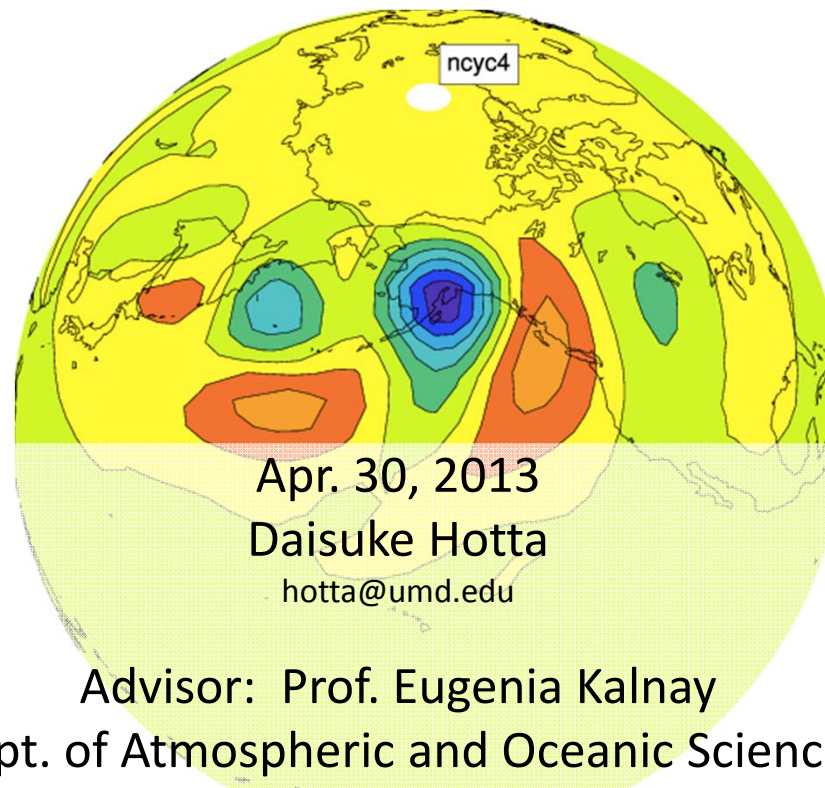


# Reduction of Temporal Discretization Error in an Atmospheric General Circulation Model (AGCM)

## Final Presentation



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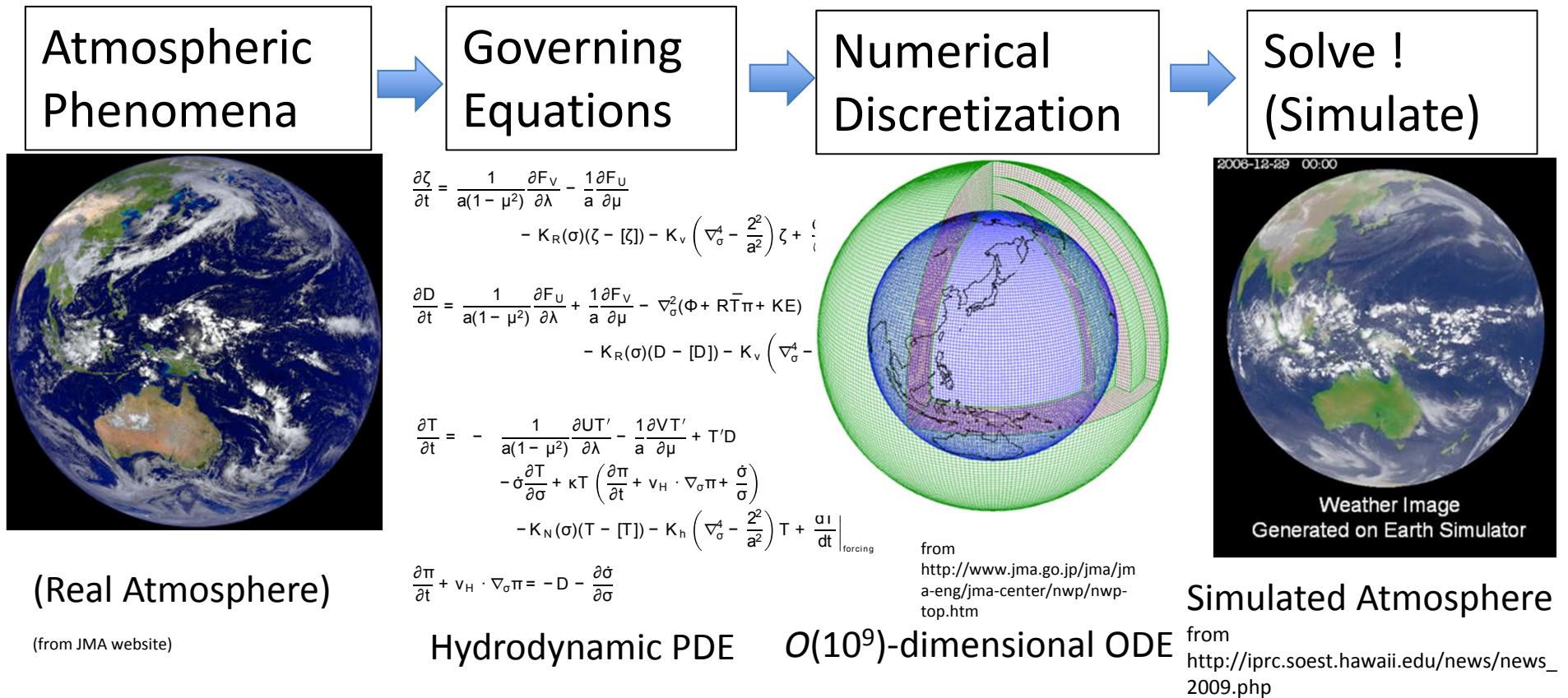
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  4. Stiffness-problem and semi-implicit method
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# Numerical Weather Prediction (NWP): = Initial Value Problem of PDE



**AGCM:** Atmospheric General Circulation Model  
= a computer program which simulates the flow of global atmosphere by numerically integrating the governing fluid dynamical PDEs

# Introduction: Motivation

Due to computational restrictions ...

- most AGCMs adopt low-order time-integration schemes, such as
  - Leap-frog with Robert-Asselin filter (1<sup>st</sup> order)
  - Explicit Backward Euler (aka. Matsuno; 1<sup>st</sup> order)
- Often,  $\Delta t$  is taken as the largest value for which computational instability is suppressed,
- under the premise that temporal discretization errors are negligible compared to those associated with spatial discretization or Physical Parameterizations.

# Introduction: Motivation

However ...

- Spatial resolutions become finer and finer as the supercomputers become faster.
- Is the premise that time truncation errors are negligible justified ?
- If not, how can we alleviate such errors ?

→ Goal of the Project: Reduction of such model errors

Approaches :

Phase 1: Use a more accurate scheme with the same computational cost

Phase 2: Identify and parameterize the error, and reduce it using data assimilation

Phase I:  
Implementation of Lorenz  $N$ -cycle  
to SPEEDY model

# Phase 1 : A Better integration scheme (Lorenz $N$ -cycle)

Lorenz (1971) proposed an incredibly smart time-integration scheme which:

- requires only 1 function evaluation per step
- but yet (**every  $N$  steps**) it is of
  - (up to) 4<sup>th</sup>-order accuracy (for nonlinear systems)
  - arbitrary order of accuracy (for linear systems)

However, this scheme seems to have remained forgotten.  
No applications have been made to AGCMs.

→ Apply Lorenz  $N$ -cycle to an AGCM (Phase 1)

# Phase 1: Approach

- Implement Lorenz  $N$ -cycle to an existing AGCM
- Implement 4<sup>th</sup> order Runge-Kutta (RK4) as well as a reference
- Compare the accuracy and efficiency of the newly introduced schemes with the original scheme
- Perform verification by the Jablonowski-Williamson (2006) Dynamical-core Tests



# Phase 1: Algorithms

ODE to be solved:

$$\frac{du}{dt} = F(u)$$

Lorenz $N$ -cycle	<b>(existing)</b> Leap-frog with Robert-Asselin Filter	4 <sup>th</sup> order Runge-Kutta
$G = 0, w^1 = 1, w^k = \frac{N}{N-k},$ do $k = 1, \dots, N - 1$ $G \rightarrow w^k F(u) + (1 - w^k)G$ $u \rightarrow u + G\Delta t$ end do	do $k = 1, \dots,$ $u^{+1} \rightarrow u^{-1} + 2\Delta t F(u^0)$ $u^0 \rightarrow u^0 + \alpha(u^{+1} - 2u^0 + u^{-1})$ $u^{-1} \rightarrow u^0$ $u^0 \rightarrow u^{+1}$ end do	do $k = 1, \dots,$ $h^1 \rightarrow F(u)$ $h^2 \rightarrow F(u + \frac{\Delta t}{2}h^1)$ $h^3 \rightarrow F(u + \frac{\Delta t}{2}h^2)$ $h^4 \rightarrow F(u + \Delta t h^3)$ $u_{+} = \frac{\Delta t}{6}(h^1 + 2h^2 + 2h^3 + h^4)$ end do
Memory consumption: <b>2 x</b> dim{model state}	Memory consumption: <b>2 x</b> dim{model state}	Memory consumption: <b>4 x</b> dim{model state}
$F$ -evaluation: <b>1</b> per time step	$F$ -evaluation: <b>1</b> per time step	$F$ -evaluation: <b>4</b> per time step
accuracy: ( $N \leq 4$ ) <b><math>O((N\Delta t)^N)</math></b> (every $N$ steps) <b><math>O(N\Delta t)</math></b> (in between)	accuracy: <b><math>O(\Delta t)</math></b>	accuracy: <b><math>O(\Delta t^4)</math></b>

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# AGCM: SPEEDY model

- A fast AGCM with simplified physical parameterizations
- Developed in ICTP (Italy) by Drs. F. Molteni and F. Kucharski
- Horizontal Discretization:
  - Spectral Representation with Spherical Harmonics
  - truncated at total wavenumber 30 (T30)  $\approx$  400km mesh
- Vertical Discretization:
  - 8-layers Finite Difference on  $\sigma$ -coordinate
- Temporal Discretization:
  - Leap-Frog scheme with Robert-Asselin Filter
  - (1<sup>st</sup> order Forward Euler for the physical parameterizations)

# The equations solved: the “primitive” equation system (PDEs) on a spherical geometry + parametrized processes

## Dynamical Core

## Sub-grid Parametrizations

$$\frac{\partial \zeta}{\partial t} = \frac{1}{a(1-\mu^2)} \frac{\partial F_V}{\partial \lambda} - \frac{1}{a} \frac{\partial F_U}{\partial \mu} - K_R(\sigma)(\zeta - [\zeta]) - K_V \left( \nabla_\sigma^4 - \frac{2^2}{a^2} \right) \zeta + \left. \frac{d\zeta}{dt} \right|_{\text{forcing}}$$

$$\frac{\partial D}{\partial t} = \frac{1}{a(1-\mu^2)} \frac{\partial F_U}{\partial \lambda} + \frac{1}{a} \frac{\partial F_V}{\partial \mu} - \nabla_\sigma^2(\Phi + R\bar{T}\pi + KE) - K_R(\sigma)(D - [D]) - K_V \left( \nabla_\sigma^4 - \frac{2^2}{a^2} \right) D$$

$$\frac{\partial T}{\partial t} = - \frac{1}{a(1-\mu^2)} \frac{\partial U T'}{\partial \lambda} - \frac{1}{a} \frac{\partial V T'}{\partial \mu} + T'D - \dot{\sigma} \frac{\partial T}{\partial \sigma} + \kappa T \left( \frac{\partial \pi}{\partial t} + v_H \cdot \nabla_\sigma \pi + \frac{\dot{\sigma}}{\sigma} \right) - K_N(\sigma)(T - [T]) - K_h \left( \nabla_\sigma^4 - \frac{2^2}{a^2} \right) T + \left. \frac{dT}{dt} \right|_{\text{forcing}}$$

$$\frac{\partial \pi}{\partial t} + v_H \cdot \nabla_\sigma \pi = -D - \frac{\partial \dot{\sigma}}{\partial \sigma}$$

$$\theta \equiv T (p/p_{00})^{-\kappa}$$

$$\kappa \equiv R/C_p$$

$$\phi \equiv gz = - \int_1^\sigma \frac{RT}{\sigma} d\sigma$$

$$\pi \equiv \ln p_s$$

$$\dot{\sigma} \equiv \frac{d\sigma}{dt}$$

$$\mu \equiv \sin \phi$$

$$U \equiv u \cos \phi$$

$$V \equiv v \cos \phi$$

$$\zeta \equiv \frac{1}{a(1-\mu^2)} \frac{\partial V}{\partial \lambda} - \frac{1}{a} \frac{\partial U}{\partial \mu}$$

$$D \equiv \frac{1}{a(1-\mu^2)} \frac{\partial U}{\partial \lambda} + \frac{1}{a} \frac{\partial V}{\partial \mu}$$

$$F_U \equiv (\zeta + f)V - \dot{\sigma} \frac{\partial U}{\partial \sigma} - \frac{RT'}{a} \frac{\partial \pi}{\partial \lambda}$$

$$F_V \equiv -(\zeta + f)U - \dot{\sigma} \frac{\partial V}{\partial \sigma} - \frac{RT'}{a} (1-\mu^2) \frac{\partial \pi}{\partial \mu}$$

$$KE \equiv \frac{U^2 + V^2}{2(1-\mu^2)}$$

$$v_H \cdot \nabla \equiv \frac{u}{a \cos \phi} \left( \frac{\partial}{\partial \lambda} \right)_\sigma + \frac{v}{a} \left( \frac{\partial}{\partial \phi} \right)_\sigma = \frac{U}{a(1-\mu^2)} \left( \frac{\partial}{\partial \lambda} \right)_\sigma + \frac{V}{a} \left( \frac{\partial}{\partial \mu} \right)_\sigma$$

$$\nabla_\sigma^2 \equiv \frac{1}{a^2(1-\mu^2)} \frac{\partial^2}{\partial \lambda^2} + \frac{1}{a^2} \frac{\partial}{\partial \mu} \left[ (1-\mu^2) \frac{\partial}{\partial \mu} \right]$$

$$T \equiv \bar{T}(\sigma) + T'$$

$$\dot{\sigma} = -\sigma \frac{\partial \pi}{\partial t} - \int_0^\sigma D d\sigma - \int_0^\sigma v_H \cdot \nabla_\sigma \pi d\sigma,$$

$${}^u F_{N-h}^m = {}^d F_{N-h}^m = F^*$$

$${}^u F_{k-h}^{LR} = {}^u F_{k+h}^{LR} \tau_k^{LR} + (1-\tau_k^{LR}) {}^d B_k$$

$${}^u F_{N-h}^Q = F^* \cdot Q_{N-h}^{ext}; \quad {}^d F_{N-h}^Q = F^* \cdot Q_{N-h}$$

$${}^u B_k = f_b(T_k) \sigma_{SB} (T_k^i + w_k^{LR} (T_{k-h}^i - T_k^i))$$

$${}^u F_{N-h}^{SE} = F^* \cdot SE_N; \quad {}^d F_{N-h}^{SE} = F^* \cdot SE_{N-h}$$

$$U_{sa} = f_{wind} U_N; \quad V_{sa} = f_{veg} V_N$$

$$\left( \frac{\partial Q_N}{\partial t} \right)_{cnv} = - \frac{g F^* (Q_{N-h}^{ext} - Q_{N-h})}{\Delta p_N} = - \frac{Q_N - Q_{thr}}{\tau_{cnv}}$$

$$|V_0| = (U_{sa}^2 + V_{sa}^2 + V_{gust}^2)^{1/2}$$

$$F^* = \frac{\Delta p_N}{g \tau_{cnv}} \frac{Q_N - Q_{thr}}{Q_N^{ext} - Q_{N-h}}$$

$$E_k^m = \epsilon(\sigma_k) F_{k+h}^m$$

$$F_{k-h}^m = F_{k+h}^m + E_k^m$$

$${}^u F_{k-h}^Q = {}^u F_{k+h}^Q + E_k^m Q_k; \quad {}^d F_{k-h}^Q = {}^d F_{k+h}^Q$$

$${}^u F_{k-h}^{SE} = {}^u F_{k+h}^{SE} + E_k^m SE_k; \quad {}^d F_{k-h}^{SE} = {}^d F_{k+h}^{SE}$$

$$P_{cnv} = {}^u F_{k0+h}^Q - F_{k0+h}^m Q_{k0+h}^{ext}$$

$$\Delta F_{k0}^Q = {}^u F_{k0+h}^Q - {}^d F_{k0+h}^Q - P_{cnv}$$

$$\Delta F_{k0}^{SE} = {}^u F_{k0+h}^{SE} - {}^d F_{k0+h}^{SE} + L_c P_{cnv}$$

$$\left( \frac{\partial Q_k}{\partial t} \right)_{lac} = - \frac{Q_k - RH(\sigma_k)_{lac} Q_k^{ext}}{\tau_{lac}}$$

$$\left( \frac{\partial T_k}{\partial t} \right)_{lac} = - \frac{L_c}{C_p} \left( \frac{\partial Q_k}{\partial t} \right)_{lac}$$

$$P_{lac} = - \frac{1}{g} \sum_{k=2}^N \Delta p_k \left( \frac{\partial Q_k}{\partial t} \right)_{lac}$$

$$\tau_{k,1}^{LR} = \exp [ - ( \alpha_{win}^{LR} + \alpha_{cl}^{LR} CLC ) \Delta p'_k ]$$

$$\tau_{k,2}^{LR} = \exp ( - \alpha_{CO_2}^{LR} \Delta p'_k )$$

$$\tau_{k,3}^{LR} = \exp ( - \alpha_{wv1}^{LR} Q_k \Delta p'_k )$$

$$\tau_{k,4}^{LR} = \exp ( - \alpha_{wv2}^{LR} Q_k \Delta p'_k )$$

$${}^d F_{k+h}^{LR} = {}^d F_{k-h}^{LR} \tau_k^{LR} + (1-\tau_k^{LR}) {}^d B_k$$

$$\alpha_{sw} = \frac{D_{top} W_{top} + f_{veg} D_{veg}}{D_{top} W_{top} + f_{veg} D_{veg}}$$

$${}^u F_{ls}^U = \tau_{ls}^U = \dots$$

$${}^u F_{ls}^V = \tau_{ls}^V = \dots$$

$${}^u F_{ls}^{SE} = SHF_{ls} = \rho_{sa} C_p |V_0|$$

$${}^u F_{ls}^Q = E_{ls} = \rho_{sa} C_l^H |V_0| \max [ \dots ]$$

$${}^u F_{ss}^{SE} = SHF_{ss} = \rho_{sa} C_p |V_0|$$

$${}^u F_{ss}^Q = E_{ss} = \rho_{sa} C_s^H |V_0| \max [ \dots ]$$

$$F_s^{SR} + F_s^{LR} + SHF_{ls} + E_{ls}$$

$$d_s c_s \frac{\partial T_s}{\partial t} = GHF_{ls} - d_s c_s \tau_s^{-1} T_s$$

$$\Delta T(t+1) = \frac{\tau}{\tau + \delta t} \Delta T(t)$$

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# Stiffness Problem

- The “Primitive” Equations = Stiff system :
  - Fast but insignificant modes (= gravity waves) are superposed on the slow but meteorologically meaningful modes (= Rossby waves)
  - Due to the CFL condition for the fast modes, we need very small timestepping  $\Delta t$
- Semi-implicit method (Robert, 1969)

