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The leapfrog is dead.  
Long live the leapfrog!

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# Time-stepping methods

weather  
and climate  
models

Method	Order	Formula	Storage factor	Efficiency factor	Amplitude error	Phase error	Maximum $\omega\Delta t$
Forward (Adams-Bashforth)	1	$\phi^{n+1} = \phi^n + hF(\phi^n)$	2	0	$1 + \frac{p^2}{2}$	$1 - \frac{p^2}{3}$	0
Backward (Adams-Moulton)	1	$\phi^{n+1} = \phi^n + hF(\phi^{n+1})$	Implicit	$\infty$	$1 - \frac{p^2}{2}$	$1 - \frac{p^2}{3}$	$\infty$
Matsuno	1	$\phi^* = \phi^n + hF(\phi^n)$ $\phi^{n+1} = \phi^n + hF(\phi^*)$	2	.5	$1 - \frac{p^2}{2}$	$1 + \frac{2}{3}p^2$	1
Asselin-filtered leapfrog	1	$\phi^{n+1} = \phi^{n-1} + 2hF(\phi^n)$ $\phi^n = \phi^n + \gamma(\phi^{n-1} - 2\phi^n + \phi^{n+1})$	3	<1	$1 - \frac{\gamma}{2(1-\gamma)}p^2$	$1 + \frac{1+2\gamma}{6(1-\gamma)}p^2$	<1
Leapfrog	2	$\phi^{n+1} = \phi^{n-1} + 2hF(\phi^n)$	2	1	1	$1 + \frac{p^2}{6}$	1
Runge-Kutta (Williamson/Huen)	2	$q_1 = hF(\phi^n), \phi_1 = \phi^n + q_1$ $q_2 = hF(\phi_1) - q_1, \phi^{n+1} = \phi_1 + q_2/2$	2	0	$1 + \frac{p^4}{8}$	$1 + \frac{p^2}{6}$	0
Adams-Bashforth	2	$\phi^{n+1} = \phi^n + \frac{h}{2}[3F(\phi^n) - F(\phi^{n-1})]$	3	0	$1 + \frac{p^4}{4}$	$1 + \frac{5}{12}p^2$	0
Adams-Moulton (Trapizoidal)	2	$\phi^{n+1} = \phi^n + \frac{h}{2}[F(\phi^{n+1}) + F(\phi^n)]$	Implicit	$\infty$	1	$1 - \frac{p^2}{12}$	$\infty$
Leapfrog, then Adams-Bashforth (Magazenkov)	2	$\phi^n = \phi^{n-2} + 2hF(\phi^{n-1})$ $\phi^{n+1} = \phi^n + \frac{h}{2}[3F(\phi^n) - F(\phi^{n-1})]$	3	.67	$1 - \frac{p^4}{4}$	$1 + \frac{p^2}{6}$	.67
Leapfrog-Trapizoidal (Kurihara)	2	$\phi^* = \phi^{n-1} + 2hF(\phi^n)$ $\phi^{n+1} = \phi^n + \frac{h}{2}[F(\phi^n) + F(\phi^*)]$	3	.71	$1 - \frac{p^4}{4}$	$1 - \frac{p^2}{12}$	1.41
Young's method A	2	$\phi_1 = \phi^n + hF(\phi^n)/2$ $\phi_2 = \phi_1 + hF(\phi_1)/2$ $\phi^{n+1} = \phi^n + hF(\phi_2)$	3	0	$1 + \frac{p^6}{128}$	$1 + \frac{p^2}{24}$	0
Runge-Kutta (Williamson)	3	$q_1 = hF(\phi^n), \phi_1 = \phi^n + q_1/3$ $q_2 = hF(\phi_1) - 5q_1/9, \phi_2 = \phi^1 + 15q_2/16$ $q_3 = hF(\phi_2) - 153q_2/128, \phi^{n+1} = \phi_2 + 8q_3/15$	2	.58	$1 - \frac{p^4}{24}$	$1 + \frac{p^4}{30}$	1.73
ABM predictor- corrector	3	$\phi^* = \phi^n + \frac{h}{2}[3F(\phi^n) - F(\phi^{n-1})]$ $\phi^{n+1} = \phi^* + \frac{5h}{12}[F(\phi^*) - 2F(\phi^n) + F(\phi^{n-1})]$	4	.60	$1 - \frac{19}{144}p^4$	$1 + \frac{1243}{8640}p^4$	1.20
Adams-Moulton	3	$\phi^{n+1} = \phi^n + \frac{h}{12}[5F(\phi^{n+1}) + 8F(\phi^n) - F(\phi^{n-1})]$	Implicit	0	$1 + \frac{p^4}{24}$	$1 - \frac{11}{720}p^4$	0
Adams-Bashforth	3	$\phi^{n+1} = \phi^n + \frac{h}{12}[23F(\phi^n) - 16F(\phi^{n-1}) + 5F(\phi^{n-2})]$	4	.72	$1 - \frac{3}{8}p^4$	$1 + \frac{289}{720}p^4$	0.72
Runge-Kutta (Classical)	4	$k_0 = hF(\phi^n)$ $k_1 = hF(\phi^n + k_0/2)$ $k_2 = hF(\phi^n + k_1/2)$ $k_3 = hF(\phi^n + k_3)$ $\phi^{n+1} = \phi^n + \frac{1}{6}(k_0 + 2k_1 + 2k_2 + k_3)$	3*	.70	$1 - \frac{p^6}{144}$	$1 - \frac{p^4}{120}$	2.82
ABM predictor- corrector	4	$\phi^* = \phi^n + \frac{h}{12}[23F(\phi^n) - 16F(\phi^{n-1}) + 5F(\phi^{n-2})]$ $\phi^{n+1} = \phi^* + \frac{3h}{8}[F(\phi^*) - 3F(\phi^n) + 3F(\phi^{n-1}) - F(\phi^{n-2})]$	5	.59	$1 - \frac{265}{1536}p^6$	$1 - \frac{329}{2880}p^4$	1.18
Adams-Moulton	4	$\phi^{n+1} = \phi^n + \frac{h}{24}[9F(\phi^{n+1}) + 19F(\phi^n) - 5F(\phi^{n-1}) + F(\phi^{n-2})]$	Implicit	0	$1 + \frac{p^6}{48}$	$1 + \frac{19}{720}p^4$	0
Adams-Bashforth	4	$\phi^{n+1} = \phi^n + \frac{h}{24}[55F(\phi^n) - 59F(\phi^{n-1}) + 37F(\phi^{n-2}) - 9F(\phi^{n-3})]$	5	.43	$1 - \frac{13}{24}p^6$	$1 - \frac{251}{720}p^4$	.43

