

Energy Estimates and Weak Boundary Procedures for LAM

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Conclusion II

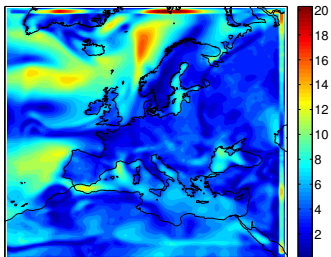
Introduction/Motivation for this work

- ▶ Boundary conditions are set using *ad hoc* intuition (Davies relaxation)
- ▶ Different effects in e.g. Arctic and Europe
- ▶ "The apparent success of spectral nudging for one-way nesting is at least partly an artifact of very bad procedures for windowing and blending"¹

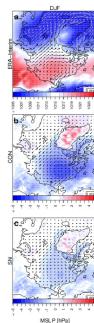
¹John P. Boyd "Limited-Area Fourier Spectral Models and Data Analysis Schemes: Windows, Fourier Extension, Davies Relaxation and All That", Mon. Weat. Rev. Vol. 133 2005

Motivating example

- ▶ Europe has "standing waves"
- ▶ Similar problems over Arctic (mitigated with Spectral Nudging)



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- ▶ Analyze boundary procedures!

Model problem

Simplification for purpose of illustration, without losing generality!

- ▶ $\vec{u}_t + \sum_{i=1}^3 \nabla \vec{F}_i(\vec{u}) = 0$ N-S/Euler/primitive equations

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- ▶ $u_t + u_x = 0$

Why do we impose Boundary Conditions?

(Energy point of view)

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- ▶ $\frac{d}{dt} \|u\|^2 = \underbrace{(u(0, t))^2}_{\text{Faster than exponential growth}} - \underbrace{(u(1, t))^2}_{\text{Decay}}$

- ▶ We must set conditions to bound the energy for $u(0, t)$ with data: $g(t) < C$ for some constant $C \in \mathcal{R}$

Discrete Boundary procedures

CKD : $U_t + P^{-1}QU = 0$

$$U(\cdot, t) = (I - W)U(\cdot, t) + WG(t) \quad W = \text{diag}(\tanh)$$

- ▶ G is data!
- ▶ $U = (U_0, U_1, \dots, U_N)^T$ is the solution.
- ▶ CKD = Classic Källberg-Davies Relaxation

Discrete Boundary procedures

$$\text{WKD} : U_t + P^{-1}QU = P^{-1}W(G - U) \quad W = \text{diag}(\tanh)$$

- ▶ WKD = Weak Källberg-Davies Relaxation, proven stable

Discrete Boundary procedures

$$\text{SAT} : U_t + P^{-1}QU = P^{-1}E_0(G - U) \quad E_0 = \text{diag}(1, 0, \dots, 0)$$

- ▶ SAT = Simultaneous Approximation Term (Carpenter et. al.), proven stable

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- ▶ CKD = Classic Källberg-Davies Relaxation
- ▶ WKD = Weak Källberg-Davies Relaxation, proven stable
- ▶ SAT = Simultaneous Approximation Term (Carpenter et. al.), proven stable

Energy estimate for Davies-relaxation

Strong Davies relaxation

$$U_t + P^{-1}QU = 0$$

$$U = U + W_{\tanh}(G - U)$$

- ▶ No energy estimate! (For stability(?) proof use GKS theory *DIFFICULT!*)

- ▶ In fact there are counterexamples showing instability for strong methods.²

Weak Davies relaxation

$$U_t + P^{-1}QU = \tau P^{-1}W_{\tanh}(G - U)$$

- ▶ Discrete Energy Method to prove stability
- ▶ Multiply with $U^T P$
- ▶ Use summation-by-parts property of difference operator

²"High Order Difference Methods for Time Dependent PDE", Bertil

Energy estimate for Davies-relaxation cont.

$$\begin{aligned} \frac{d}{dt} \|U\|_P^2 &= U_1^2(1 - 2w_1) + 2w_1 U_1 G_1 - U_N^2(1 + 2w_N) + 2w_N U_N G_N \\ &\quad - \sum_{i=2}^{N-1} (2w_i U_i^2 - 2w_i U_i G_i) \quad (1) \end{aligned}$$

Since $w_1 \geq \frac{1}{2}$ and $w_i \geq 0 \therefore$ Davies Relaxation is *proven* energy-stable!