# Semi-Implicit for Leapfrog (1) Formulation

ODE to be solved:  $\frac{du}{dt} = \underbrace{F^E(u)}_{\text{Slow modes}} + \underbrace{L^I u}_{\substack{\text{Fast modes} \\ \text{assumed Linear}}}$

Solve this discretized equation

$$\frac{u^{n+1} - u^{n-1}}{2\Delta t} = \underbrace{F^E(u^n)}_{\text{Explicit}} + \underbrace{L^I (\alpha u^{n+1} + (1 - \alpha)u^{n-1})}_{\text{Implicit}}$$

# Semi-Implicit for Leapfrog (2)

## How to solve

Define: 
$$\delta u = \frac{u^{n+1} - u^{n-1}}{2\Delta t}$$

Substitute  $u^{n+1} = u^{n-1} + 2\Delta t\delta u$   
to the discretized equation,

$$\begin{aligned}\delta u &= F^E(u^n) + L^I u^{n-1} + 2\alpha\Delta t L^I \delta u \\ \Leftrightarrow \delta u &= (I - 2\alpha\Delta t L^I)^{-1} (F^E(u^n) + L^I u^{n-1})\end{aligned}$$

Once you get  $\delta u$ , you can integrate  
the equation by

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



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# Semi-implicit Lorenz $N$ -cycle

## Explicit (original)

```
do  $k = 0, \dots$   
   $w \leftarrow w^{\text{mod}(k, N)}$   
   $G \leftarrow wF(u) + (1 - w)G$   
   $u \leftarrow u + G\Delta t$   
end do
```

## Semi-implicit

```
do  $k = 0, \dots$   
   $w \leftarrow w^{\text{mod}(k, N)}$   
   $G \leftarrow wF^E(u) + (1 - w)G$   
   $\delta u = (I - \alpha\Delta tL^I)^{-1}(G + L^I u)$   
   $u \leftarrow u + \Delta t\delta u$   
end do
```

# Accuracy(Consistency) & Stability Analysis: Method

- Following Durran (1991, 1999; *MWR*) and Williamson (2011; *MWR*), apply semi-implicit modification to the second term of the equation:

$$\frac{du}{dt} = i\omega_L u + i\omega_H u$$

- Examine the modulus of Amplification factor  $A$ .
- If  $|A| < 1$ , the scheme is stable.
- Range of interest:

$$\omega_L \Delta t < 0.5, \quad \omega_H \Delta t > 2$$

i.e., CFL condition is met for low-frequency part but is violated for high-frequency part.

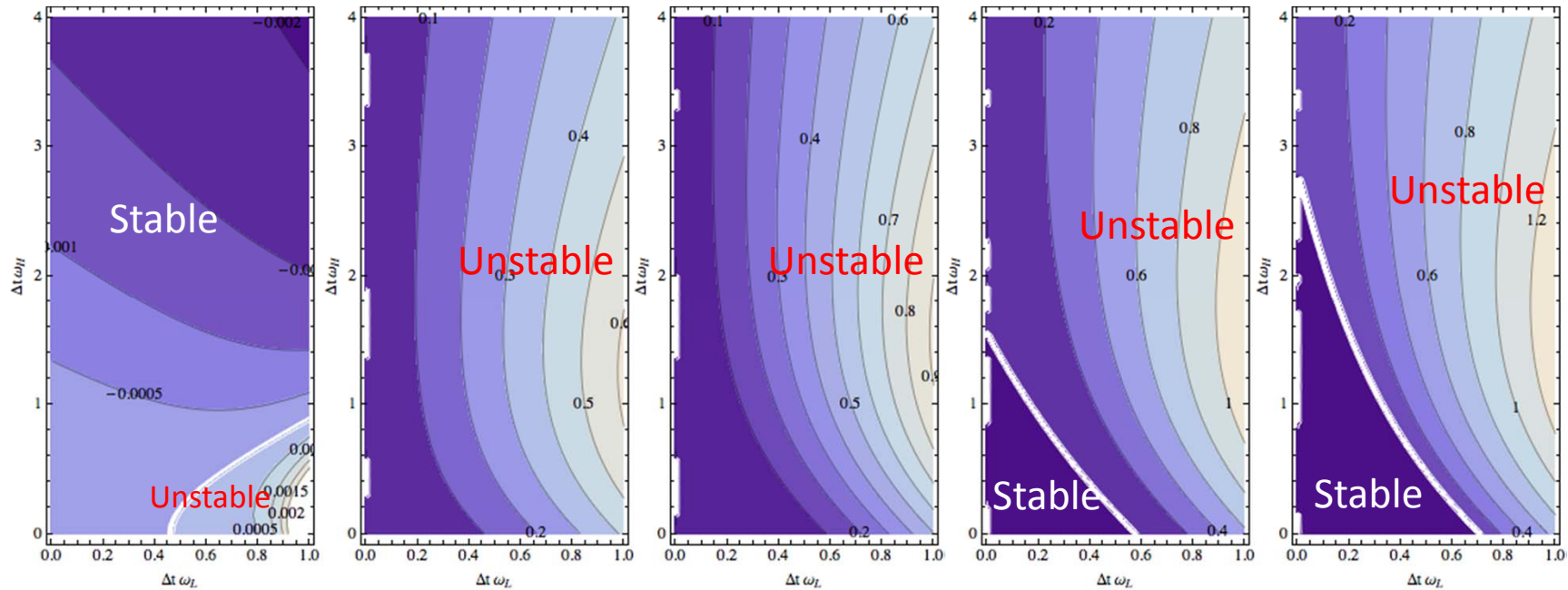
# Accuracy(Consistency)

- Truncation Error:

$$\frac{u^N - u^{\text{Exact}}}{u^0} = \frac{1}{2N}(1 - 2\alpha)\omega_H(\omega_H + \omega_L)(N\Delta t)^2 + O(\Delta t^3)$$

- By taking  $\alpha=1/2$  (Crank-Nicolson), the scheme becomes 2<sup>nd</sup>-order

# Plots of $|A| - 1$ ( $\alpha=1/2$ : Crank-Nicolson)



Leapfrog with  
R/A filter:  
Stable almost  
everywhere

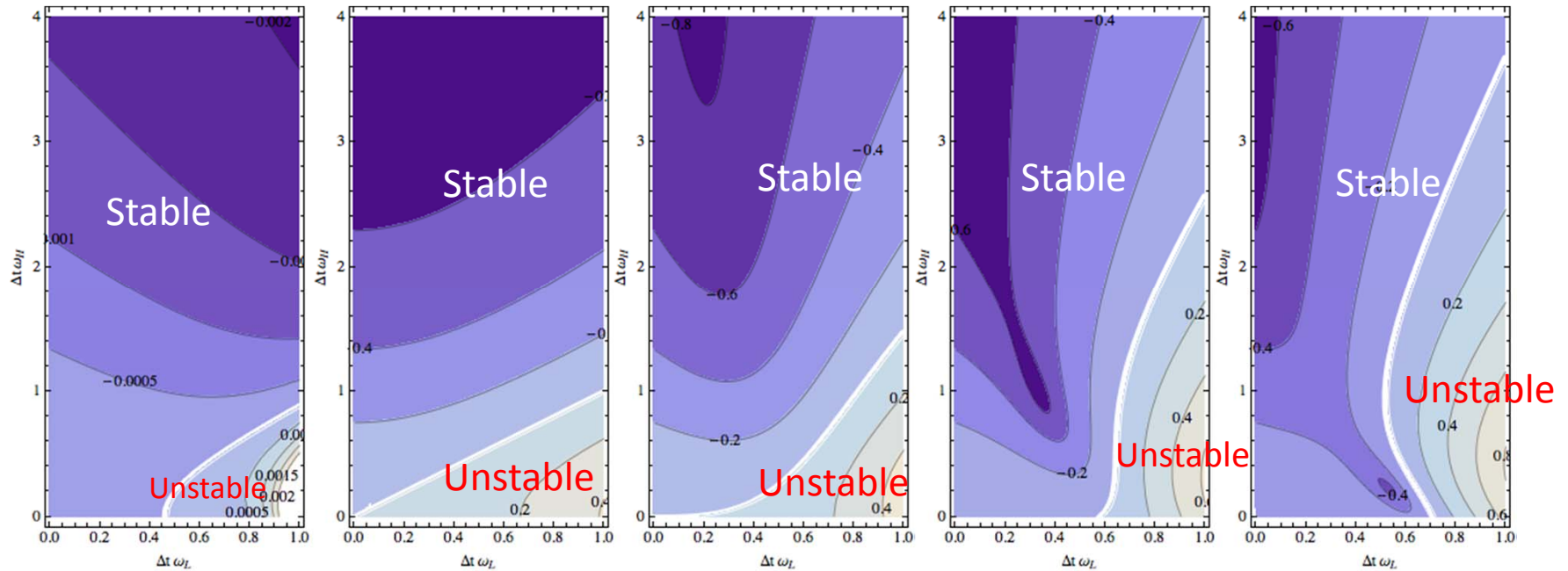
Lorenz 1-cycle  
(=Forward Euler):  
Unstable  
everywhere

Lorenz 2-cycle:  
Unstable  
everywhere

Lorenz 3-cycle:  
Absolutely  
Unstable for  
> 1.5

Lorenz 4-cycle:  
Absolutely  
Unstable for  
> 2.7

# Plots of $|A| - 1$ ( $\alpha=1$ : Backward Euler)



Ref:  
Leapfrog+R/A filter

Lorenz 1-cycle  
(=Forward Euler):

Lorenz 2-cycle:

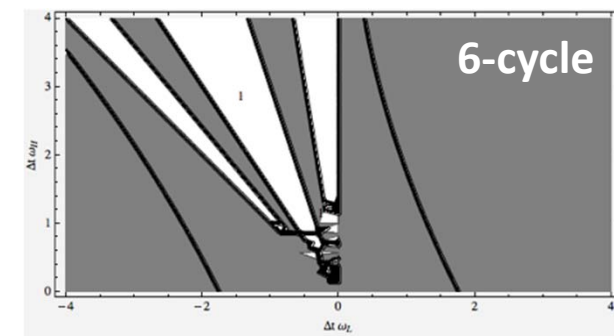
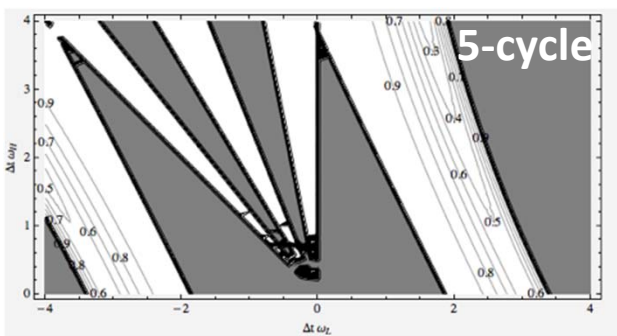
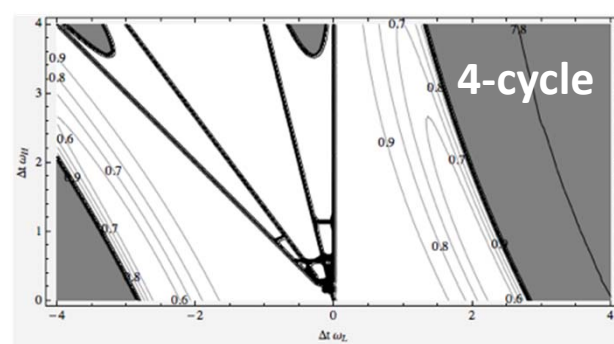
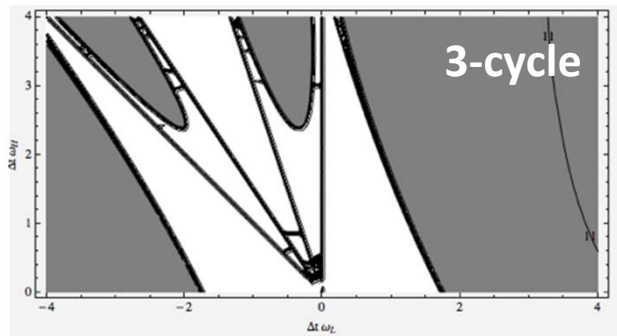
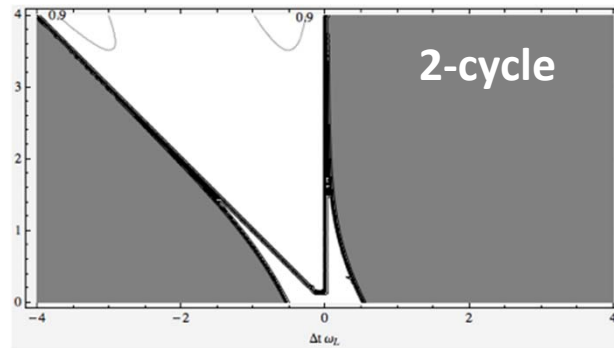
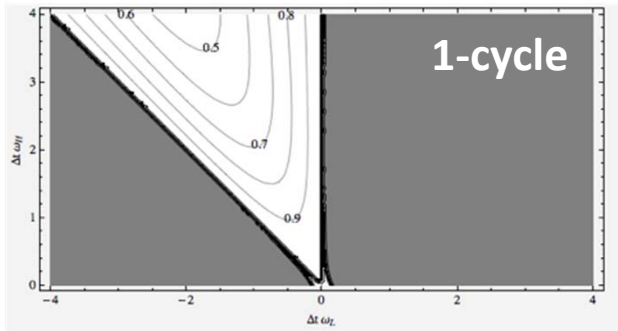
Lorenz 3-cycle:

Lorenz 4-cycle:

Stability is good, but damping is too strong (~50%)

# Stability for $\alpha=1/2$ : Crank-Nicolson

## $N=1, \dots, 6$

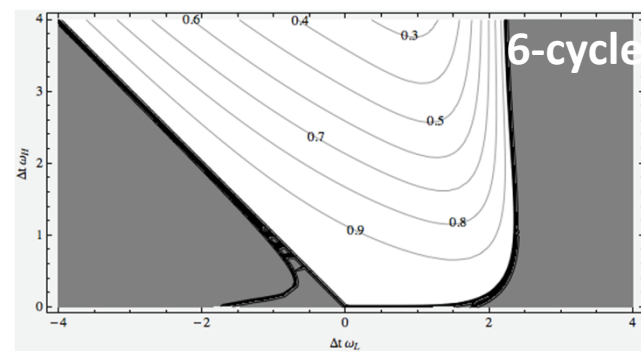
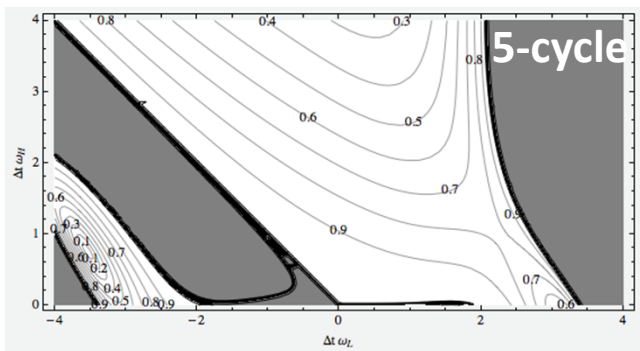
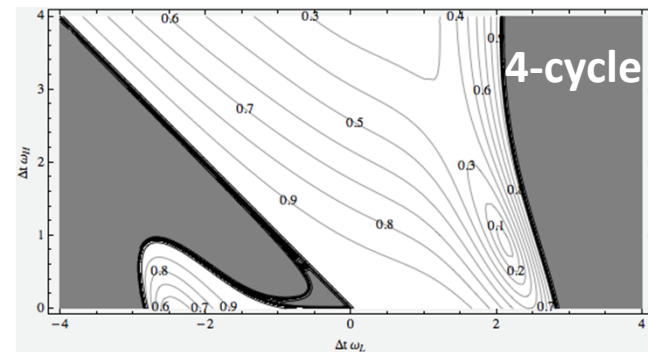
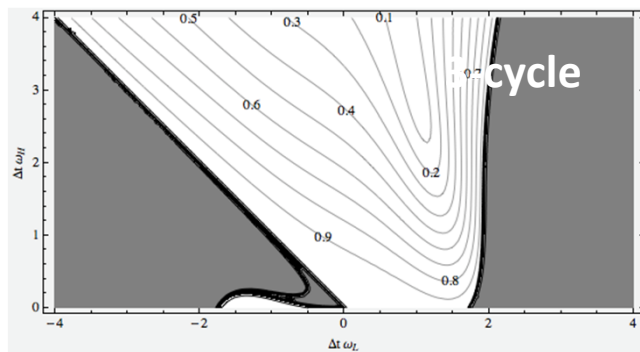
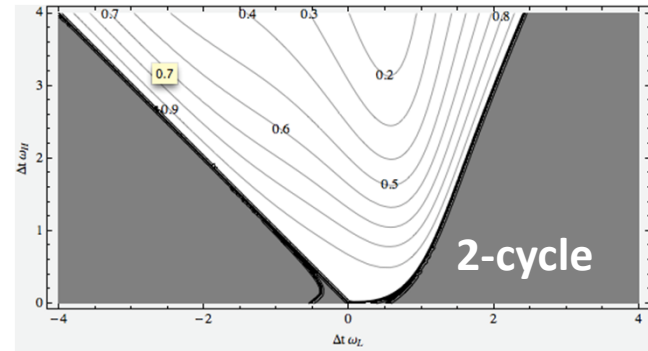
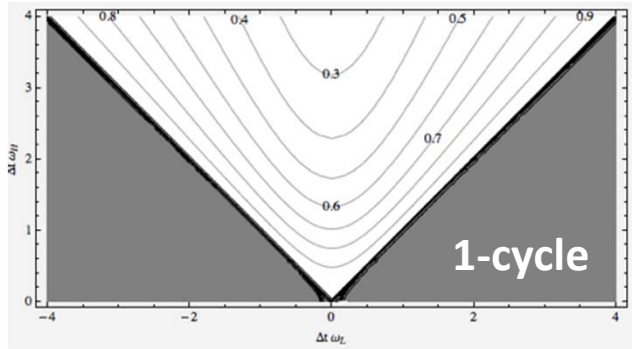


White: stable  $|A| < 1$

Gray: Unstable  $|A| > 1$

# Stability for $\alpha=1$ : Backward Euler

## $N=1, \dots, 6$



White: stable  $|A| < 1$

Gray: Unstable  $|A| > 1$



# Stability: Summary

- Crank-Nicolson semi-implicit  $N$ -cycle is unstable for  $N=1,2$
- Stability region is largest for  $N=4$
- More unstable than Leapfrog.
  