increasing accuracy ↓

computational  
fluid  
dynamics

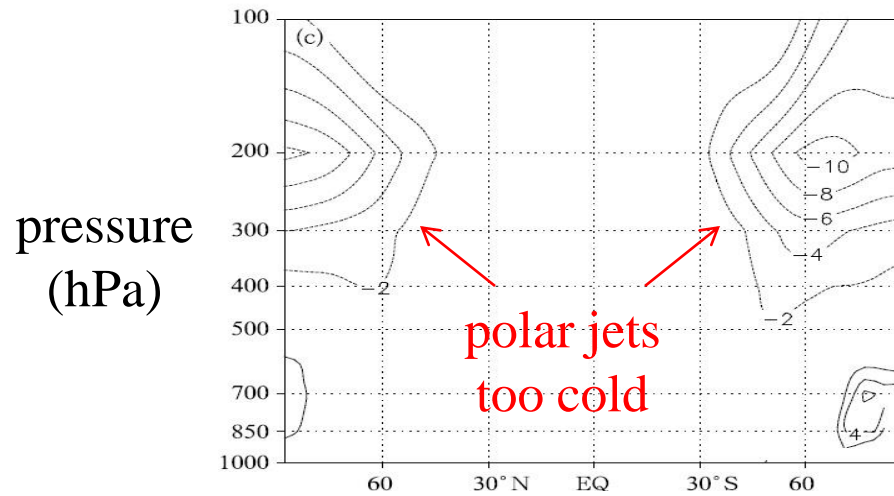
(Durran 1991)

# Impact of time-stepping schemes

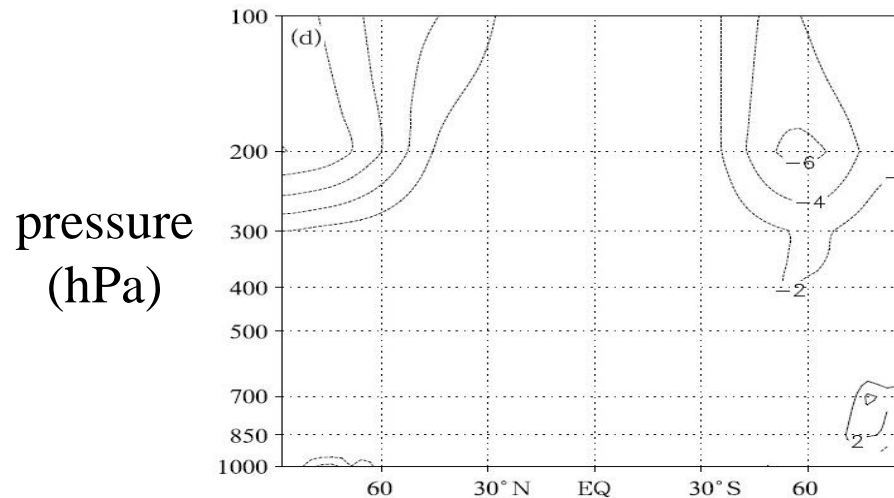
- “In the weather and climate prediction community, when thinking in terms of model predictability, there is a tendency to associate model error with the physical parameterizations. In this paper, it is shown that **time truncation error can be a substantial part of the total forecast error.**” (Teixeira et al. 2007)
- “Climate simulations are sensitive not only to physical parameterizations of subgrid-scale processes **but also to the numerical methodology employed.**” (Pfeffer et al. 1992)
- “Many published conclusions on parameter sensitivity, calibrated values and associated uncertainty may be questionable due to **numerical artifacts introduced by unreliable time stepping schemes.**” (Kavetski & Clark 2010)

# Impact of time stepping in weather and climate prediction

annual-mean zonal-mean temperature error ( C) in CAM relative to ERA40



The default scheme  
(leapfrog + RA filter)



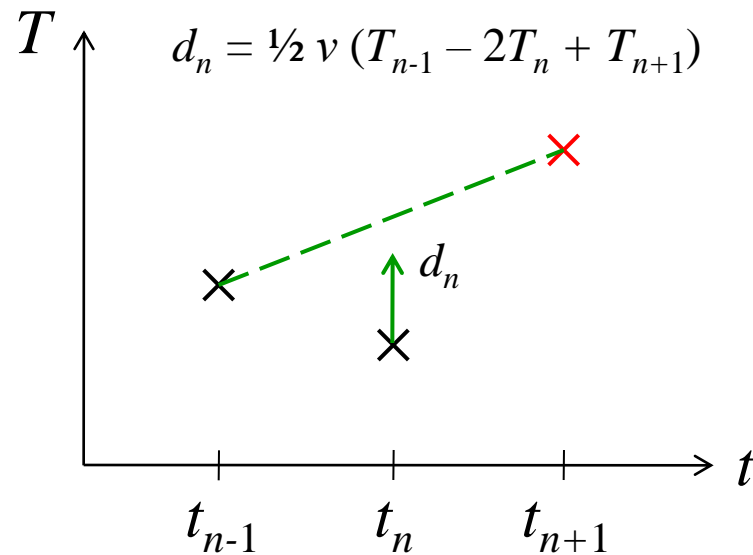
A more accurate scheme  
(AB2)

(Zhao & Zhong 2009)

# Leapfrog with Robert–Asselin filter

LF+RA

(Robert 1966, Asselin 1972)