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- ▶ Solution quality is not affected if $\|U_i - G_i\|$ is "small".

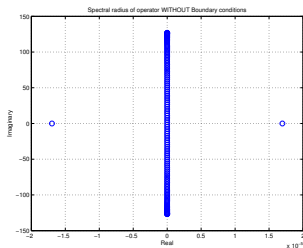
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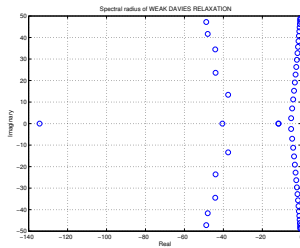
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- ▶ Stable, *BUT* imposing conditions where not needed!
- ▶ Solution quality is not affected if $\|U_i - G_i\|$ is "small".
- ▶ Usually we impose time-interpolated G i.e. $G_i(t) = \pi_{6h} G_i(t_n)$

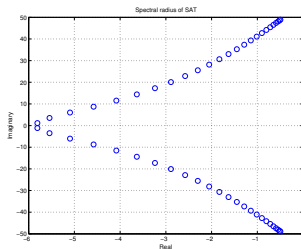
Spectral Radius of operators part I



(a) Operator without boundary conditions



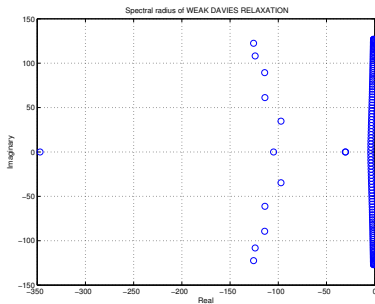
(b) WKD



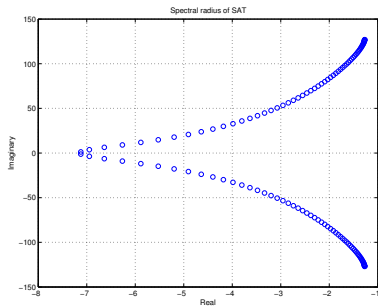
(c) SAT

Spectral Radius of operators part II

Increasing resolution by a factor of 5:



(d) WKD



(e) SAT

$$\rho(WKD) \approx 350, \quad \rho(SAT) \approx 140$$

Computational results

$$u_t + u_x = 0$$

$$u(0, t) = G(0, t), \quad x \in [0, 1]$$

Exact solution is $u(x, t) = \sin(2\pi(x - t - \frac{1}{2})) = G(x, t)$

We use the exact solution as initial data, i.e. assume perfect assimilation

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- ▶ Use exact $G(\cdot, t)$ as boundary data (show movie)

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- ▶ πG is linear interpolation in time (show movie)

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- ▶ πG is constant interpolation in time (closest in time) (show movie)

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- ▶ πG is P_3 -Hermite (no new minima or maxima are introduced)

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- ▶ Mismatching data on outflow boundary

Conclusion

- ▶ Davies relaxation in its weak form is a penalty method
- ▶ WKD is proven stable and yields similar results with CKD
- ▶ CKD and WKD impose data on all boundaries, even when not needed.
- ▶ SAT is proven stable, imposes data only where needed!
- ▶ The WKD operator is much more stiff than SAT by a factor of $\approx 20!$
- ▶ When data is not close to the solution on outflow boundaries (and WKD) yields unsatisfactory results \Rightarrow *horizontal diffusion mitigates this problem*
- ▶ The excessive diffusion can be the reason for the poor results over the Arctic
- ▶ Non-matching data causes a "standing wave" on the outflow boundary with WKD and CKD
- ▶ SAT yields similar results for exact and almost matching data (time interpolation error is visible), but outperforms CKD and WKD for non-matching data.

Conclusion II

- ▶ Theory for penalty-based boundary conditions is considered mature and ready to be used for NWP and climate
- ▶ Results are already extended to non-linear multidimensional systems, but for purpose of illustration a model problem was shown here.

Conclusion II

▶ THANK YOU FOR LISTENING!