- Backward Euler semi-implicit  $N$ -cycle is more stable, but damping is too strong

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# Code Validation: Idea

- Lorenz  $N$ -cycle:
  - Lorenz 1-cycle is equivalent to Forward Euler, which is built-in in the SPEEDY model.
  - $\rightarrow$  Compare Lorenz 1-cycle with Forward Euler.
- RK4 :
  - For a linear system, RK4 with  $4\Delta t$  is equivalent to Lorenz 4-cycle with  $\Delta t$ .
  - $\rightarrow$  Remove all nonlinear terms from SPEEDY and Compare RK4 ( $4\Delta t$ ) with Lorenz 4-cycle ( $\Delta t$ ).

# Code Validation: Results

- Outputs of single step integrations of Lorenz 1-cycle and Forward Euler from the same initial condition are compared using UNIX diff command.
  - Result → no difference (Success)!
- Similarly, outputs of single step integration of RK4 and four-step integration of Lorenz 4-cycle from the same initial condition are compared using UNIX diff command.
  - Result → no difference (Success)!

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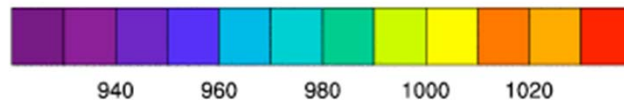
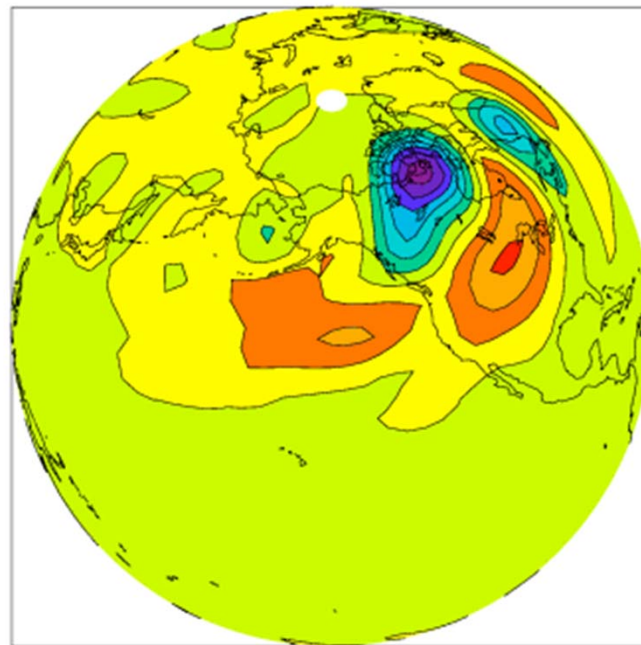
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# Verification: Dynamical-Core test cases

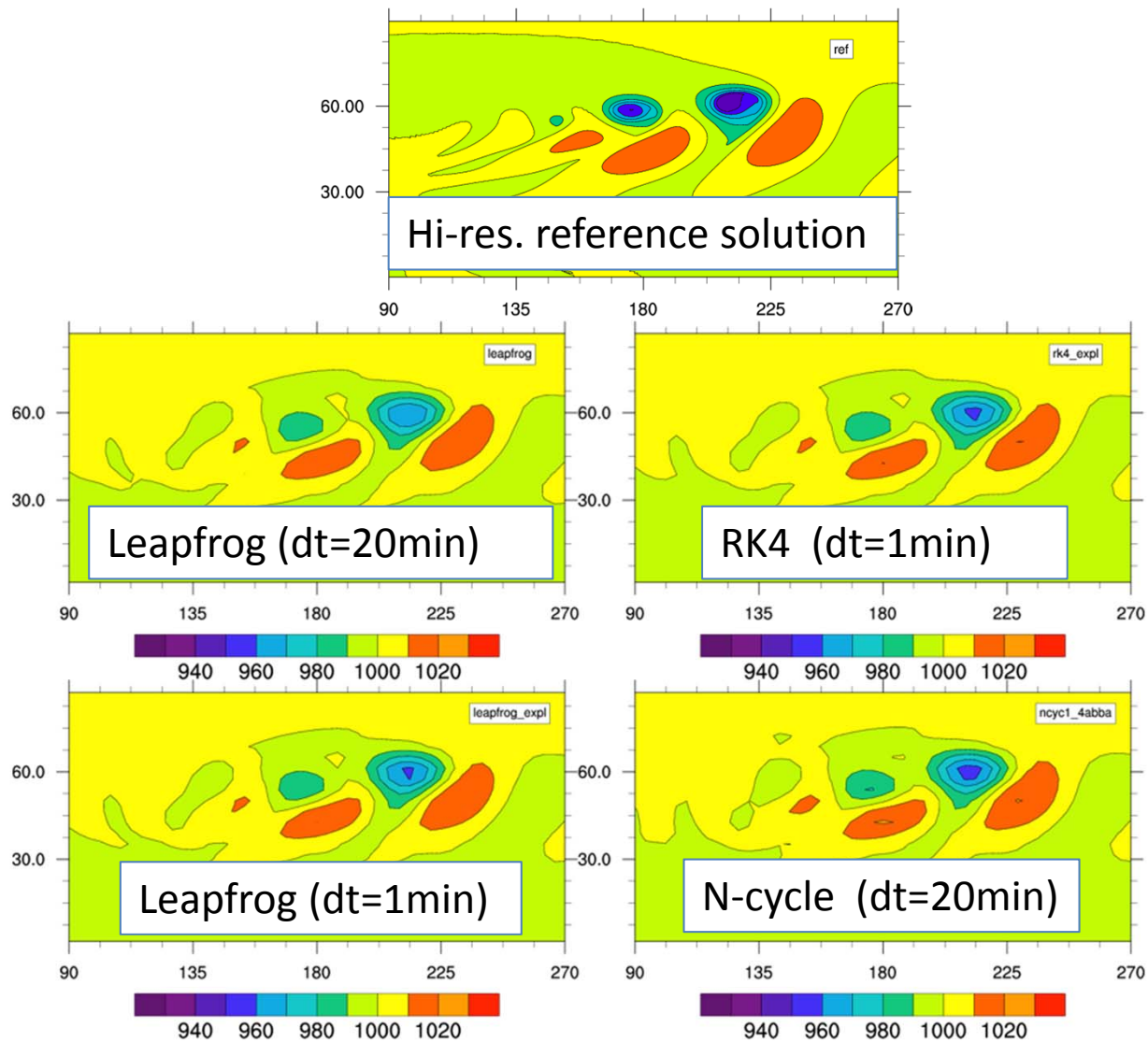
- Standardized tests for dynamical cores of AGCM proposed by Jablonowski and Williamson (2006) which consists of two tests:
  1. Steady-State test:
    - A model is integrated from an analytical steady-state solution of the primitive equation
    - The model is evaluated by how well it can keep the steady-state intact.
  2. Baroclinic-Wave test:
    - The model is integrated with initial and boundary conditions which is designed to produce an idealized baroclinic wave (= extratropic cyclones and highs)

# Baroclinic-wave Test: Result (Lorenz 4-cycle)

Baroclinic wave test: ncyc1\_4abba Day 11



# Snapshots of surface-pressure at Day 9

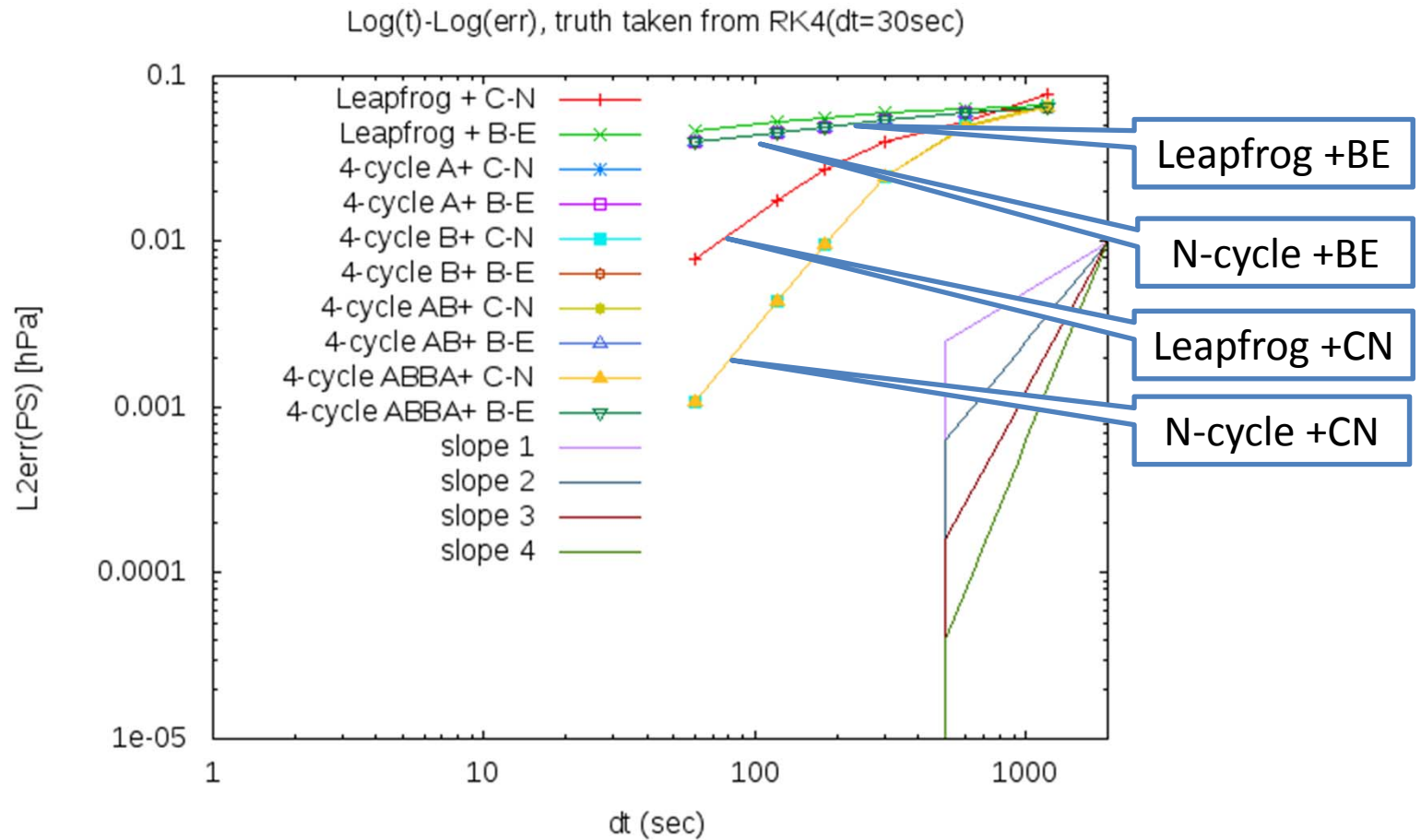




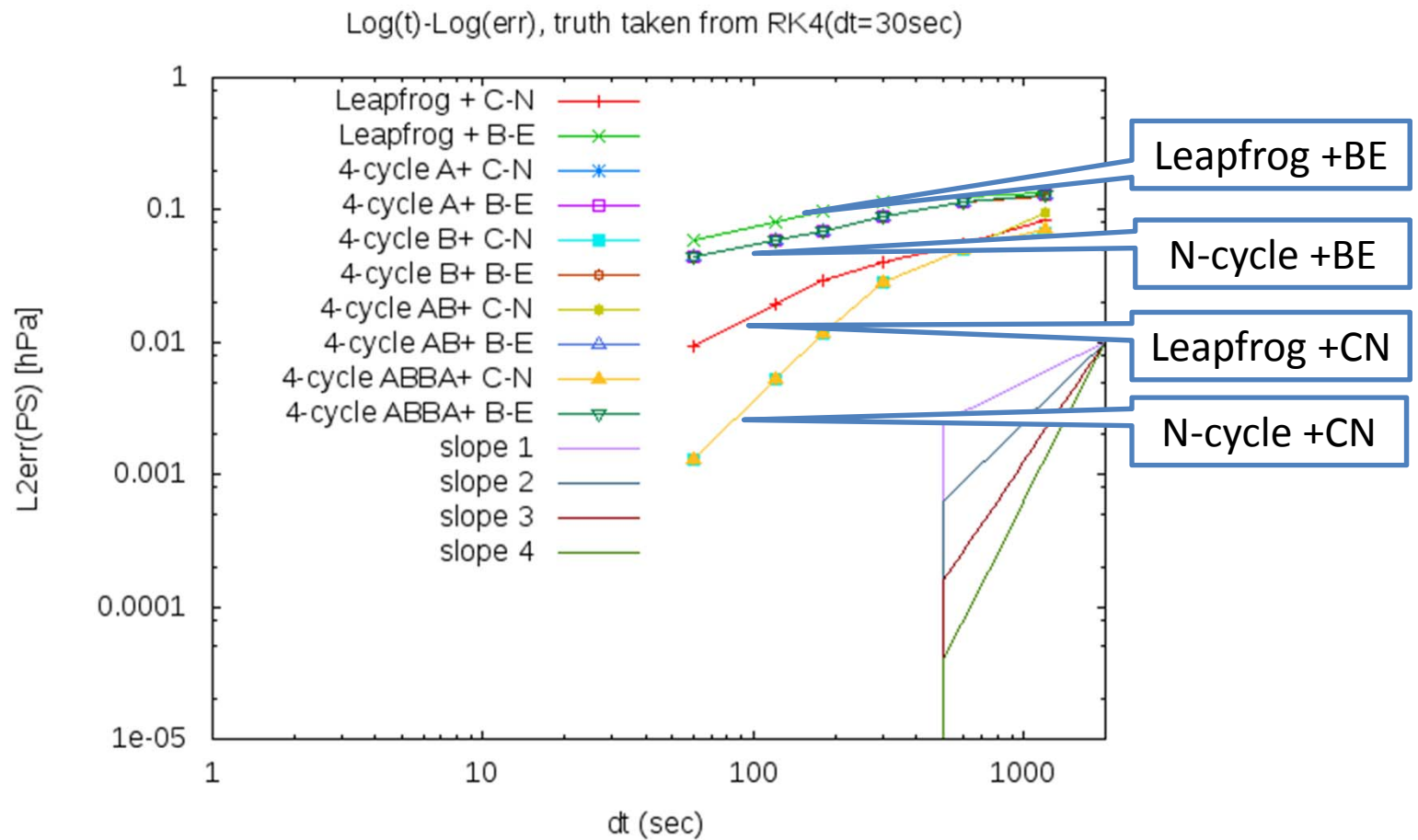
# Order estimation

- Lorenz 4-cycle is supposedly of 4<sup>th</sup>-order, while Leapfrog (with R/A filter) is only of 1<sup>st</sup>-order
- Confirm this by plotting  $L^2$  error vs.  $\Delta t$  on a log-log plane.
- Experimental set-up: Jablonowski-Williamson baroclinic-wave test
- Measure of the error: difference in surface pressure with respect to the reference solution produced by RK4 with  $\Delta t=0.5\text{min}$  in  $L^2$ -norm