- use leapfrog to calculate  $T_{n+1}$
- RA filter nudges  $T_n$
- reduces curvature but does not conserve mean
- amplitude accuracy is 1st order

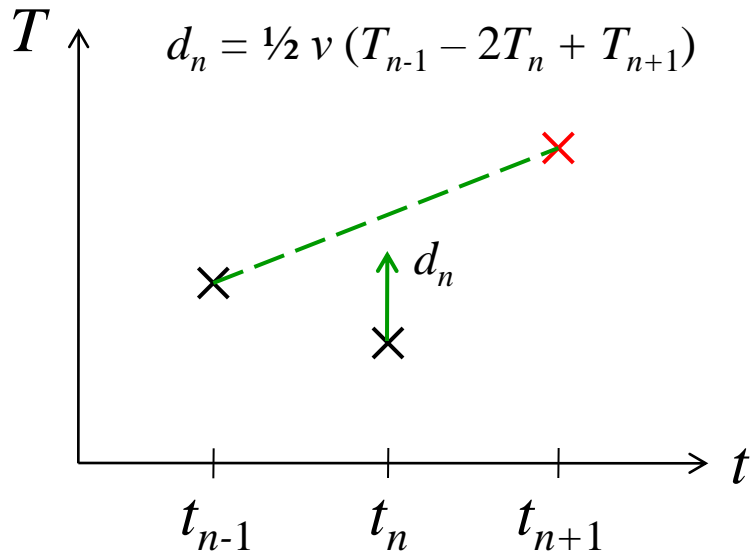
# Leapfrog with Robert–Asselin filter

- Widely used in current numerical models
  - **atmosphere**: ECHAM, MAECHAM, MM5, CAM, MESO-NH, HIRLAM, KMCM, LIMA, SPEEDY, IGCM, PUMA, COSMO, FSU-GSM, FSU-NRSM, NCEP-GFS, NCEP-RSM, NSEAM, NOGAPS, RAMS, CCSR/NIES-AGCM
  - **ocean**: OPA, ORCA, NEMO, HadOM3, DieCAST, TIMCOM, GFDL-MOM, POM, MICOM, HYCOM, POSEIDON, NCOM, ICON, OFES, SOM
  - **coupled**: HiGEM (oce), COAMPS (atm), PlaSim (atm), ECHO (atm), MIROC (atm), FOAM (oce), NCAR-CCSM (atm), BCM (oce), NCEP-CFS (atm/oce), QESM (oce), CHIME (oce), FORTE (atm)
  - **others**: GTM, ADCIRC, QUAGMIRE, MORALS, SAM, ARPS, CASL, CReSS, JTGCM, ECOMSED, UKMO-LEM, MPI-REMO
- Asselin (1972) has received over 450 citations
- Has many problems
  - “*The Robert–Asselin filter has proved immensely popular, and has been widely used for over 20 years. However, it is not the last word...*” (Lynch 1991)
  - “*Replacement of the Asselin time filter... can be a feasible way to improve the ability of climate models*” (Zhao & Zhong 2009)
  - “*The Robert–Asselin filter can produce slewing frequency as well as the well-known damping and phase errors*” (Thrustarson & Cho 2011)

# A proposed improvement

## LF+RA

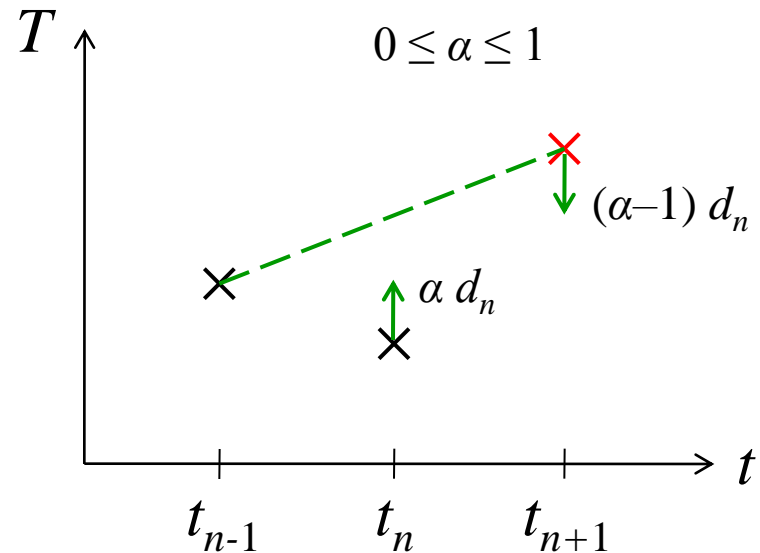
(Robert 1966, Asselin 1972)



- use leapfrog to calculate  $T_{n+1}$
- RA filter nudges  $T_n$
- reduces curvature but does not conserve mean
- amplitude accuracy is 1st order

## LF+RAW

(Williams 2009, 2011)



- use leapfrog to calculate  $T_{n+1}$
- RAW filter nudges  $T_n$  and  $T_{n+1}$
- reduces curvature and conserves mean (for  $\alpha=1/2$ )
- amplitude accuracy is 3rd order

# Simple test integration

$$\begin{aligned}\frac{dX}{dt} &= -\omega Y \\ \frac{dY}{dt} &= +\omega X\end{aligned}$$

