. Szmelter

Loughborough University, UK

Acknowledgements: Mike Gillard, Francesco Cocetta, Piotr Smolarkiewicz



Funded by the European Union



Generalised Nonhydrostatic Model

Smolarkiewicz et. al. JCP 2016

$$\frac{\partial \mathcal{G}\varrho}{\partial t} + \nabla \cdot (\mathcal{G}\varrho \boldsymbol{v}) = 0$$

$$\frac{\partial \mathcal{G} \varrho \theta'}{\partial t} + \nabla \cdot \left(\mathcal{G} \varrho \boldsymbol{v} \theta' \right) = -\mathcal{G} \varrho \left(\tilde{G}^T \boldsymbol{u} \cdot \nabla \theta_a - \mathcal{H} \right)$$

$$\frac{\partial \mathcal{G} \varrho \boldsymbol{u}}{\partial t} + \nabla \cdot (\mathcal{G} \varrho \boldsymbol{v} \otimes \boldsymbol{u}) = -\mathcal{G} \varrho \left(\Theta \tilde{G} \nabla \varphi + g \Upsilon_B \frac{\theta'}{\theta_b} + \boldsymbol{f} \times (\boldsymbol{u} - \Upsilon_C \boldsymbol{u}_a) - \mathcal{M}'(\boldsymbol{u}, \boldsymbol{u}, \Upsilon_C) - \mathcal{D} \right)$$
$$\varrho := \left[\rho(\boldsymbol{x}, t), \frac{\rho_b(z)\theta_b(z)}{\theta(\boldsymbol{x}, t)}, \rho_b(z) \right] \quad \varphi := \left[c_p \theta_0 \pi', c_p \theta_0 \pi, c_p \theta_b \pi \right] \quad \Theta := \left[\frac{\theta}{\theta_0}, \frac{\theta}{\theta_0}, 1 \right] \quad \Upsilon_B := \left[\frac{\theta_b(z)}{\theta_a(\boldsymbol{x})}, \frac{\theta_b(z)}{\theta_a(\boldsymbol{x})}, 1 \right] \quad \Upsilon_C := \left[\frac{\theta}{\theta_a(\boldsymbol{x})}, \frac{\theta}{\theta_a(\boldsymbol{x})}, 1 \right]$$

[compressible, pseudo-incompressible, anelastic]

$$\varphi = c_p \theta_0 \left[\left(\frac{R_d}{p_0} \rho \theta \right)^{R_d/c_v} - \pi_a \right]$$
. Ideal

Ideal gas law for compressible option

NFT based Nonhydrostatic Models

Non-oscillatory Forward-in-Time (NFT)

$$\frac{\partial G\Phi}{\partial t} + \nabla \cdot (\boldsymbol{V}\Phi) = G\mathcal{R}$$

$$\Phi_i^{n+1} = \mathcal{A}_i \left(\Phi^n + \frac{1}{2} \delta t \mathcal{R}, \mathbf{V}^{n+\frac{1}{2}}, G^n, G^{n+1} \right) + \frac{1}{2} \delta t \mathcal{R}^{n+1}$$

- Semi-implicit
- Multidimensional Positive Definite Advection Transport Algorithm (MPDATA)
- Krylov solver

For details see: Kühnlein et. al. GMD 2019 for global model (compressible) FVM (Finite Volume Module) of IFS Szmelter et. al. JCP 2019 for local model (anelastic)

Global flows - FVM: discrete operators



Deconinck et. al. Parallel Proc & Appl Math 2016

Generalised Conjugate Residual (k) scheme

For any initial guess, ϕ^0 , set $r^0 = \mathcal{L}(\phi^0) - R$, $p^0 = \mathcal{P}^{-1}(r^0)$; then iterate:

For $n = 1, 2, \dots$ until convergence

Smolarkiewicz & Margolin 2000

$$\sum_{I=1}^{M} \frac{\partial}{\partial x^{I}} \left(\sum_{J=1}^{M} C^{IJ} \frac{\partial \phi}{\partial x^{J}} + D^{I} \phi \right) - A\phi = R$$

 $\mathcal{L}(\phi) - R = 0 \; .$

$$\frac{\partial^k \mathcal{P}(\phi)}{\partial \tau^k} + \frac{1}{T_{k-1}(\tau)} \frac{\partial^{k-1} \mathcal{P}(\phi)}{\partial \tau^{k-1}} + \dots + \frac{1}{T_1(\tau)} \frac{\partial \mathcal{P}(\phi)}{\partial \tau} = \mathcal{L}(\phi) - R$$

for
$$\nu = 0, ..., k - 1$$

$$\beta = -\frac{\langle r^{\nu} \mathcal{L} (p^{\nu}) \rangle}{\langle \mathcal{L} (p^{\nu}) \mathcal{L} (p^{\nu}) \rangle},$$