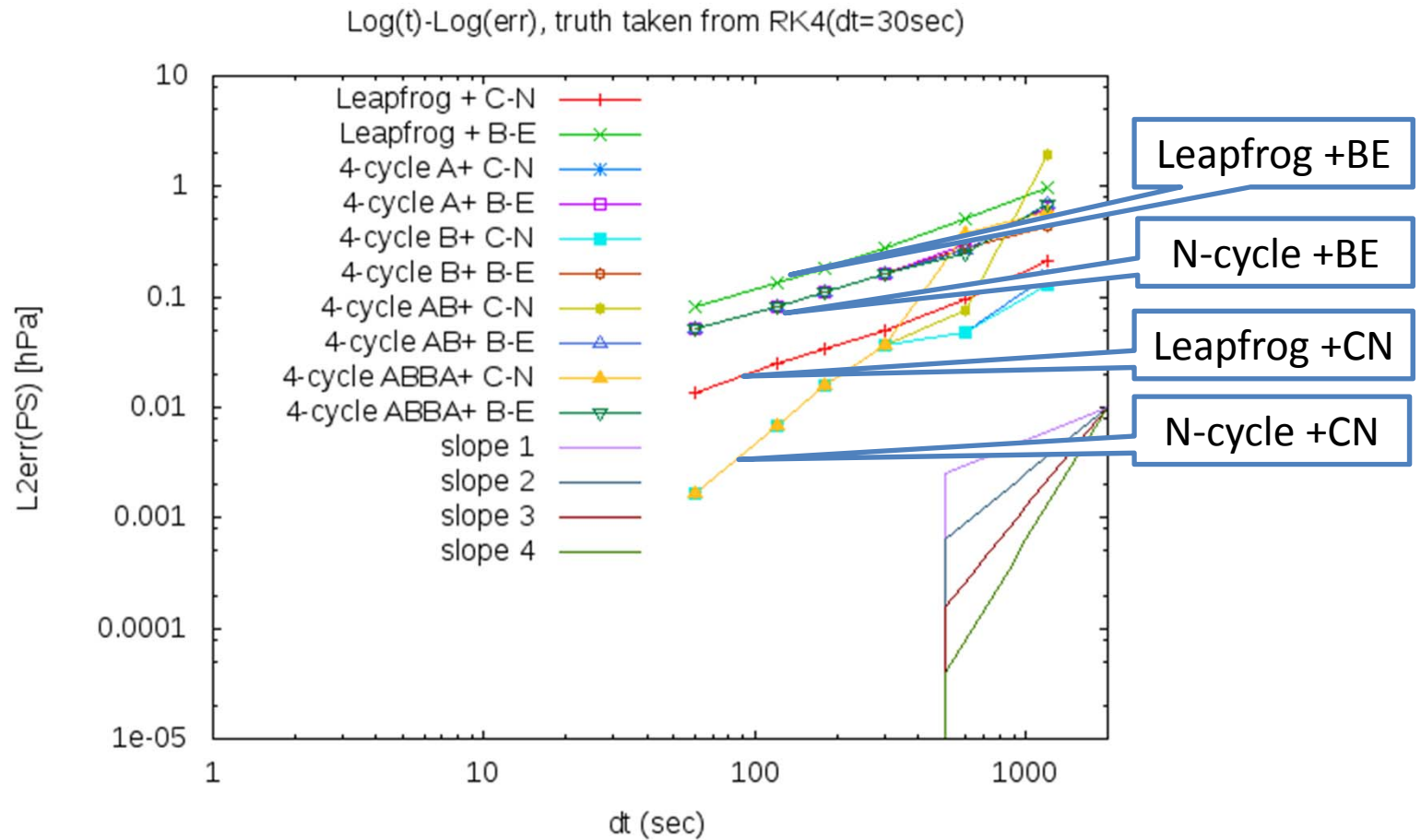
# Result : $L^2(PS)$ at $t=3\text{days}$



# Result : $L^2(PS)$ at $t=5\text{days}$



# Result : $L^2(PS)$ at $t=10\text{days}$



# Dynamical-core test

## Summary of the results

- For  $\Delta t \leq 10\text{min}$ , the order of Lorenz 4-cycle is  $3^{\text{rd}} \sim 4^{\text{th}}$
  - For a large  $\Delta t$ , A-B-B-A cycle is inferior to version A or version B, which contradicts with Lorenz (1971)'s claim.
  - Possible reasons:
    - Cancellation of truncation errors of version A and B does may hold because of the introduction of semi-implicit method.
    - Cancellation between A and B itself is not attained for the AGCM.
    - A-B-B-A cycle is more unstable for nonlinear models than A-only or B-only.
- ➔ To be examined

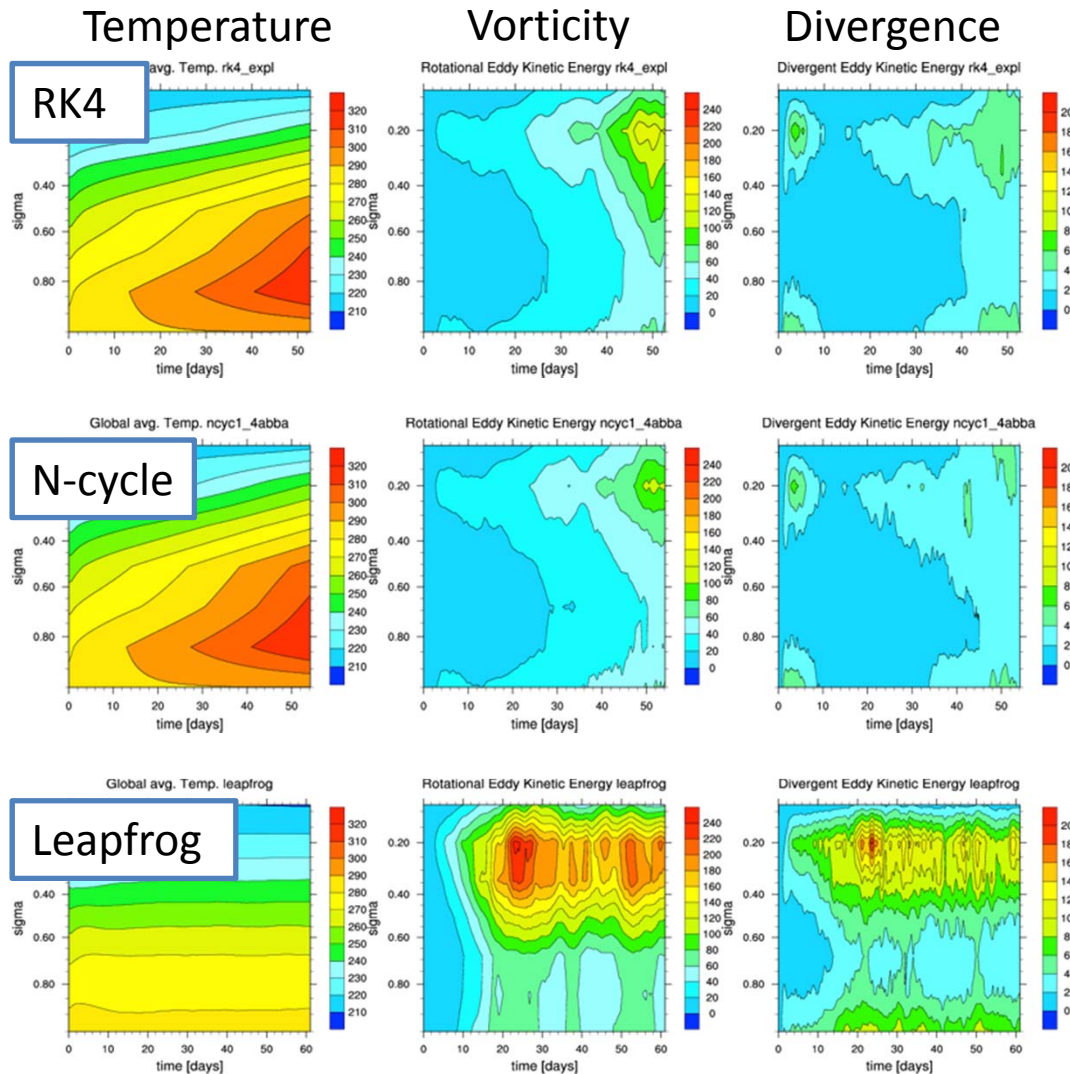
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# Inclusion of Physics

- Having proved that  $N$ -cycle works without physical parameterizations, I next tried to include physics in the  $N$ -cycle SPEEDY model.

# Problem encountered in the Last Semester: *N*-cycle and RK4 blow up when run with physics on



➔ Resolved:  
It was a Stupid  
Bug !



# Stability

- After fixing the bug, I tried to find the largest  $\Delta t$  with which the  $N$ -cycle can be integrated stably.

Bad news: the largest stable  $\Delta t$

- for  $N$ -cycle : 15 mins, whereas
- for Leapfrog: 40mins.

Possible remedy (with some accuracy degradation) :

- the largest stable  $\Delta t$  can be made 30mins
- by using Backward Euler for the gravity-wave part (instead of the default Crank-Nicolson)
- $N$ -cycle can be integrated with  $\Delta t=15$ mins. for at least 100 years (=3,504,000 steps)

# Comparison of Climatology (= long-term mean)

- In climate application, it is important that the long-term mean of the atmospheric state (called *climatology*) does not change.
- → Statistically compared the climatologies using Welch's *t*-test.
- Examined Climatologies:  
two seasons (DJF: 12-2 and JJA: 6-8), each from 3 models:
  - Leapfrog with  $dt=40$ mins (default)
  - 4-cycle (Crank-Nicolson for gravity-wave)  $dt=15$ mins
  - 4-cycle (Backward Euler for gravity-wave)  $dt=30$ mins

# *t*-test for the difference of climatologies

- Experimental design:
  - Run all the models from the same initial condition (state of rest)
  - Integrate for 30 years.
  - Discard the first 10 years as a spin-up.
  - Use the remaining 20 years as the samples.
- Null-hypothesis:

climatologies (= sample means) are taken from the same population.
- The probability of the differences between two sampled climatologies being larger than the observed difference under the null-hypothesis is computed.
- If this probability is larger than 95%, then the null-hypothesis is not rejected.

# Results for T, Z, U and V, MSLP and OLR: no significant difference !

- all results are uploaded on

[http://www.atmos.umd.edu/~dhotta/speedy\\_clim2/clim.html](http://www.atmos.umd.edu/~dhotta/speedy_clim2/clim.html)

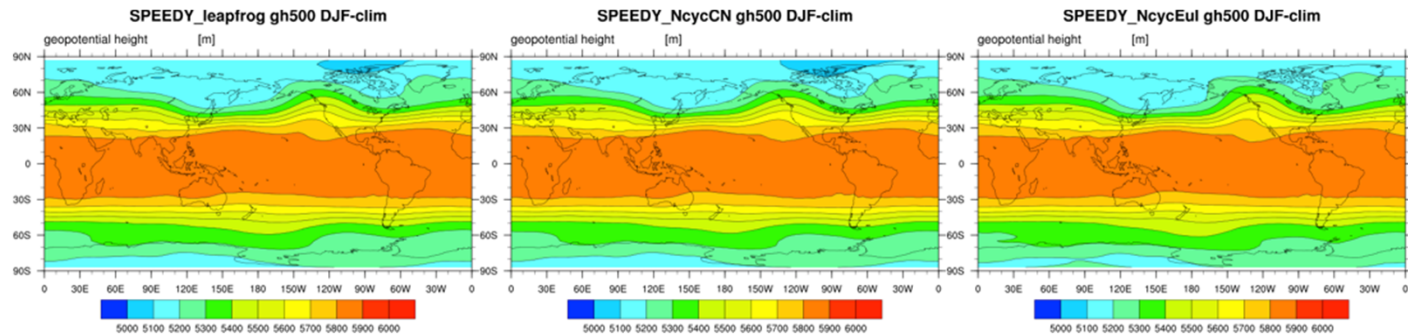
Example:  
Z500

Leapfrog +  
CrankNicolson

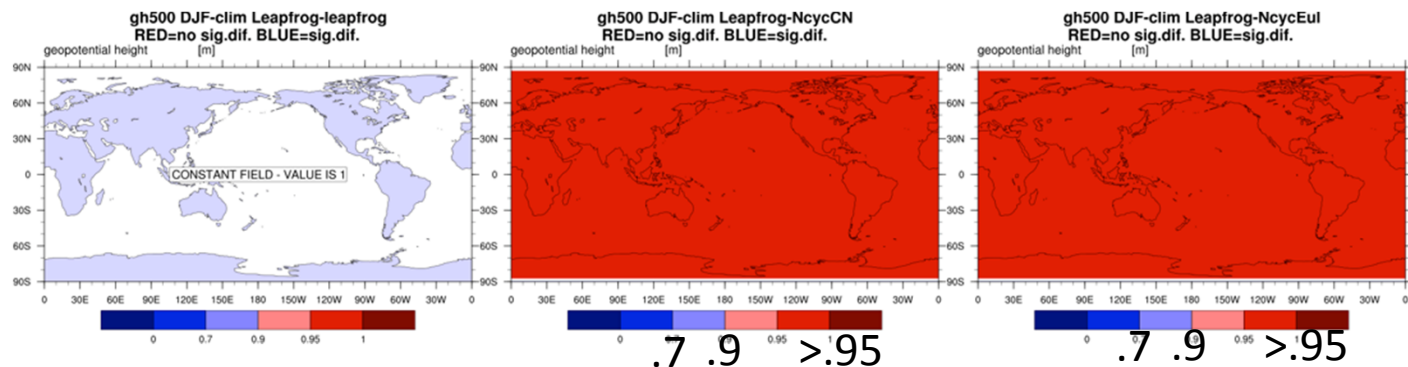
4-cycle +  
CrankNicolson

4-cycle +  
Backward Euler

climatology



Probability  
of null-  
hypothesis  
not being  
rejected



# Conclusion for Phase I

- Designed semi-implicit version of  $N$ -cycle
  - Analyzed stability using a toy-model
  - 4-cycle is the most stable
  - For semi-implicit method, Crank-Nicolson is more accurate than Backward-Euler but is more unstable
- Verification through Dynamical-Core test:
  - 4-cycle with Crank-Nicolson exhibits order-of-accuracy which is higher ( $2^{\text{nd}} \sim 3^{\text{rd}}$ ) than filtered Leapfrog ( $1^{\text{st}}$ -order)

# Schedule for Phase 1: Planned and Actual

## Planned

- Implement RK4 and  $N$ -cycle, **Nov.**
- Write the mid-year report, prepare the oral presentation, **Dec.**
- Switch-off physical parameterizations, prepare flat topography, **Jan.**
- perform the dynamical core tests. **Feb.**

## Actual

- Formulate semi-implicit  $N$ -cycle, **Oct.**
- Implement RK4 and  $N$ -cycle, **Nov.**
- Switch-off physical parameterizations, prepare flat topography, **Dec.**
- perform the dynamical core tests, **Dec.**
- Write the mid-year report, prepare the oral presentation, **Dec.**
- Coded a bug, and fixed the bug, **Jan. (winter break)**
- Compare climatologies, performed statistical test, **Feb.**

Phase II:  
Empirical characterization of Model  
Errors

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# Phase 2: Approach

- Objective: Characterize the model errors due to temporal discretizations
- Take the Truth from NCEP/NCAR reanalysis (Kalnay et al. 1996)

NCEP=National Centers for Environmental Prediction

NCAR=National Center for Atmospheric Research

- Extract model errors by applying the method of Danforth et al. (2007) to the models with:
  1. the original scheme (Leap-Frog;  $M^{LF}$ )
  2. Lorenz N-cycle scheme ( $M^{NCYC}$ )
- (time permitting) Correct the model errors on-line during the course of model integration  
(→ Phase 3&4)

# Phase 2: Algorithm

1. Generate initial values from the Truth (NCEP/NCAR reanalysis)
2. Perform short-range forecasts using the 2 models ( $M^{LF}$ ,  $M^{NCYC4}$ ) from the initial conditions
3. find the bias of the model errors for each model
4. Build the covariance matrix

$$\left\langle (x(t) - \bar{x}) (M^{true}(x(t)) - M^{LF}(x(t)))^T \right\rangle$$