$$\omega = 1 \text{ rad s}^{-1}$$

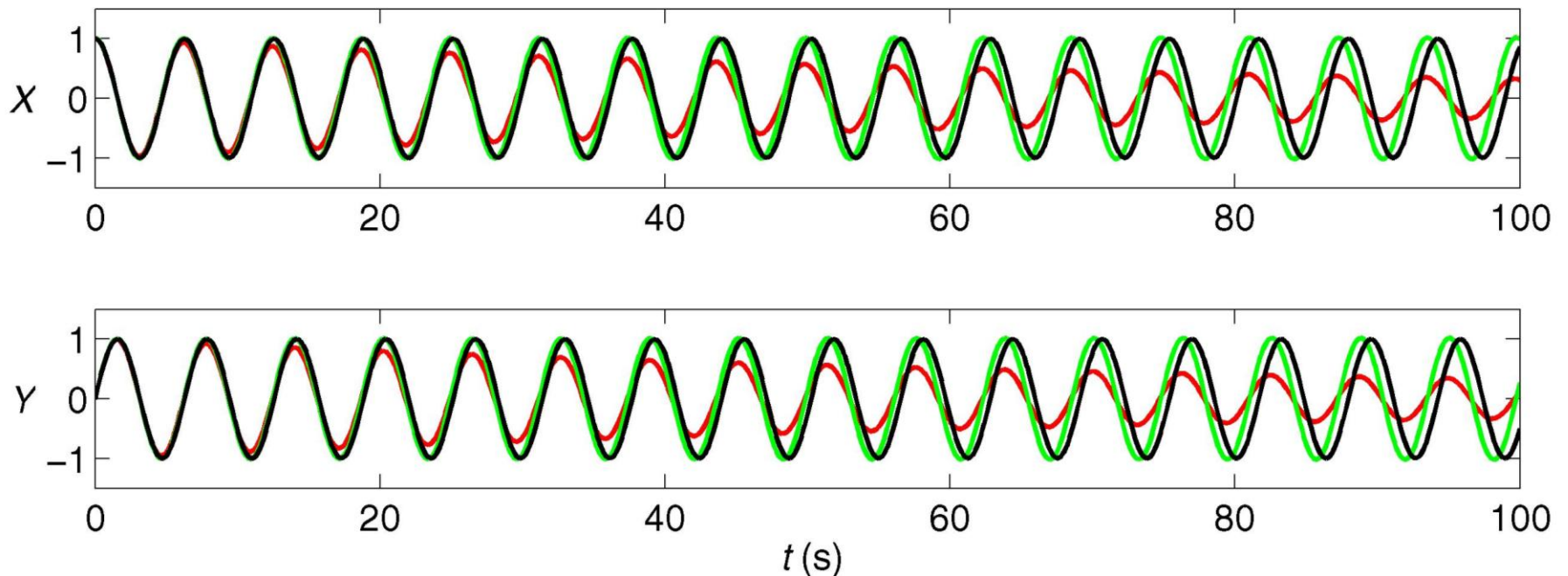
**exact**

**LF+RA**

**LF+RAW<sub>α=1/2</sub>**

$$\Delta t = 0.2 \text{ s}$$

$$\nu = 0.2$$

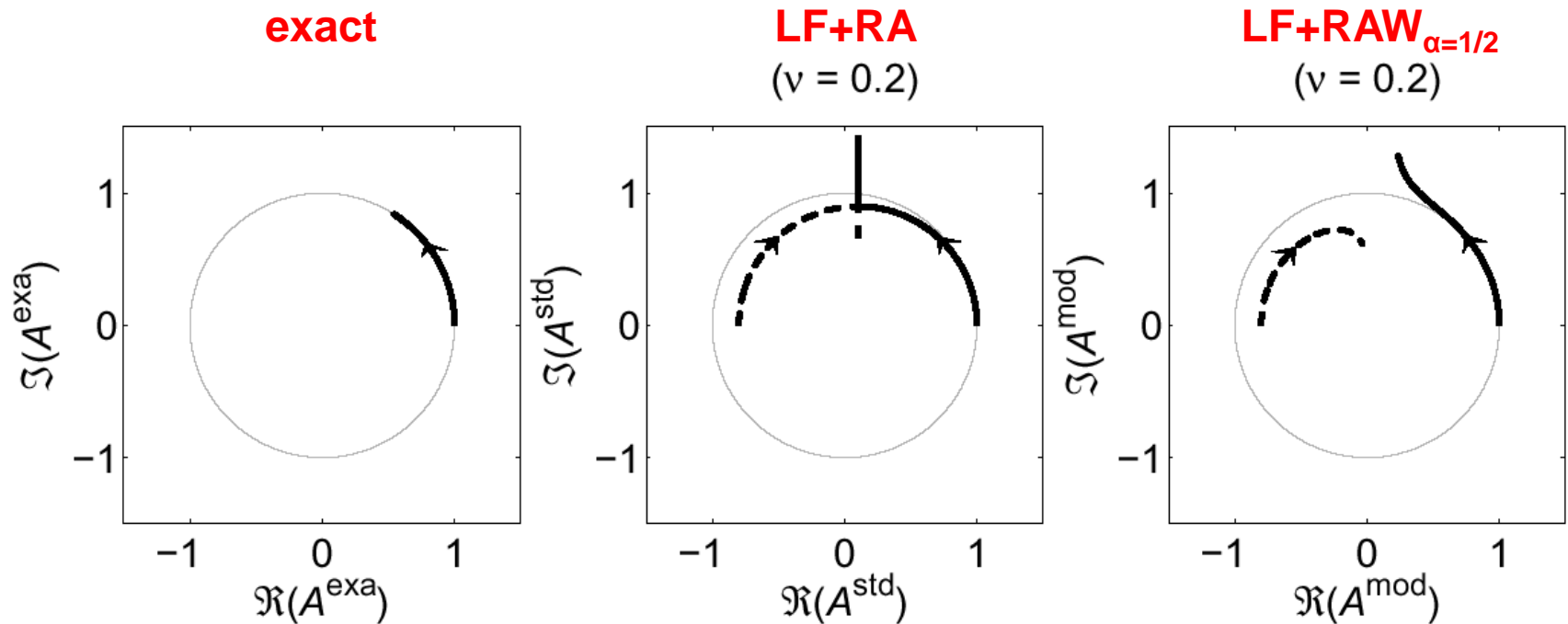


(Williams 2009)