$$\phi^{\nu+1} = \phi^{\nu} + \beta p^{\nu},$$

$$r^{\nu+1} = r^{\nu} + \beta \mathcal{L} (p^{\nu}),$$
exit if $||r^{\nu+1}|| \le \epsilon,$

$$e = \mathcal{P}^{-1} (r^{\nu+1}),$$

evaluate $\mathcal{L}(e)$

for
$$l = 0, ...\nu$$

$$\alpha_l = \frac{\langle \mathcal{L}(e)\mathcal{L}(p^l) \rangle}{\langle \mathcal{L}(p^l) \mathcal{L}(p^l) \rangle},$$

$$p^{\nu+1} = e + \sum_l^{\nu} \alpha_l p^l,$$

$$\mathcal{L}(p^{\nu+1}) = \mathcal{L}(e) + \sum_l^{\nu} \alpha_l \mathcal{L}(e^l),$$

reset $[\phi, r, e, \mathcal{L}(e)]^k$ to $[\phi, r, e, \mathcal{L}(e)]^0$

Richardson iteration

Take an operator $\mathcal{P} \approx \mathcal{L}$

$$\sum_{I=1}^{M} \frac{\partial}{\partial x^{I}} \left(\sum_{J=1}^{M} C^{IJ} \frac{\partial \phi}{\partial x^{J}} + D^{I} \phi \right) - A\phi = R$$

by neglecting cross derivative terms i.e. $I \neq J$

Solve the resulting residual equation:

$$\frac{\partial e}{\partial \tilde{\tau}} = \mathcal{P}(e) - r \quad \Rightarrow \quad e^{\mu+1} = e^{\mu} + \Delta \tilde{\tau} \left[\mathcal{P}(e^{\mu}) - r^{\nu+1} \right]$$

Due to the thin shell atmosphere: treat the unstructured horizontal direction explicitly and the structured vertical direction implicitly, i.e.

let
$$\mathcal{P}(e) \equiv \mathcal{P}_H(e^{\mu}) + \mathcal{P}_z(e^{\mu+1})$$

$$(I - \Delta \tilde{\tau} \mathcal{P}_z) e^{\mu + 1} = e^{\mu} + \Delta \tilde{\tau} \left(\mathcal{P}_H(e^{\mu}) - r^{\nu + 1} \right) = \mathcal{R}^{\mu}$$

$$e^{\mu+1} = \left[I - \Delta ilde{ au} \mathcal{P}_z
ight]^{-1} \mathcal{R}^{\mu}$$
 Use Thomas Algorithm

Smolarkiewicz & Margolin 2000

Unstructured mesh Jacobi iteration

Solve the residual equation by lagging the off-diagonal $\ensuremath{\mathcal{P}}$ components:

 $e^{\mu+1} = e^{\mu} - \frac{\mathcal{P}(e^{\mu}) - r^{\nu+1}}{\mathcal{D}}$

Apply the thin shell atmosphere treatment again:

let $\mathcal{P}(e) \equiv \mathcal{P}_{H}(e^{\mu}) + \mathcal{P}_{z}(e^{\mu+1})$ For the Helmholtz $0 = -\sum_{\ell=1}^{3} \left(\frac{A_{\ell}^{\star}}{\zeta_{\ell}} \nabla \cdot \zeta_{\ell} \widetilde{\mathbf{G}}^{T}(\check{\mathbf{u}} - \mathbf{C} \nabla \varphi') \right) - B^{\star}(\varphi' - \widehat{\varphi'})$ $\mathcal{D}_{k,i} = -\frac{1}{4V_{i}} \sum_{l} \frac{A_{l\,k,i}^{\star}}{\zeta_{l\,k,i}} \sum_{j=1}^{nbrs} \frac{\zeta_{l\,k,j}}{V_{j}} \left(\mathcal{S}_{xj}^{2} (\widetilde{\mathbf{G}}^{T} \mathbf{C})_{xx_{k,j}} + \mathcal{S}_{yj}^{2} (\widetilde{\mathbf{G}}^{T} \mathbf{C})_{yy_{k,j}} \right)$ $\left(I + \frac{\mathcal{P}_{z}}{\mathcal{D}} \right) \hat{e}^{\mu+1} = e^{\mu} - \frac{\mathcal{P}_{H}(e^{\mu}) - r^{\nu+1}}{\mathcal{D}} \equiv \frac{\mathcal{R}^{\mu}}{\mathcal{D}}$ $\hat{e}^{\mu+1} = \left[\mathcal{D} + \mathcal{P}_{z} \right]^{-1} \mathcal{R}^{\mu}$ Use Thomas Algorithm

Jacobi requires some solution relaxation:

$$e^{\mu+1} = (1-\omega) e^{\mu} + \omega \hat{e}^{\mu+1} , \quad \omega \approx \frac{2}{3}$$

 $\mathcal{P}(e) = r$

 $\mathcal{P}(e) + \mathcal{D}(e) - \mathcal{D}(e) = r$

Dry baroclinic instability, O360 mesh, Jacobi (left), Richardson (centre), Difference (right)



Table 1: 10 days, dry baroclinic instability test with 60 vertical levels, computed on the Cray XC30 at ECMWF (Dynamical core only).