5. Extract the dominant modes by conducting SVD

# Interpolation from NCEP/NCAR grid to SPEEDY grid

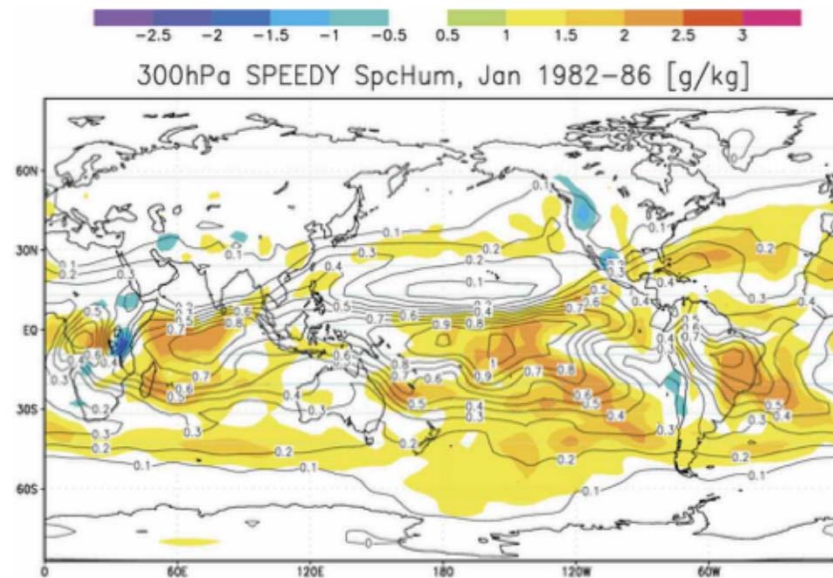
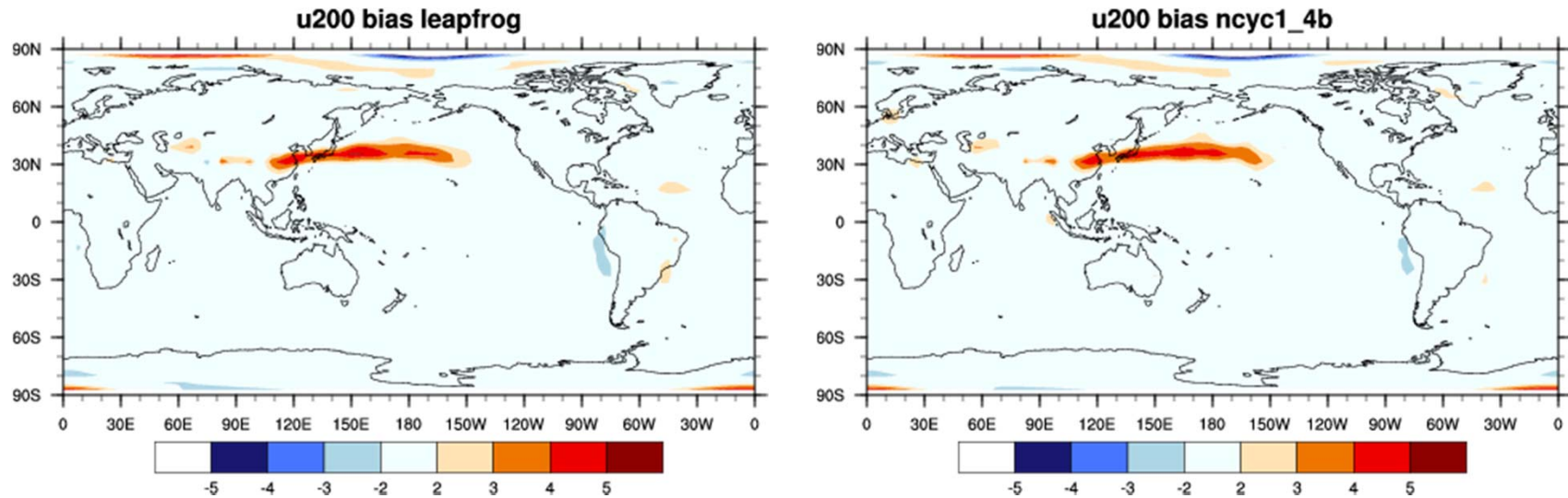
- Obtained the original code (used in Danforth et.al (2007)) and wrote a code that does exactly the same operation
- Original code:
  - written in MatLab script
  - performs Simple linear interpolation in 3D
- My code:
  - wrote in NCL (NCAR Command Language)
  - Basically a line-by-line translation from MatLab to NCL
- Validation:
  - Method:
    - produce data on SPEEDY grid from NCEP/NCAR data using the original code and my code
    - compare the two outputs, one from the original, the other from my code
  - Result: the two outputs agreed within single-precision rounding error

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# Model Error Bias : Results

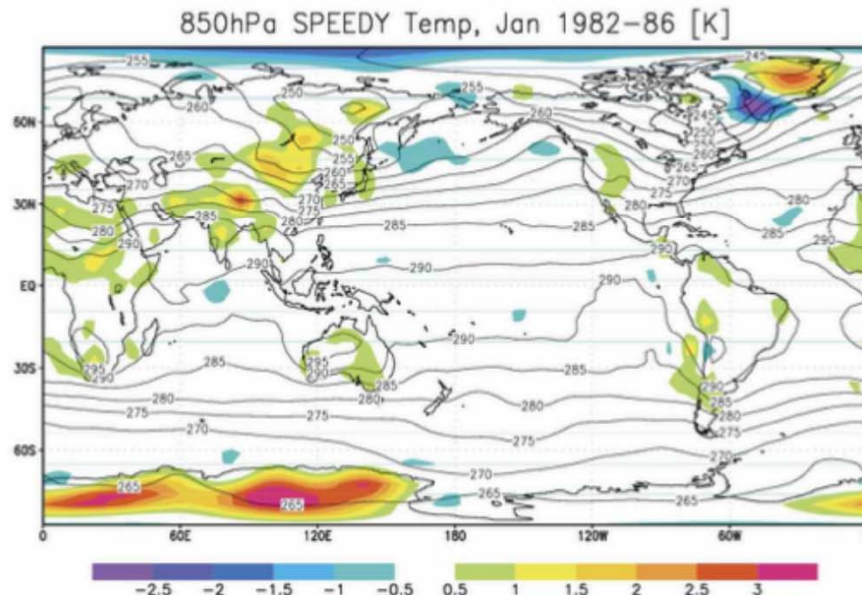
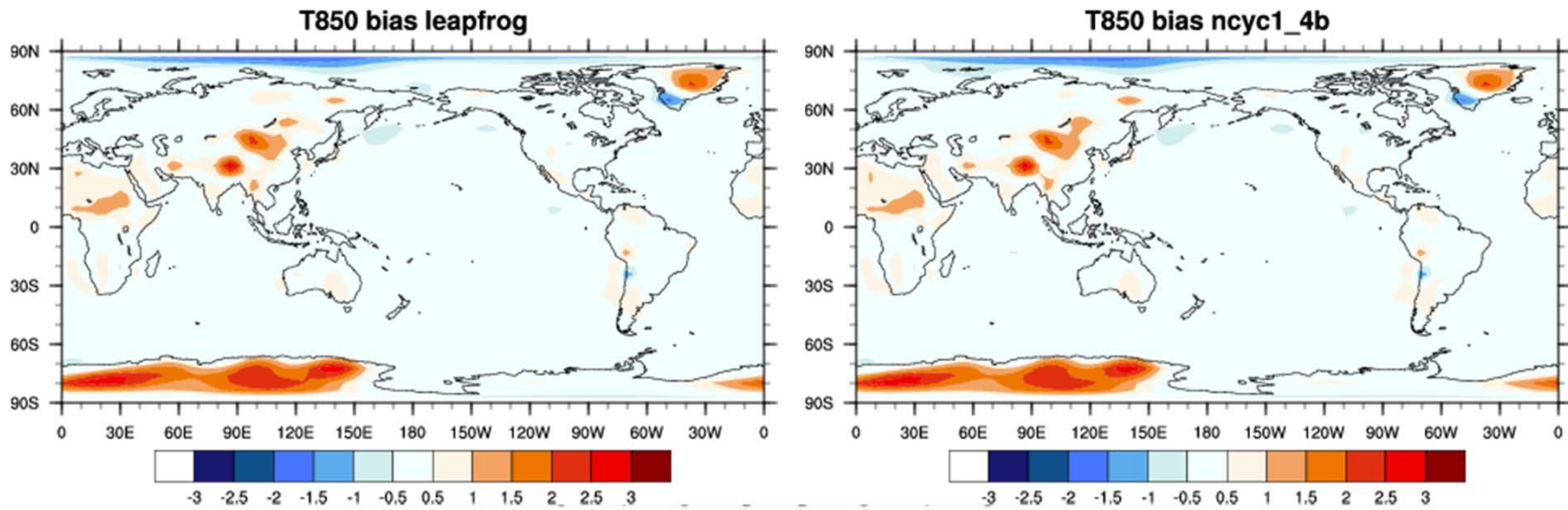
## Zonal wind at 200hPa



From Danforth et.al (2007)

# Model Error Bias : Results

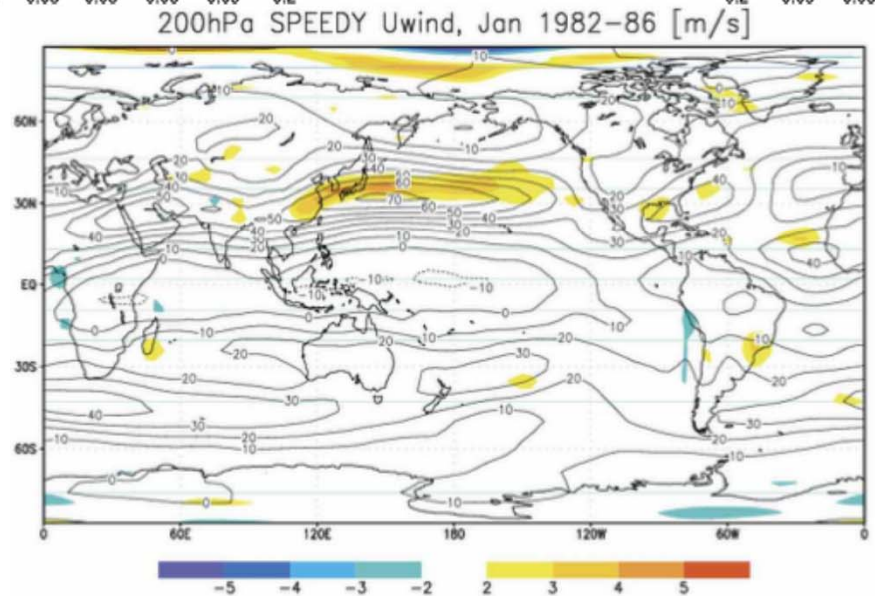
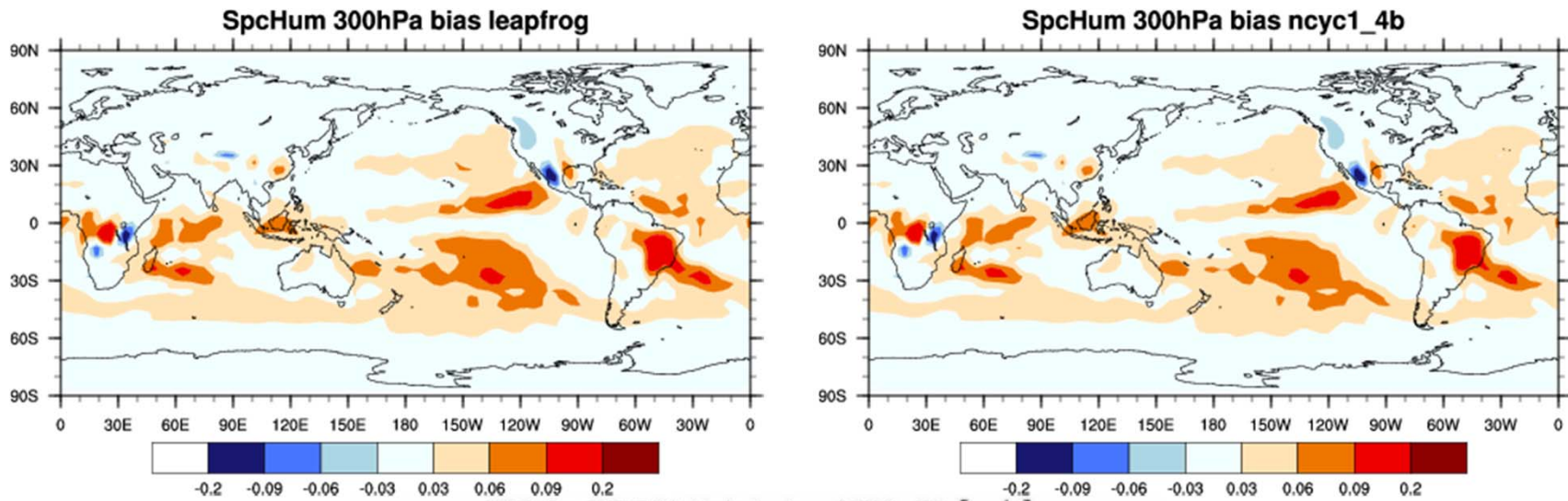
## Temperature at 850hPa



From Danforth et.al (2007)

# Model Error Bias : Results

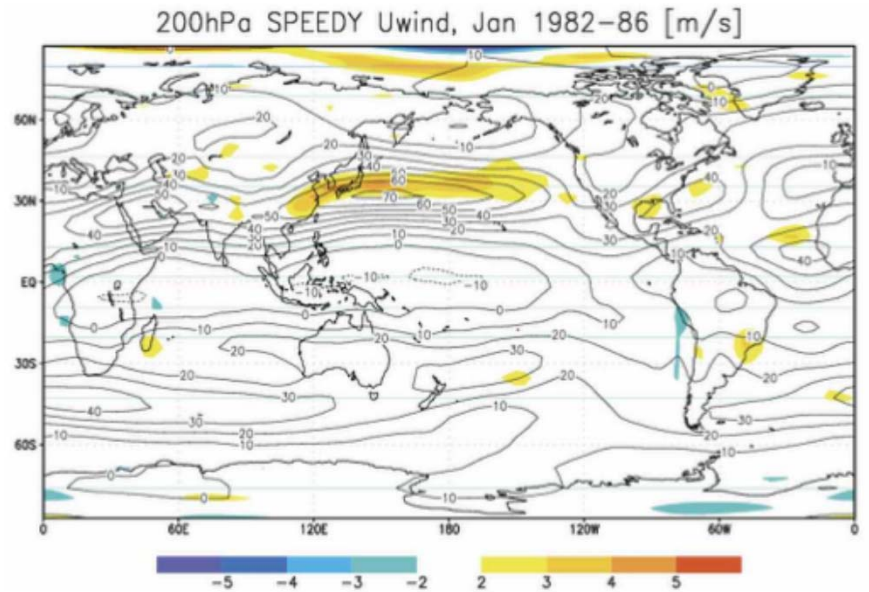
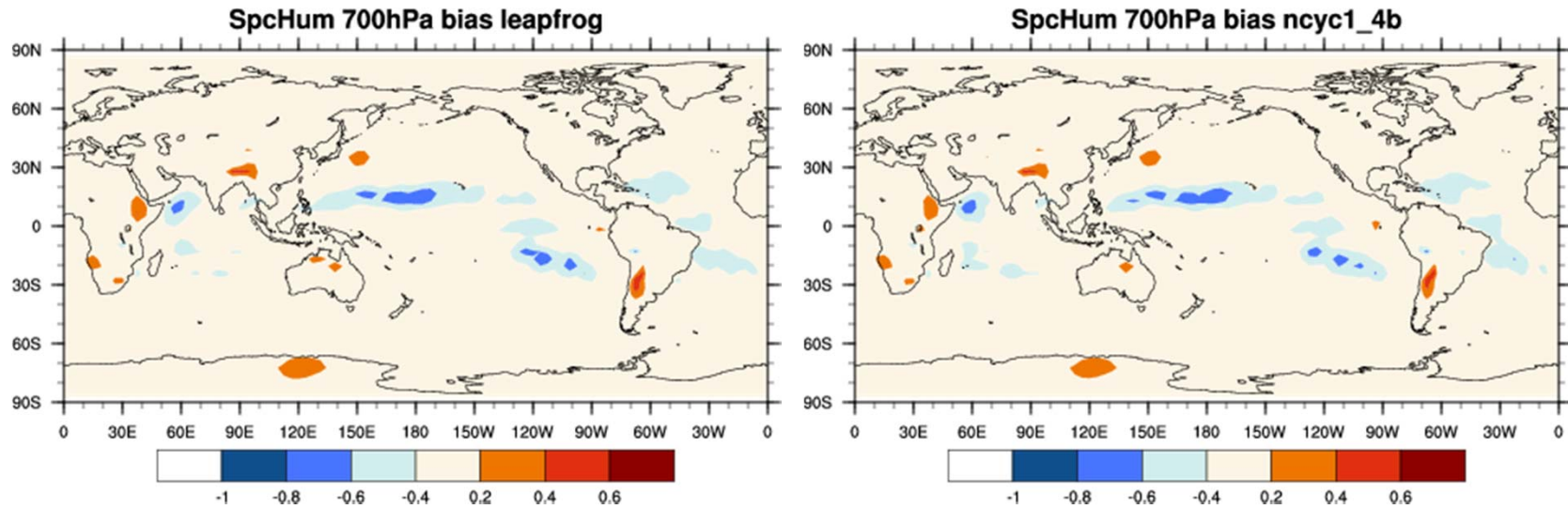
## Specific Humidity at 300hPa



From Danforth et.al (2007)

# Model Error Bias : Results

## Specific Humidity at 700hPa



From Danforth et.al (2007)



# Model Error Bias : Summary

- For all variables, Leapfrog and  $N$ -cycle produce almost identical bias pattern
  - Interpretation:  
bias is dominated by the model's deficiencies associated with physical parameterizations
- They are consistent with Danforth et.al (2007)

# Plan for Phase II

## In Danforth et.al (2007),

1. Error samples  $\{\delta x\}$  are produced with the original model  $M$
2. Bias  $\langle \delta x \rangle$  is computed
3. The model is modified by incorporating nudging to  $-\langle \delta x \rangle$  to yield  $M^+$
4. Error samples  $\{\delta x^+\}$  are resampled using the de-biased model  $M^+$
5. Diurnal errors are extracted from  $\{\delta x^+\}$  by performing EOF analysis and retaining the first two dominant modes
6. The de-biased model  $M^+$  is again modified by incorporating nudging to negative of diurnal biases to yield  $M^{++}$
7. Error samples  $\{\delta x^{++}\}$  are again resampled using  $M^{++}$
8. Perform SVD analysis to extract dominant co-variation of model error  $\delta x^{++}$  and the anomaly of the model states

## while, in my original plan,

1. Error samples  $\{\delta x\}$  are produced with the original model  $M$
2. Bias  $\langle \delta x \rangle$  is computed
3. The model is modified by incorporating nudging to  $-\langle \delta x \rangle$  to yield  $M^+$
4. Error samples  $\{\delta x^+\}$  are resampled using the de-biased model  $M^+$
5. Diurnal errors are extracted from  $\{\delta x^+\}$  by performing EOF analysis and retaining the first two dominant modes
6. The de-biased model  $M^+$  is again modified by incorporating nudging to negative of diurnal biases to yield  $M^{++}$
7. Error samples  $\{\delta x^{++}\}$  are again resampled using  $M^{++}$
3. Perform SVD analysis to extract dominant co-variation of model error  $\delta x^{++}$  and the anomaly of the model states
4. Validation: Results are compared with Danforth et.al (2007)

# Plan for Phase II

- Danforth et.al (2007) involves much more tricks than I originally planned.
- Reproducing all the procedures in Danforth et.al (2007) is impossible given that I have only 2-weeks left.