# Analysis: numerical amplification

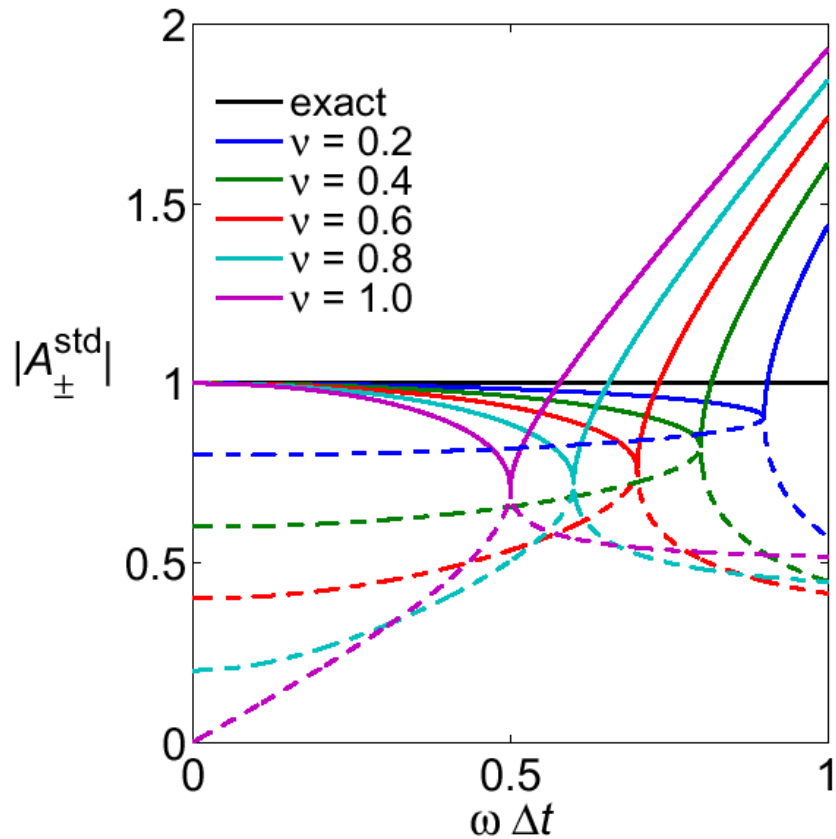
Let  $\dot{F} = i\omega F$  and  $A = F(t+\Delta t) / F(t)$  and trace  $A$  as  $\omega\Delta t = 0 \rightarrow 1$ :



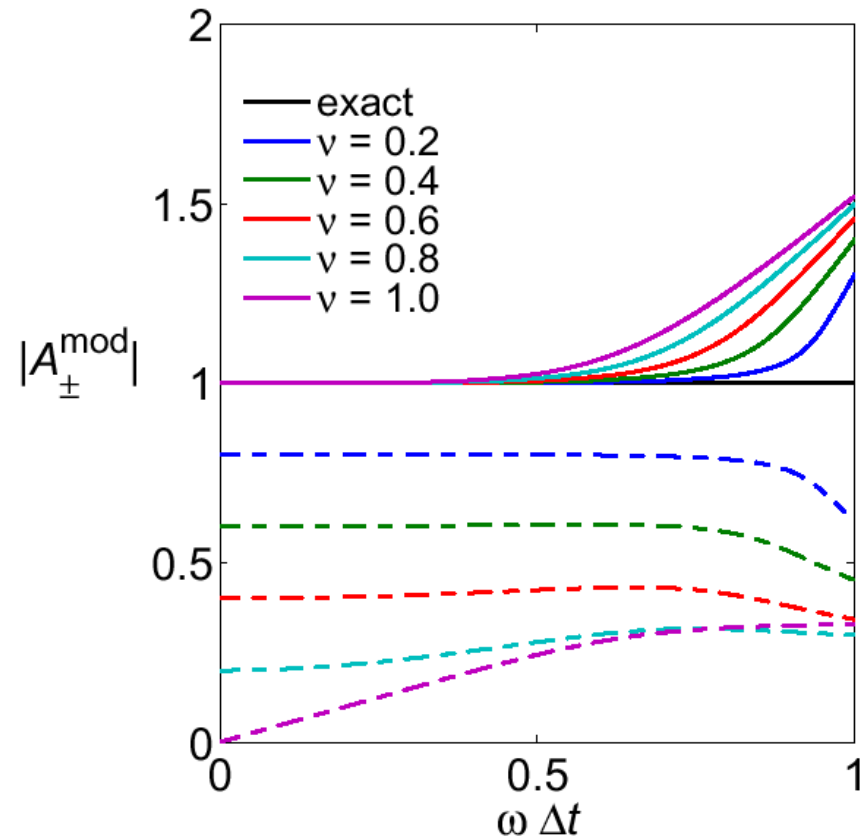
(Williams 2009)