Grid	Richardson runtime	Richardson convergence	Jacobi runtime	Jacobi convergence	Tasks x Threads	Total cores
O180 O360	2897 s 3700 s	14	1540 s 1374 s	1	36 x 1 108 x 6	36 648
O300 O720	11296 s	33	3194 s	$\frac{1}{2}$	$360 \ge 6$	2160
$\begin{array}{c} O1280\\ O1800 \end{array}$	35884 s 46670 s	$\frac{47}{56}$	7976 s 8675 s	$\frac{2}{2}$	720 x 6 2400 x 6	$\begin{array}{c} 4320\\ 14400 \end{array}$

Jacobi provides ca 2x acceleration on coarse meshes expanding to ca 5-6 x acceleration on finer meshes

Preconditioners use 3 iterations

Unstructured mesh multigrid preconditioner in the horizontal

Parallel restriction, prolongation and Atlas mesh



Baroclinic Instability Testcase

O360 plots for day 10 of dry baroclinic instability test: Multigrid (top), Difference vs. Richardson (bottom).



Grid	Richardson	Multigrid	Speedup	Task x thread
O360	814 s	403 s	2.02x	108 x 6 (648 cores)
O720	3001 s	1017 s	2.95x	360 x 6 (2160 cores)
O1280	13586 s	3038 s	4.41x	720 x 6 (4320 cores)
O1800	19544 s	4060 s	4.81x	2400 x 6 (14400 cores)

10 days, dry baroclinic instability test ~27km Horizontal resolution, with 31 vertically stretched levels at a depth of 45km computed on the Cray XC30 at ECMWF (Dynamical core only)

3 iterations

1 Jacobi as a smoother and 3 Jacobi on the coarse grid

Local models: The effect of critical levels on stratified flows past an axisymmetric mountain



a = 5000 m, $h(r) = h_0 \left(1 + \frac{r^2}{a^2}\right)^{-\frac{3}{2}}$, $r \equiv \sqrt{x^2 + y^2}$

Szmelter et. al. JCP 2015

Prismatic mesh

$$\tilde{z}_{i,k} = \tilde{z}_{i,k-1} + \delta \tilde{z}_k$$
$$\tilde{z}_{i,k} = \tilde{z}_{i,k} \left(1 - \frac{h_i}{H}\right) + h_i$$



Triangular surface mesh



Fig. 12. Isentropes at T=6 in y = 0 vertical plane for experiments LS2 (left, top) LS3 (right, top), LS4 (left, bottom) and LS5 (right, bottom).



Fig. 13. As in Fig. 12 but at T=18.



the isentrope with undisturbed height $z = 0.94z_c$ LS5.



Stratified flows past a sphere

Cocetta et al. PoF 2021

Re = 200 Re = 300



Q-method (second invariant of the deformation tensor)

Neutrally-stratified flow $Fr = \infty$



Non-axisymmetric attached vortex regime



Lee-waves instability regime

 $0.8 \gtrsim Fr \gtrsim 0.6$ Re = 200



Two-dimensional vortex shedding regime



Study of drag coefficient



$$Re = 200$$

Re = 300

$$\Delta C_d = C_d (Re, \frac{1}{Fr}) - C_d (Re, 0) \qquad C_d = \frac{F_d}{0.5 \rho_0 V_0^2 A}$$

Cocetta et al. PoF 2021

Stratified flow past two spheres



Top view

Re=300 neutrally stratified

Fr =1.625 steady-state non-axisymmetric attached vortex regime





Fr=0.625 steady-state lee wave instability



Fr=0.25 two dimensional vortex shedding regime





top view

side view

Stratified flow past two spheres



top view

side view

Remarks

- The presented NFT methodology can complement established NWP methods.
- Algorithmic parallel scalability is important for future computing architectures. Technical challenges of time to solution & energy efficiency have been partially addressed by a development of specialised preconditioners for the non-symmetric Krylov-subspace solver.
 For a global atmospheric model, the Jacobi preconditioning on a single mesh provides a substantial speed up.
- General characterisation of flows past spheres confirmed that stratification dominates the flow pattern for $Fr \ge 0$ and unification of flow patterns. Flow patterns are sensitive to the spheres' relative location. Regimes characterising the flow past a single sphere are also observed for multi-sphere configurations.
- A record of consistent and accurate simulations for NFT MPDATA class of solvers continues to grow. Our recent developments targeted engineering applications and include a development of a LMN solver for strong thermally driven flows and implementation of the dynamic Smagorinski model.