- I continue the original plan, but modify the Validation part.
- New Validation:
  - Check orthogonality between the extracted modes

# Conclusion for Phase II

- Generated samples of model error
- Computed biases for Leapfrog and Lorenz 4-cycle, compared them with Danforth et.al (2007)
- Result:
  - No significant bias improvements by using a better temporal scheme. Perhaps dominated by physics errors
  - Consistent with Danforth et.al (2007)
- TODO (in two weeks):
  - SVD analysis to identify the dominant co-varying modes between model state anomaly and model error
  - Comparison with Danforth et.al (2007)

# Schedule for Phase 2: Planned and Actual

## Planned

- Generate initial values from the NCEP/NCAR reanalysis, end of **Feb.**
- build the bias and covariance matrix, **Mar.**
- Code and test a program for SVD, **Apr.**
- Compare the model errors for the new and the original schemes, **May.**
- Write the final report (paper draft), **May.**

## Actual

- Generate initial values from the NCEP/NCAR reanalysis, end of **Feb.**
- Compute the bias, **Mar.**
- Plot the bias and compare it with Danforth et.al, **Apr. (← Now I'm here)**
- Code and test a program for SVD, **May.**
- Compare the model errors for the new and the original schemes, **May.**
- Write the final report (paper draft), **May.**

# Outcome/Deliverables

## Phase 1:

- Upgraded code for SPEEDY model ✓
  - subroutines for Lorenz N-cycle and 4<sup>th</sup> order Runge-Kutta
- Test-case results for the SPEEDY model (both for the original scheme and the new schemes) ✓

Available at <https://code.google.com/p/speedy-lorenz-ncycle/>

## Phase 2:

- Archive of the model errors ✓
- Bias of model errors ✓
- Pairs of Singular Vectors for the model state and the model error
- Code for performing SVD

To be completed in two weeks



# Bibliography

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- Lorenz, Edward N., 1971: An N-cycle time-differencing scheme for stepwise numerical integration. *Mon. Wea. Rev.*, **99**, 644–648.

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- Kucharski F, Molteni F, and Bracco A, 2006: Decadal interactions between the western tropical Pacific and the North Atlantic Oscillation. *Clim. Dyn.*, **26**, 79-91

## **SPEEDY-LETKF**

- Miyoshi, T., 2005: *Ensemble Kalman filter experiments with a primitive-equation global model*. Ph.D. dissertation, University of Maryland, College Park, 197pp.

## **Atmospheric GCM Dynamical Core test cases**

- Jablonowski, C. and D. L. Williamson 2006: A baroclinic instability test case for atmospheric model dynamical cores, *Q. J. R. Meteorol. Soc.*, **132**, 2943-2975

## **NCEP/NCAR reanalysis**

- Kalnay, E., and Coauthors, 1996: The NCEP/NCAR 40-Year Reanalysis Project. *Bull. Amer. Meteor. Soc.*, **77**, 437–471.

## **Model Error Correction**

- Danforth, Christopher M., Eugenia Kalnay, Takemasa Miyoshi, 2007: Estimating and Correcting Global Weather Model Error. *Mon. Wea. Rev.*, **135**, 281–299.
- Danforth, Christopher M., Eugenia Kalnay, 2008: Using Singular Value Decomposition to Parameterize State-Dependent Model Errors. *J. Atmos. Sci.*, **65**, 1467–1478.

back-up slides



# Swinging-pendulum problem

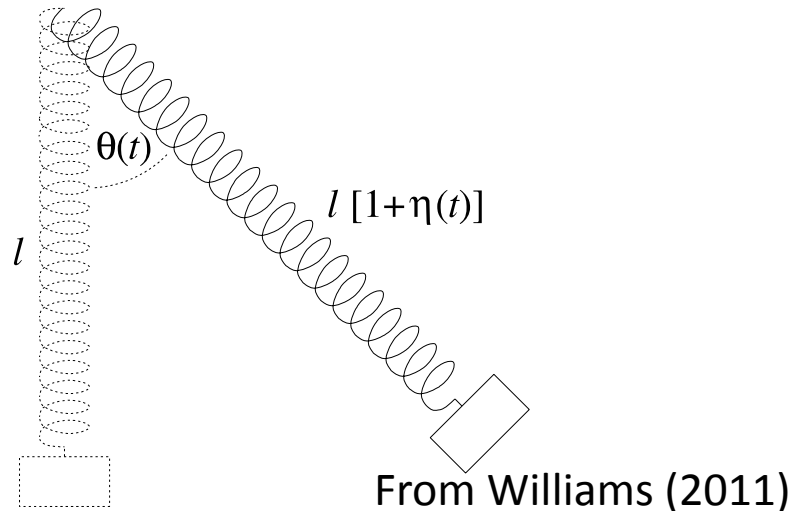


FIG. 7. Schematic diagram of the elastic pendulum, or swinging spring, showing the equilibrium position (dotted) and a non-equilibrium position defined by the values of  $\eta$  and  $\theta$  (solid).

$$\dot{\eta} = v_{\eta},$$

$$\dot{v}_{\eta} = -\omega_{\text{low}}^2(1 - \cos\theta) - \omega_{\text{high}}^2\eta + (1 + \eta)v_{\theta}^2,$$

$$\dot{\theta} = v_{\theta}, \quad \text{and}$$

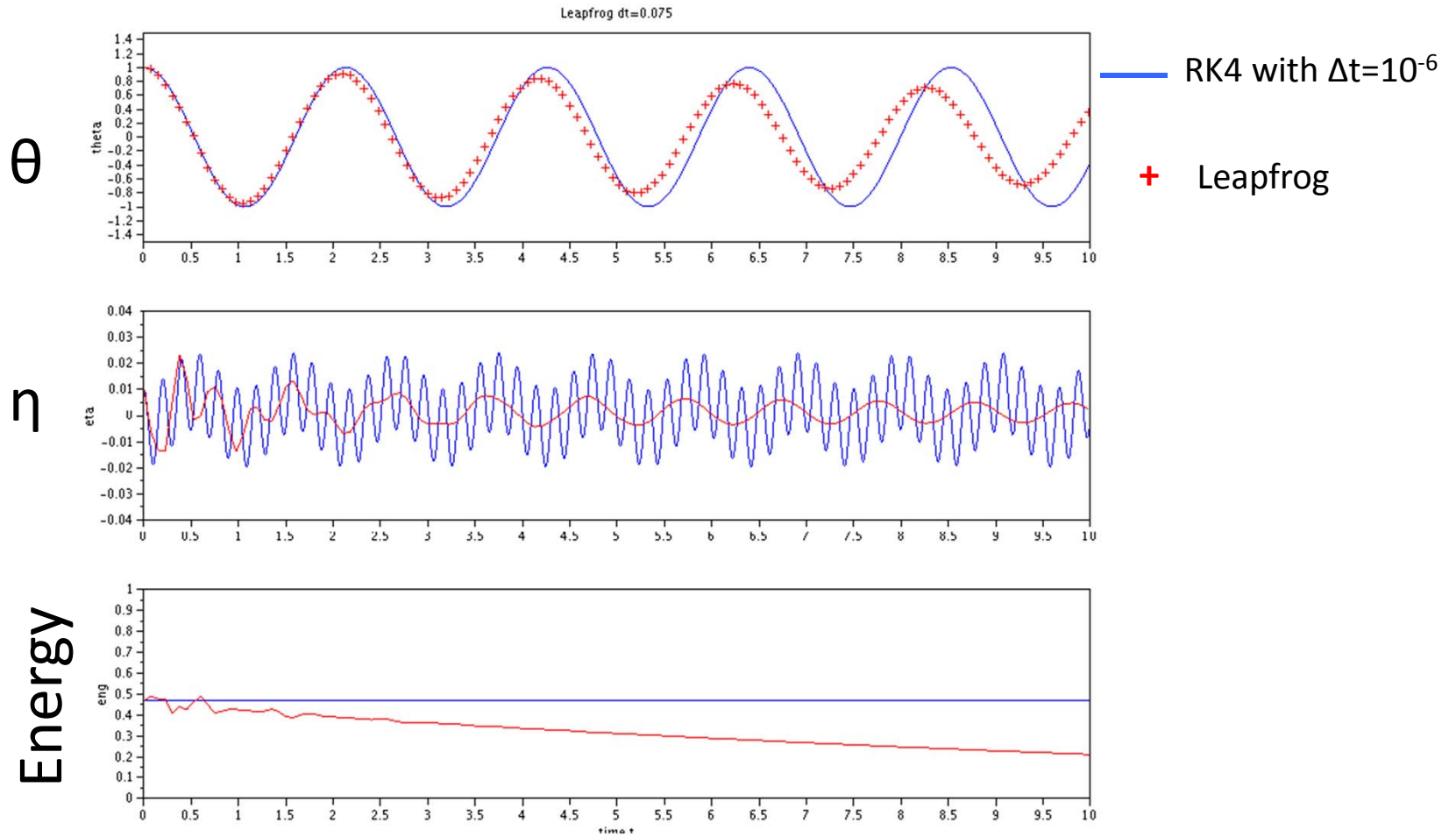
$$\dot{v}_{\theta} = \frac{-\omega_{\text{low}}^2 \sin\theta - 2v_{\eta}v_{\theta}}{1 + \eta}.$$

- A simple nonlinear test-bed for semi-implicit schemes.
- Fast oscillation = elastic spring
- Slow oscillation = pendulum
- Fast mode is treated implicitly, slow mode explicitly.

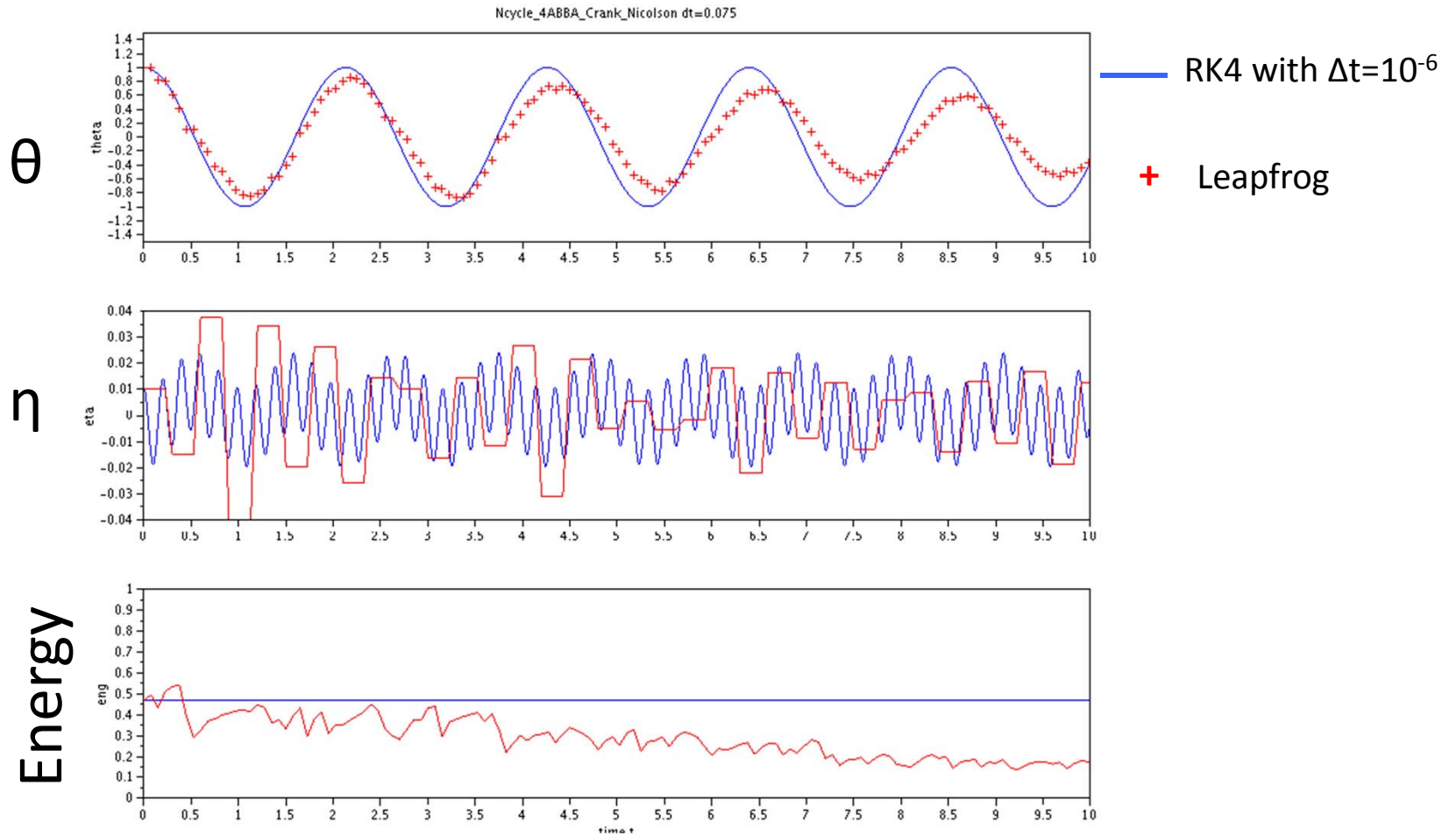
- For this test, I use  $\Delta t=0.075$  which gives