# Analysis: numerical amplification

LF+RA

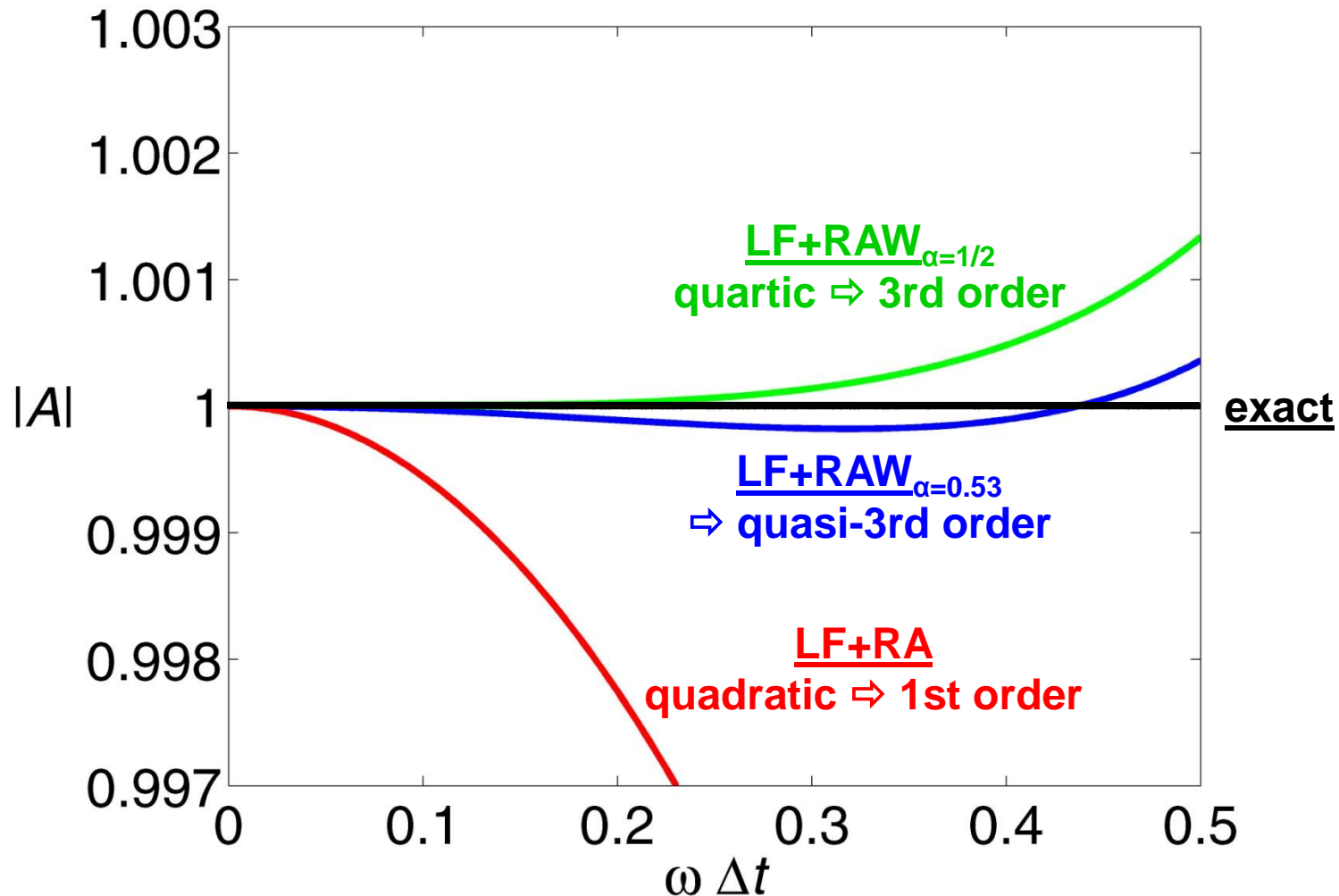


LF+RAW <sub>$\alpha=1/2$</sub>



# Analysis: numerical amplification

$$\nu = 0.2$$



(Williams 2009)

# Implementation in existing code

! Compute tendency at this time step

tendency = [...]