$$\omega_{\text{LOW}}\Delta t=0.225, \quad \omega_{\text{HIGH}}\Delta t=2.25,$$

# Leapfrog with RA filter ( $\alpha=0.01$ )

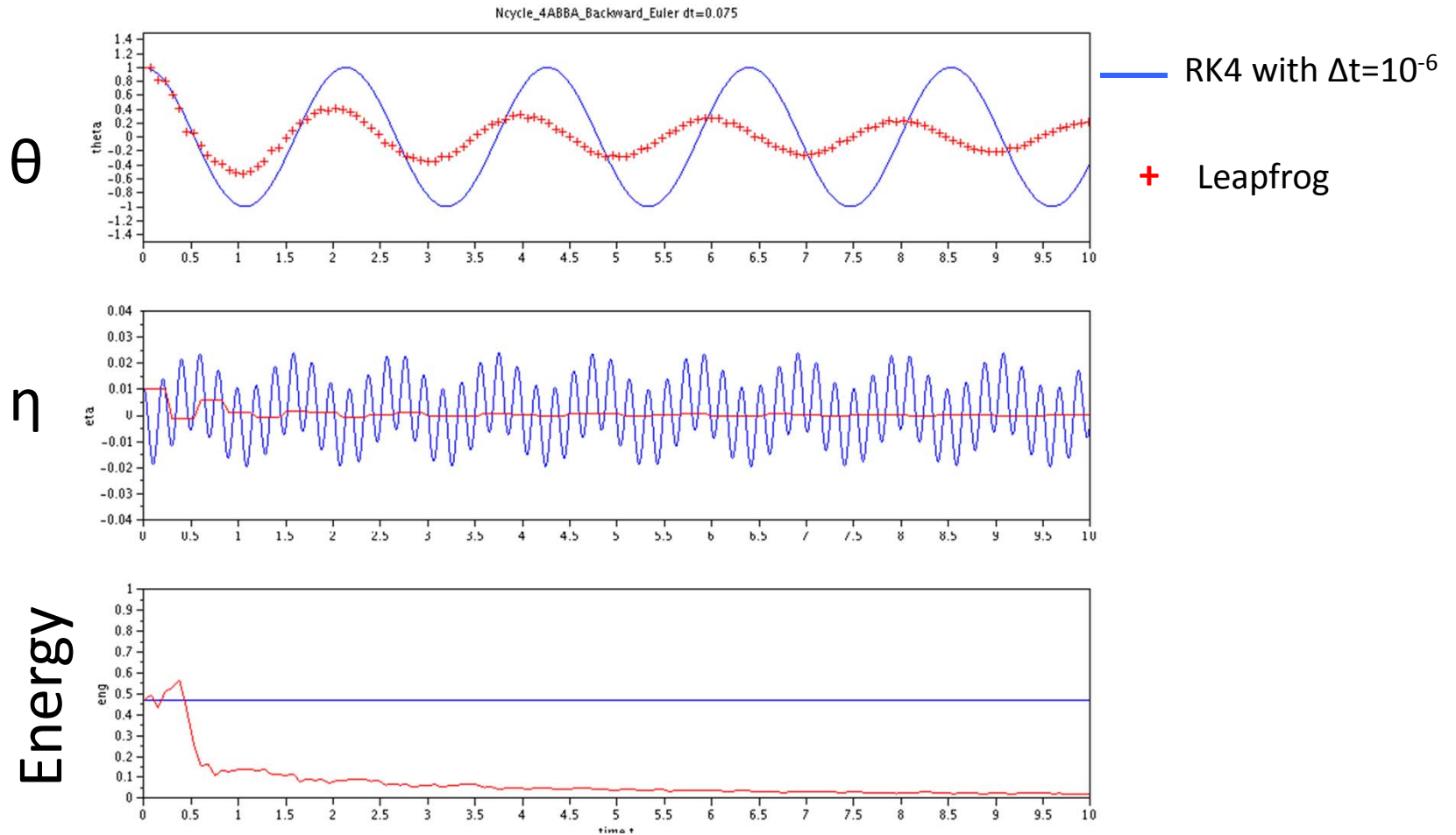


# 4-cycle with semi-implicit correction on every 4 steps (Crank-Nicolson)

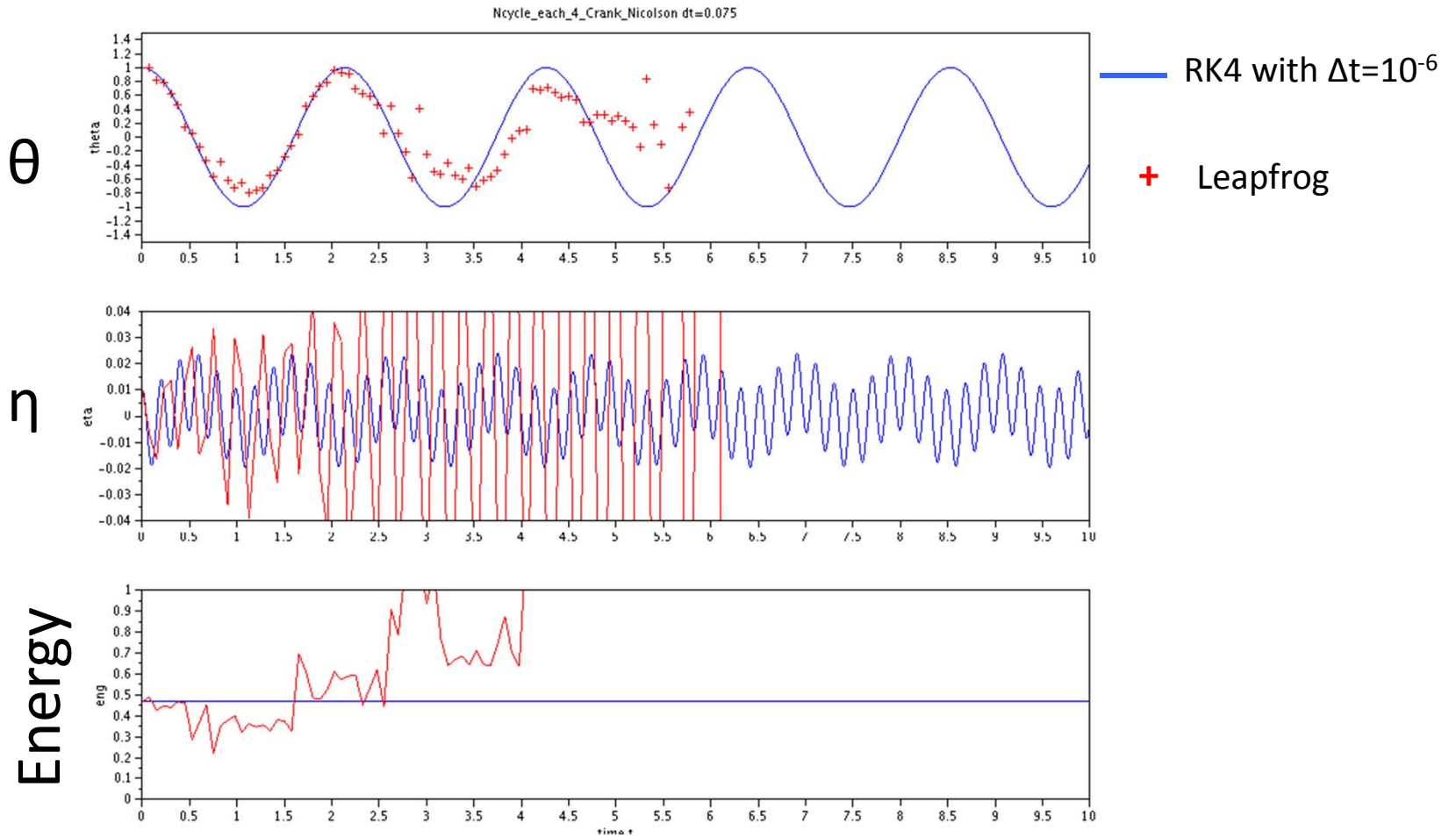


# 4-cycle with semi-implicit correction on every 4 steps (Backward)

➔ stable but very dissipative

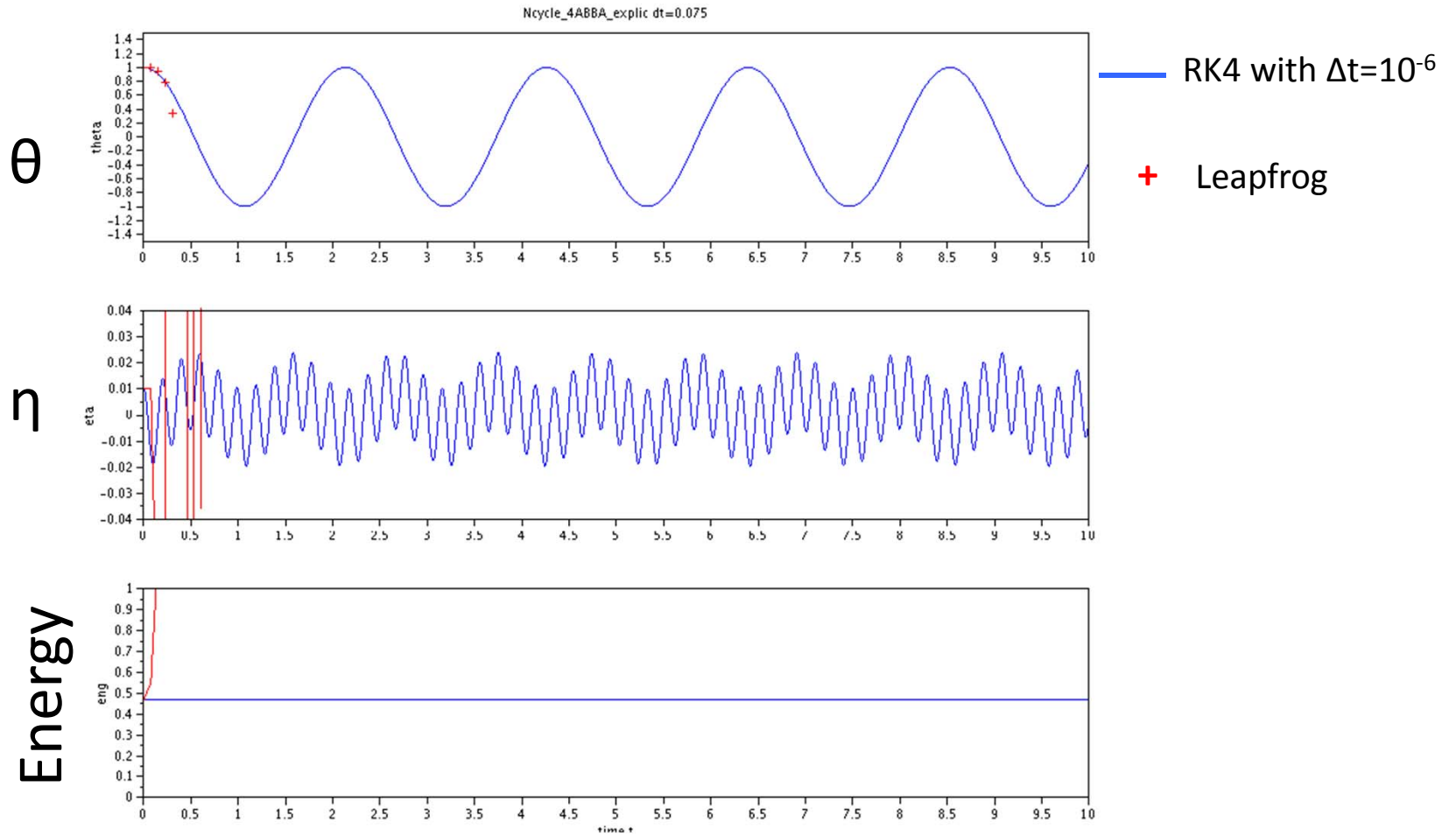


# 4-cycle with semi-implicit correction on every time step $\rightarrow$ unstable



# Explicit 4-cycle

→ unstable



# Summary

Consistent with the linear analysis,

- N-cycle with semi-implicit on each time step
  - ➔ unstable
- N-cycle with semi-implicit on every N steps
  - ➔ Crank-Nicolson: stable, but comparable accuracy with Leapfrog
  - ➔ Backward Euler: stable, very dissipative

## Reference:

- Williams, P. D. 2011: The RAW Filter: An Improvement to the Robert–Asselin Filter in Semi-Implicit Integrations, *Mon. Weath. Rev.*, **139**, 1996--2007



Runge-Kutta 4<sup>th</sup>-order scheme  
with semi-implicit correction

# Introduction

- I implemented semi-implicit Runge-Kutta 4<sup>th</sup>-order scheme to SPEEDY model and found that
  - for Crank-Nicolson, the model blows up, even for very small dt (1min).
  - for Backward Euler, the model is too diffusive that the baroclinic-wave dynamical core test fails to produce the baroclinic wave.
- Explicit Runge-Kutta 4<sup>th</sup>-order scheme with small dt (5min) works fine for the dynamical core test.
- → examine stability using the toy-model

# Method

- Following Durran (1991, 1999; MWR) and Williamson (2011; MWR), apply semi-implicit modification to the second term of the equation:

$$\frac{du}{dt} = i\omega_L u + i\omega_H u$$

- Examine the modulus of Amplification factor  $A$ .
- If  $|A| < 1$ , the scheme is stable.
- Range of interest:

$$\omega_L \Delta t < 0.5, \quad \omega_H \Delta t > 2$$

i.e., CFL condition is met for low-frequency part but is violated for high-frequency part.

# Algorithm:

## RK4 for $\omega_L$ , Crank-Nicolson for $\omega_H$

$$\begin{aligned}h^1 &= i\omega_L u^n \\u_2^* &= u^n + \frac{\Delta t}{2} h^1 \\h^2 &= i\omega_L u_2^* \\u_3^* &= u^n + \frac{\Delta t}{2} h^2 \\h^3 &= i\omega_L u_3^* \\u_4^* &= u^n + \Delta t h^3 \\h^4 &= i\omega_L u_4^* \\G &= \frac{1}{6}(h^1 + 2h^2 + 2h^3 + h^4) \\ \frac{u^{n+1} - u^n}{\Delta t} &= G + \{ \beta i\omega_H u^{n+1} (1 - \beta) i\omega_H u^n \}\end{aligned}$$

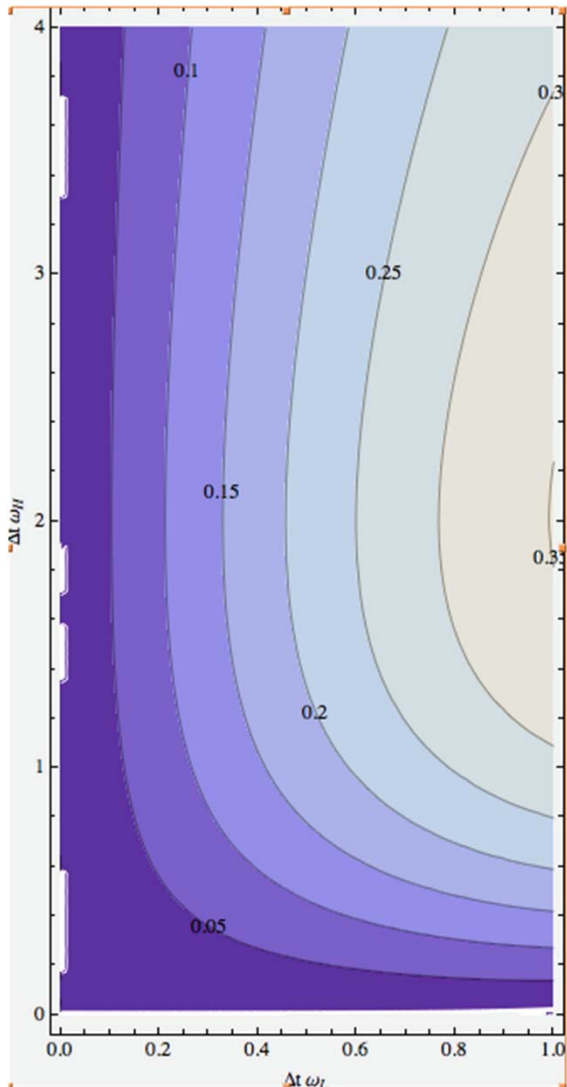
Truncation Error:

$$\frac{u^{n+1} - u^{Exact}}{u^n} = \frac{1}{2} \omega_H ((1 - 2\beta)\omega_H - 2(\beta - 1)\omega_L) \Delta t^2 + O(\Delta t^3)$$

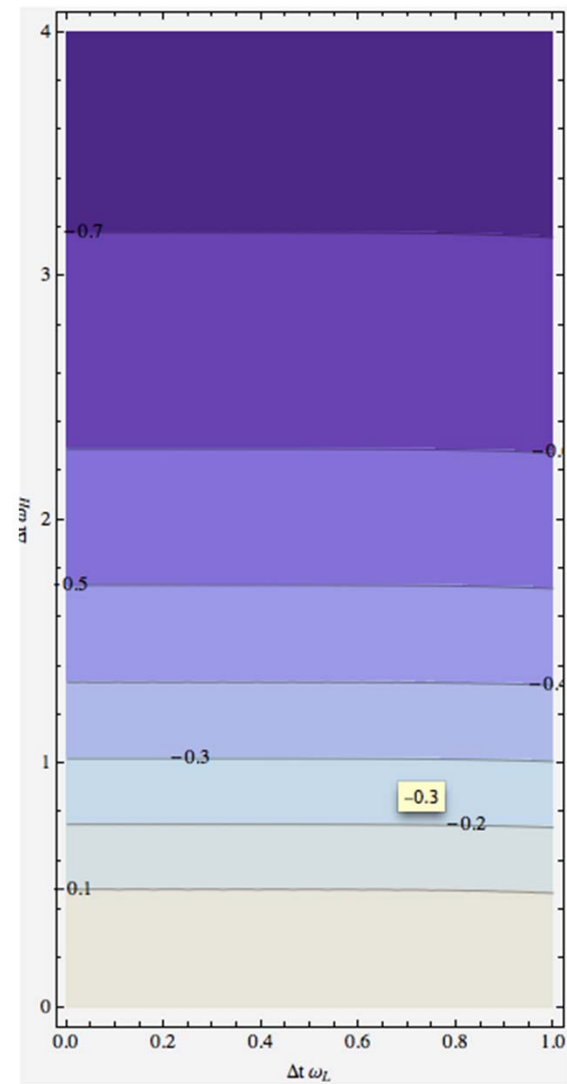
the accuracy is only 1<sup>st</sup> order

# Plot of $|A|^{-1}$

$\beta=1/2$ : Crank-Nicolson  
-> Absolutely unstable



$\beta=1$ : backward Euler  
-> Absolutely stable, but extremely dissipative



# Summary

- Runge-Kutta 4<sup>th</sup>-order scheme with semi-implicit time is
  - only of first order (with respect to fast modes)
  - absolutely unstable if Crank-Nicolson is used
  - absolutely stable if Backward Euler is used, but the numerical damping is too strong (more than halving on every time step)
- all of the above are consistent with what I found for the SPEEDY model.

# Conclusion

- Runge-Kutta 4<sup>th</sup>-order scheme with semi-implicit time-stepping for gravity waves is impossible (either unstable or too dissipative)
- The accuracy becomes only of 1<sup>st</sup> order.
- Since the motivation for implementing Runge-Kutta 4<sup>th</sup>-order scheme is to produce a reference solution, I will not try to resolve this issue, and use the explicit scheme with small  $dt$ .

Stability analysis of Forward Euler  
“split physics” for Leapfrog, Lorenz  
4-cycle and RK4



# Toy model: linear advection-diffusion equation

- Undiscretized equation:

$$\frac{du}{dt} = i\omega u - \beta u$$

- The first term of the RHS simulates the dynamics of AGCM, and the second the physics.

## Discretization: Leapfrog(dyn) + Euler(phys)

- Discretized equation:

$$\frac{u^{n+1} - u^{n-1}}{2\Delta t} = i\omega u^n - \beta u^{n-1}$$

- The amplification factor  $A$  satisfies the following equation:

$$\frac{A^2 - 1}{2\Delta t} = i\omega A - \beta$$
$$\Rightarrow A = i\omega\Delta t \pm \sqrt{1 - (\omega\Delta t)^2 - 2\Delta t\beta}$$

- (+) and (-) correspond, respectively, to physical and computational modes.
- The scheme is stable if the maximum of the moduli of them is less than 1.

# Discretization: Leapfrog for both dyn & phys

- Discretized equation:

$$\frac{u^{n+1} - u^{n-1}}{2\Delta t} = i(\omega + i\beta)u^n$$

- The amplification factor  $A$  satisfies the following equation:

$$\begin{aligned} \frac{A^2 - 1}{2\Delta t} &= i(\omega + i\beta)A \\ \Rightarrow A &= i(\omega + i\beta)\Delta t \pm \sqrt{1 - (\omega + i\beta)^2 \Delta t^2} \end{aligned}$$

- (+) and (-) correspond, respectively, to physical and computational modes.
- The scheme is stable if the maximum of the moduli of them is less than 1.

## Discretization: Lorenz 4-cycle (dyn) + Euler(phys)

- Discretized equation:

$$u^{n+N} = u^* - N\beta\Delta t u^n,$$

where  $u^* = u^n \left( \sum_{k=0}^N \frac{1}{k!} (iN\omega\Delta t)^k \right)$

- Amplification factor (per time step):

$$A = \left\{ \left( \sum_{k=0}^N \frac{1}{k!} (iN\omega\Delta t)^k \right) - \beta N \Delta t \right\}^{1/N}$$

## Discretization: Lorenz 4-cycle for both dyn & phys

- Discretized equation:

$$u^{n+N} = u^n \left( \sum_{k=0}^N \frac{1}{k!} (iN(\omega + i\beta)\Delta t)^k \right)$$

- Amplification factor (per time step):

$$A = \left( \sum_{k=0}^N \frac{1}{k!} (iN\omega\Delta t)^k \right)^{1/N}$$

# Discretization: RK4(dyn) + Euler(phys)

- Discretized equation:

$$u^{n+1} = u^* - \beta \Delta t u^n,$$

where  $u^* = u^n \left( \sum_{k=0}^N \frac{1}{k!} (i\omega \Delta t)^k \right)$

- Amplification factor :

$$A = \left( \sum_{k=0}^N \frac{1}{k!} (i\omega \Delta t)^k \right) - \beta \Delta t$$

# Discretization: RK4 for both dyn&phys

- Discretized equation:

$$u^{n+1} = u^n \left( \sum_{k=0}^N \frac{1}{k!} (i(\omega + i\beta)\Delta t)^k \right)$$

- Amplification factor (per time step):

$$A = \sum_{k=0}^N \frac{1}{k!} (i\omega\Delta t)^k$$

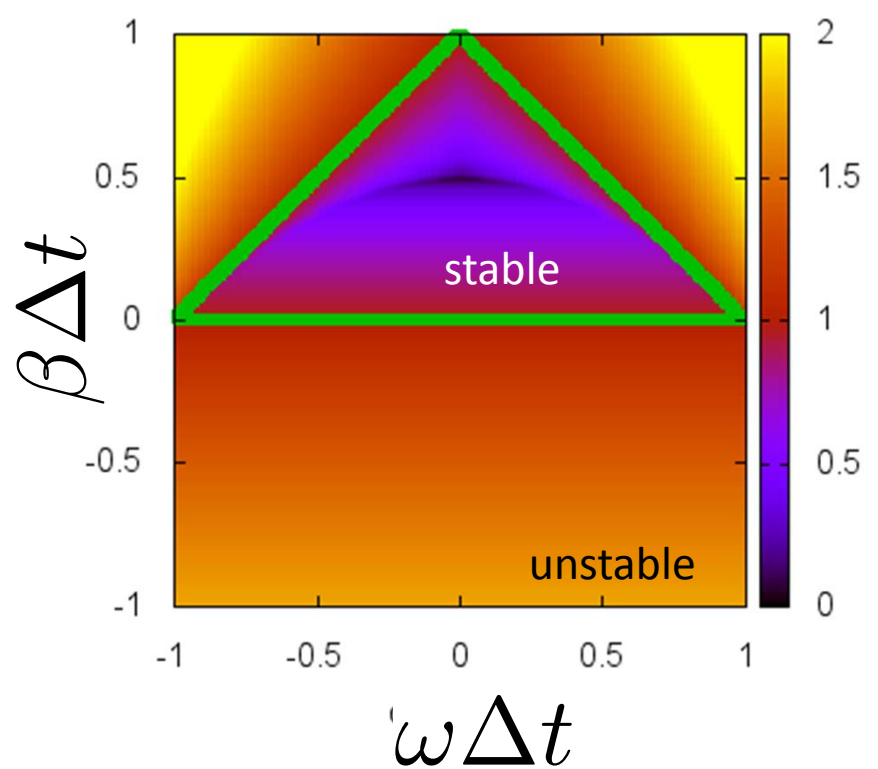
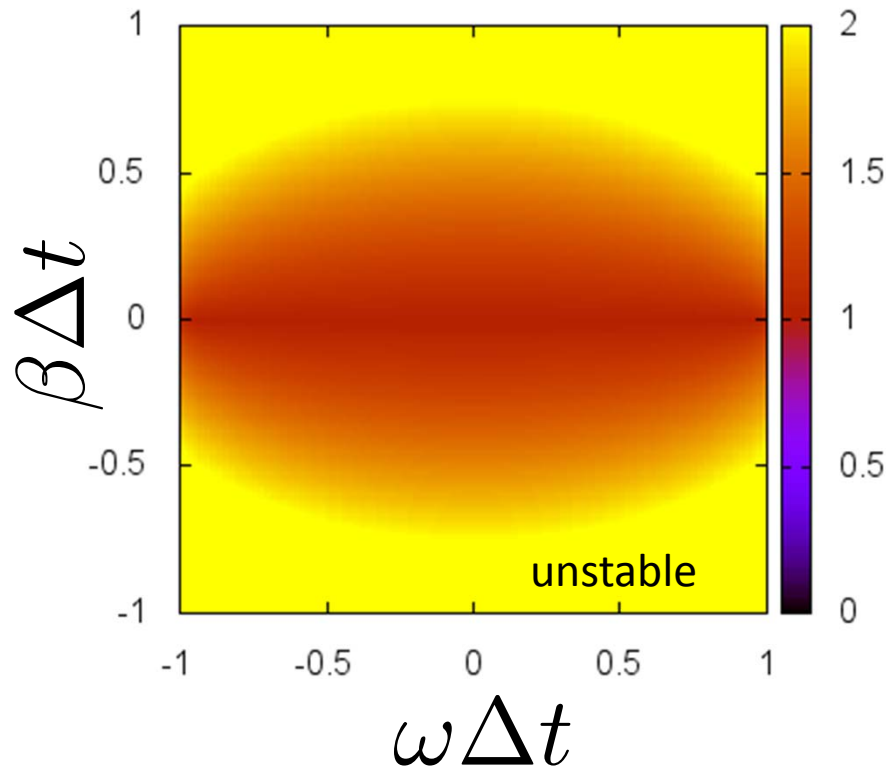
Leapfrog:  
“split physics” stabilizes the otherwise  
absolutely unstable scheme

**Leapfrog for both dyn & phys:  
absolutely unstable**

**Leapfrog (dyn) + Euler (phys):  
Stable within the triangle**

$\max(|A_{\text{phys}}|, |A_{\text{comp}}|)$  for Leapfrog (for dynamics and physics)

$\max(|A_{\text{phys}}|, |A_{\text{comp}}|)$  for Leapfrog + Euler (split physics)

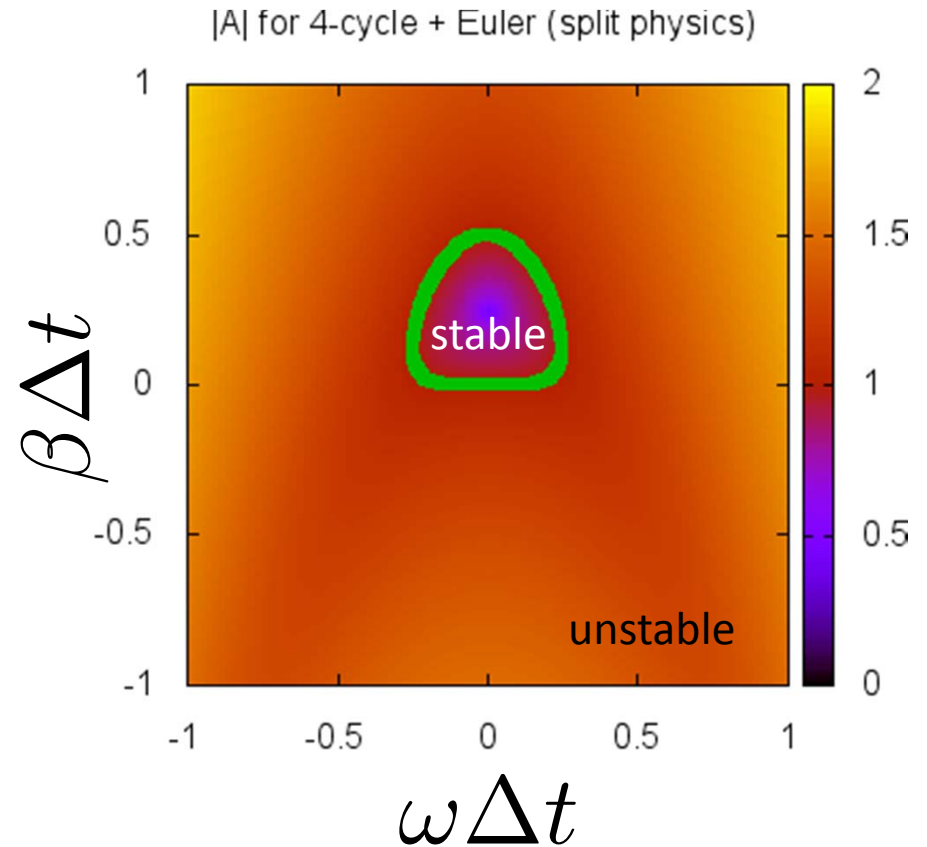
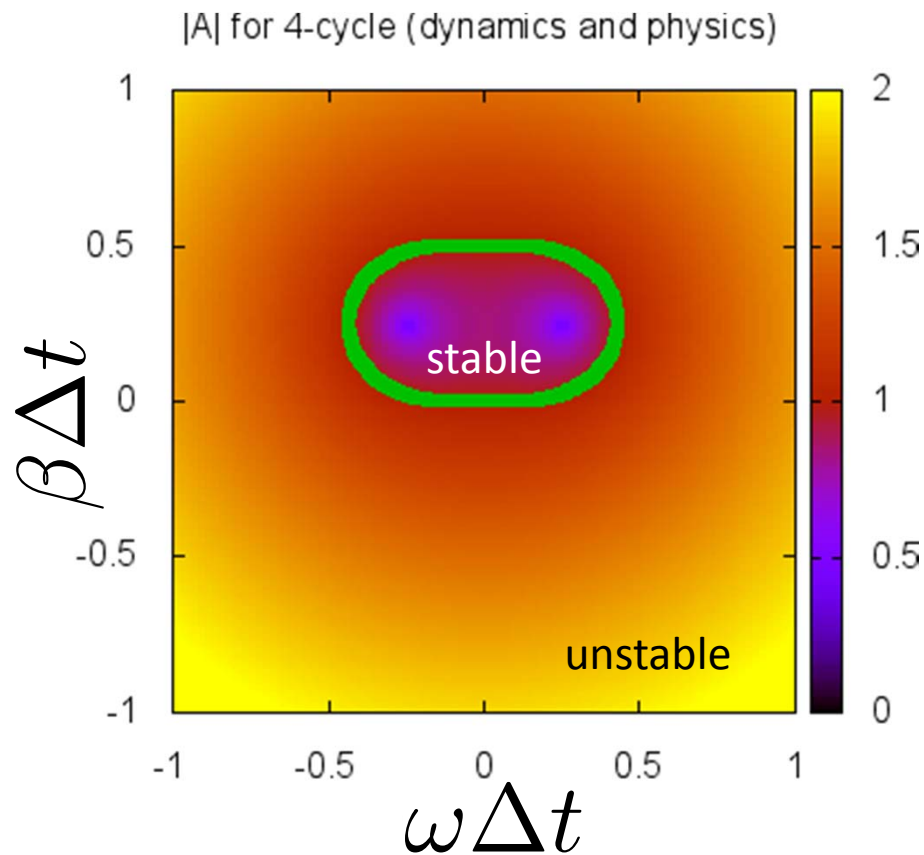




# Lorenz 4-cycle: “split physics” destabilizes the scheme

**Lorenz 4-cycle for dyn & phys:  
absolutely unstable**

**L4-cycle (dyn) + Euler (phys):  
Stable within the triangle**

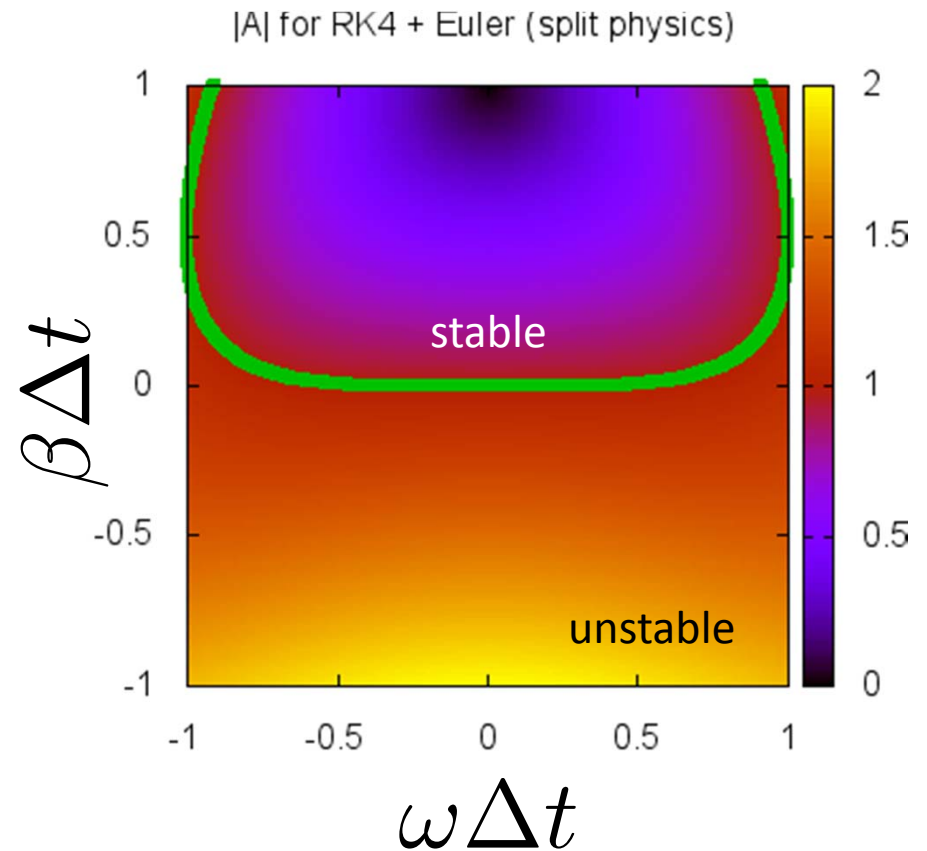
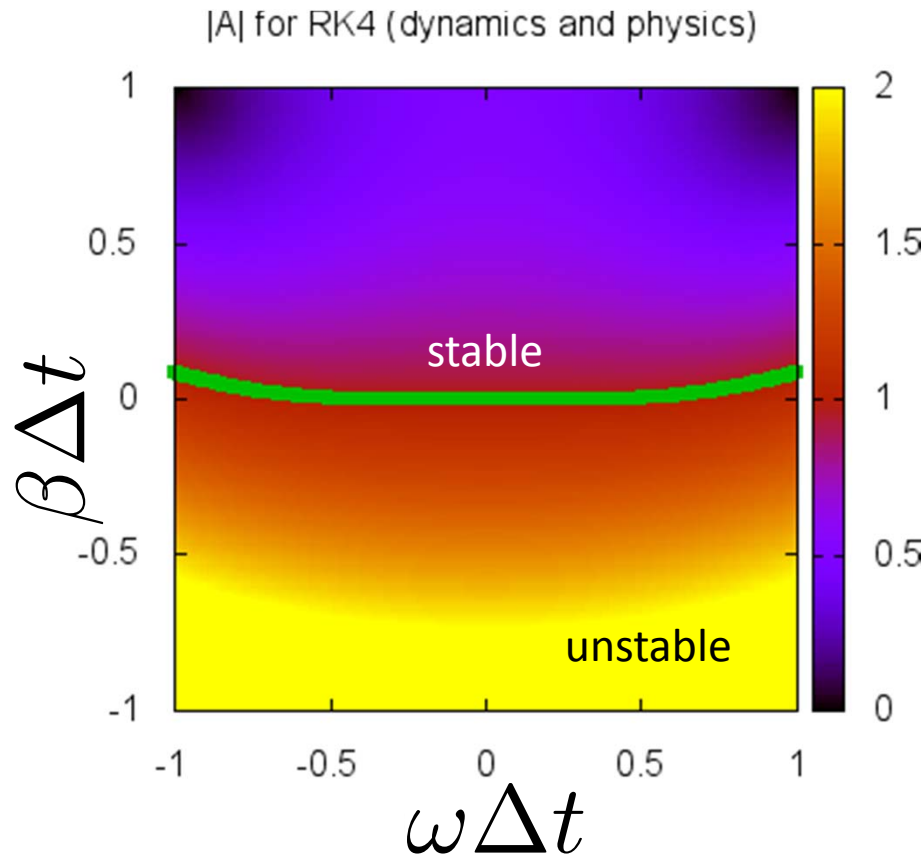


# RK4:

again, “split physics” destabilizes the scheme

**RK4 for dyn & phys:  
absolutely unstable**

**RK4 (dyn) + Euler (phys):  
Stable within the triangle**



# Summary

- For leapfrog, doing “split physics” works very well (stabilizes the scheme)
- However, for Lorenz 4-cycle (and equivalently, for RK4), “split physics” acts to destabilized the scheme.
- The latter is consistent with what I found by doing “split physics” with SPEEDY model.

# The Bug (1)

## Leapfrog (default)

```
PROGRAM agcm  
CALL iniall ()  
IDAY=0; CALL FORDATE()  
CALL STEPONE() ! 1st step by Euler Forward  
DO ! loop over a month  
  DO ! loop over a day  
    CALL FORDATE()  
    CALL STLOOP() ! integrate for a day  
  END DO  
END DO
```

## N-cycle (with bug)

```
PROGRAM agcm  
CALL iniall ()  
  
DO ! loop over a month  
  DO ! loop over a day  
    CALL FORDATE()  
    CALL STLOOP() ! integrate for a day  
  END DO  
END DO
```

# The Bug (2)

## N-cycle (fixed)

```
PROGRAM agcm  
CALL iniall ()  
IDAY=0; CALL FORDATE()  
  
DO ! loop over a month  
  DO ! loop over a day  
    CALL FORDATE()  
    CALL STLOOP() ! integrate for a day  
  END DO  
END DO
```

## N-cycle (with bug)

```
PROGRAM agcm  
CALL iniall ()  
  
DO ! loop over a month  
  DO ! loop over a day  
    CALL FORDATE()  
    CALL STLOOP() ! integrate for a day  
  END DO  
END DO
```