! Leapfrog step

$x_{\text{next}} = x_{\text{last}} + \text{tendency} * 2 * \text{delta}_t$

! Compute filter displacement

$d = \text{nu} * (x_{\text{last}} - 2 * x_{\text{this}} + x_{\text{next}}) / 2$

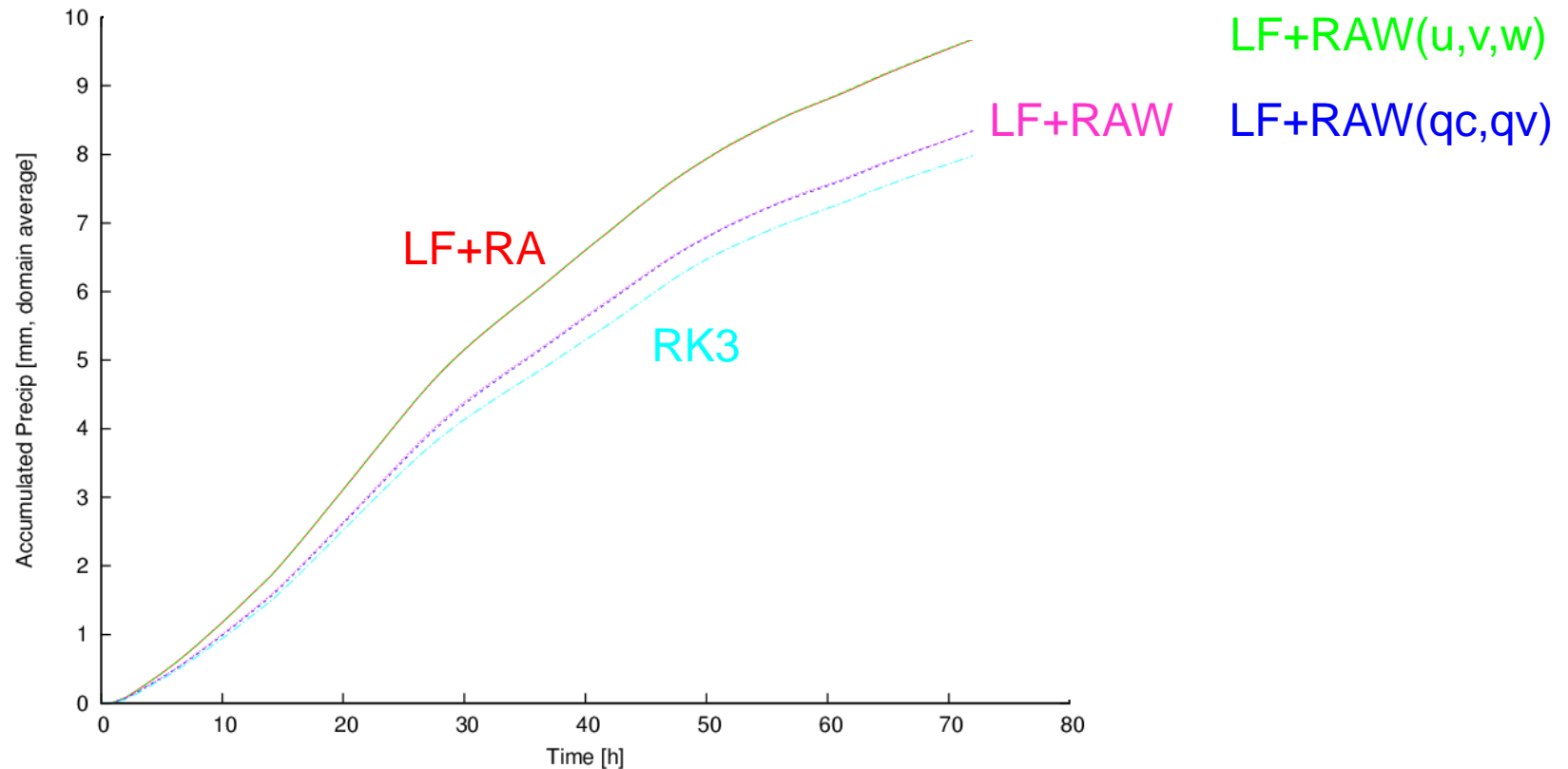
! Apply filter

$x_{\text{this}} = x_{\text{this}} + d * \text{alpha}$

$x_{\text{next}} = x_{\text{next}} + d * (\text{alpha} - 1)$

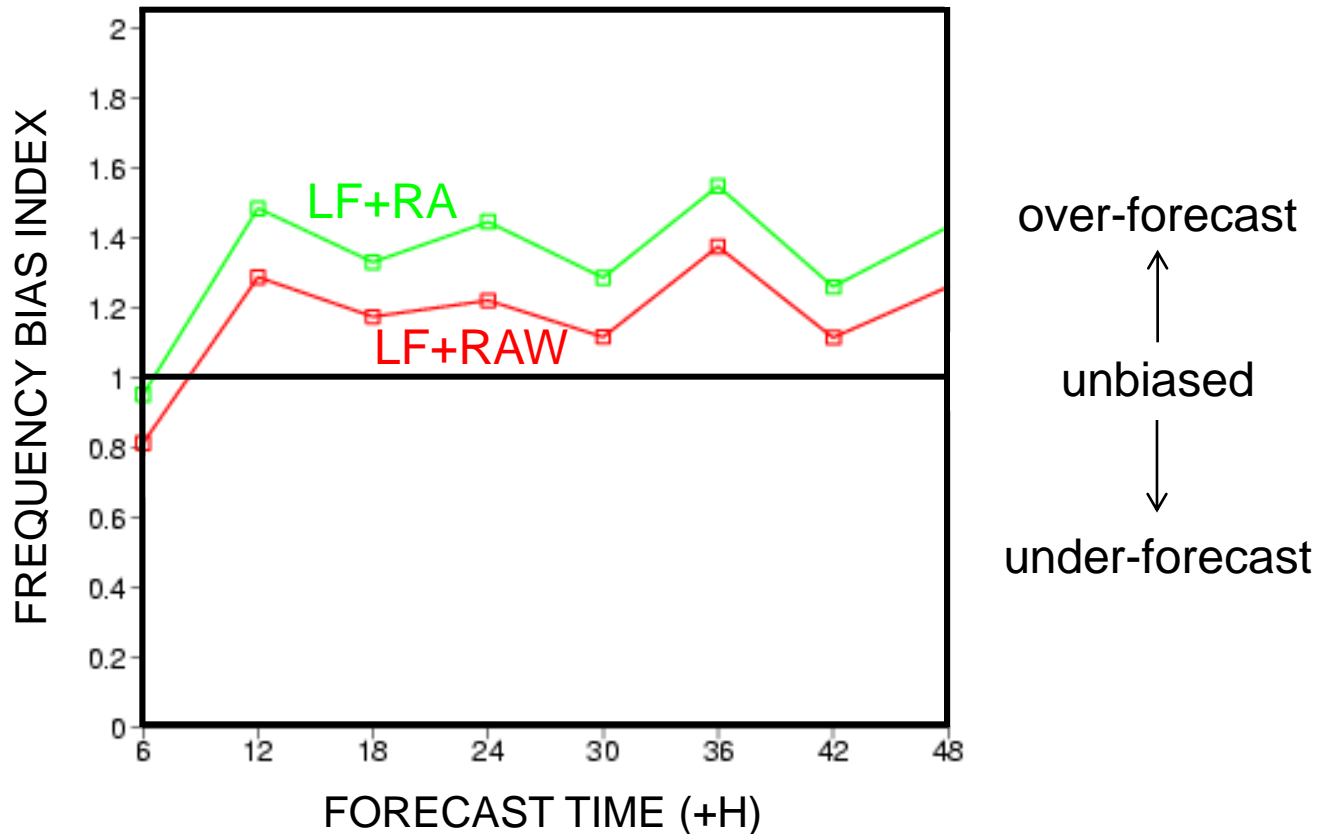
# Implementation in COSMO-7

- Test case: 72h forecast over Europe starting at 00Z on 9 Feb 2009
- LF+RAW is almost as accurate as the (more expensive) RK3 integration



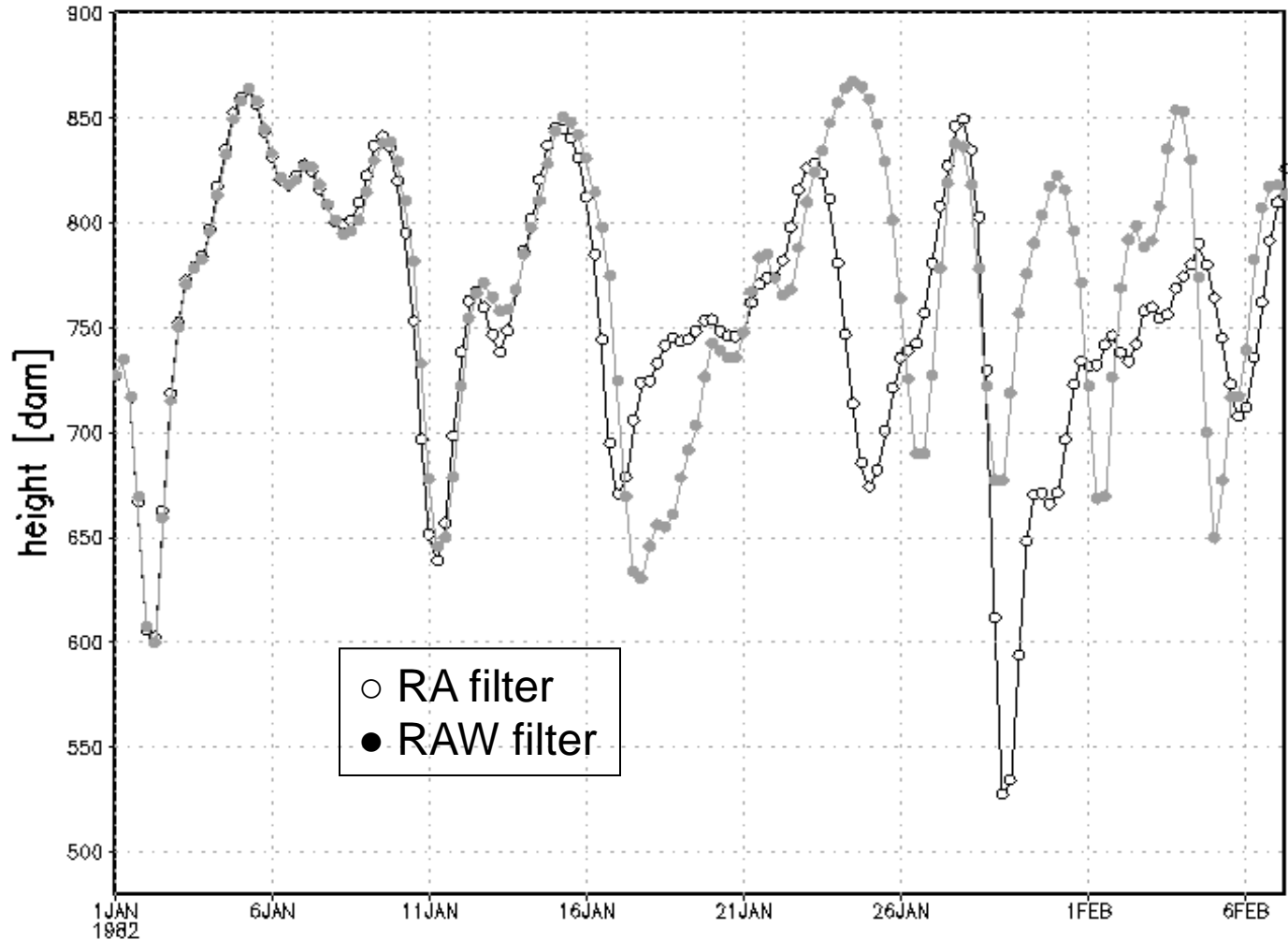
# Implementation in COSMO-ME

- Statistical analysis: two 48h forecasts per day from 16 Dec 2008 to 18 Jan 2009
- The modified filter significantly improves precipitation forecasts (2-10 mm / 6h)



# Implementation in SPEEDY

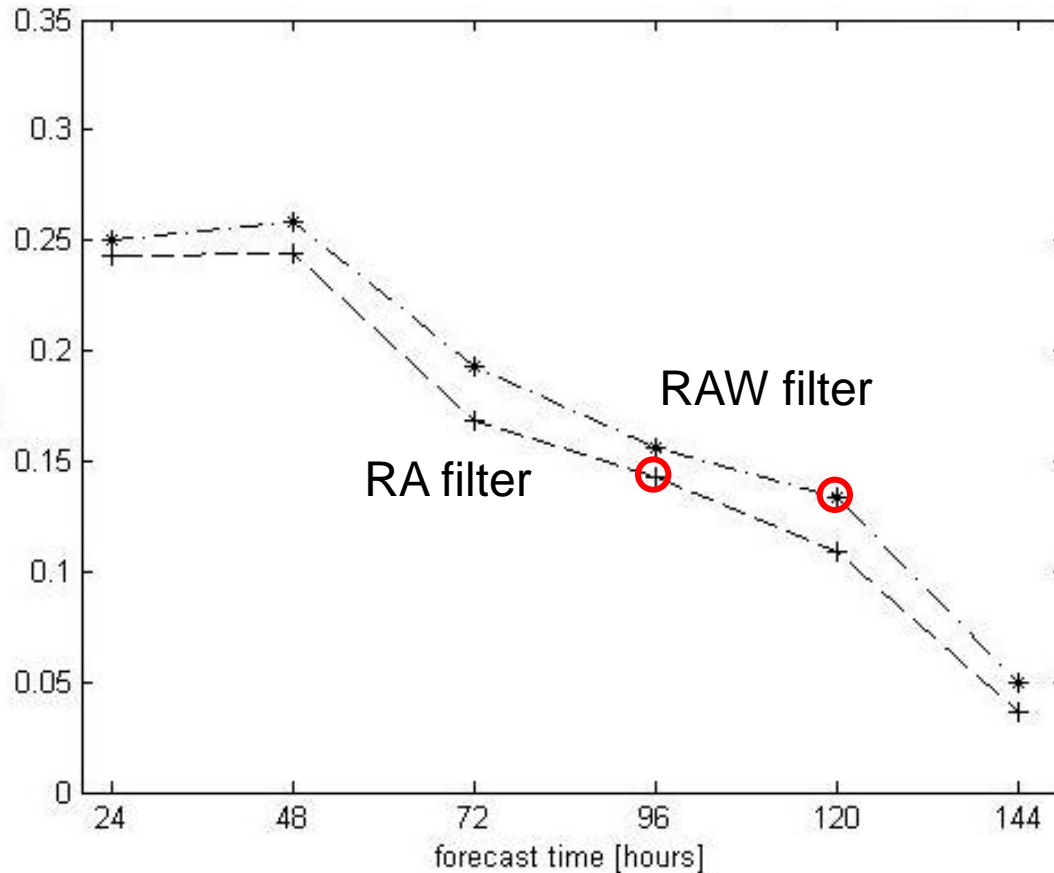
500 hPa  
geopotential  
height in  
Maryland



(Amezcuca, Kalnay & Williams 2011)

# Implementation in SPEEDY

ACC for  
surface  
pressure in  
the tropics  
(25°S-25°N)

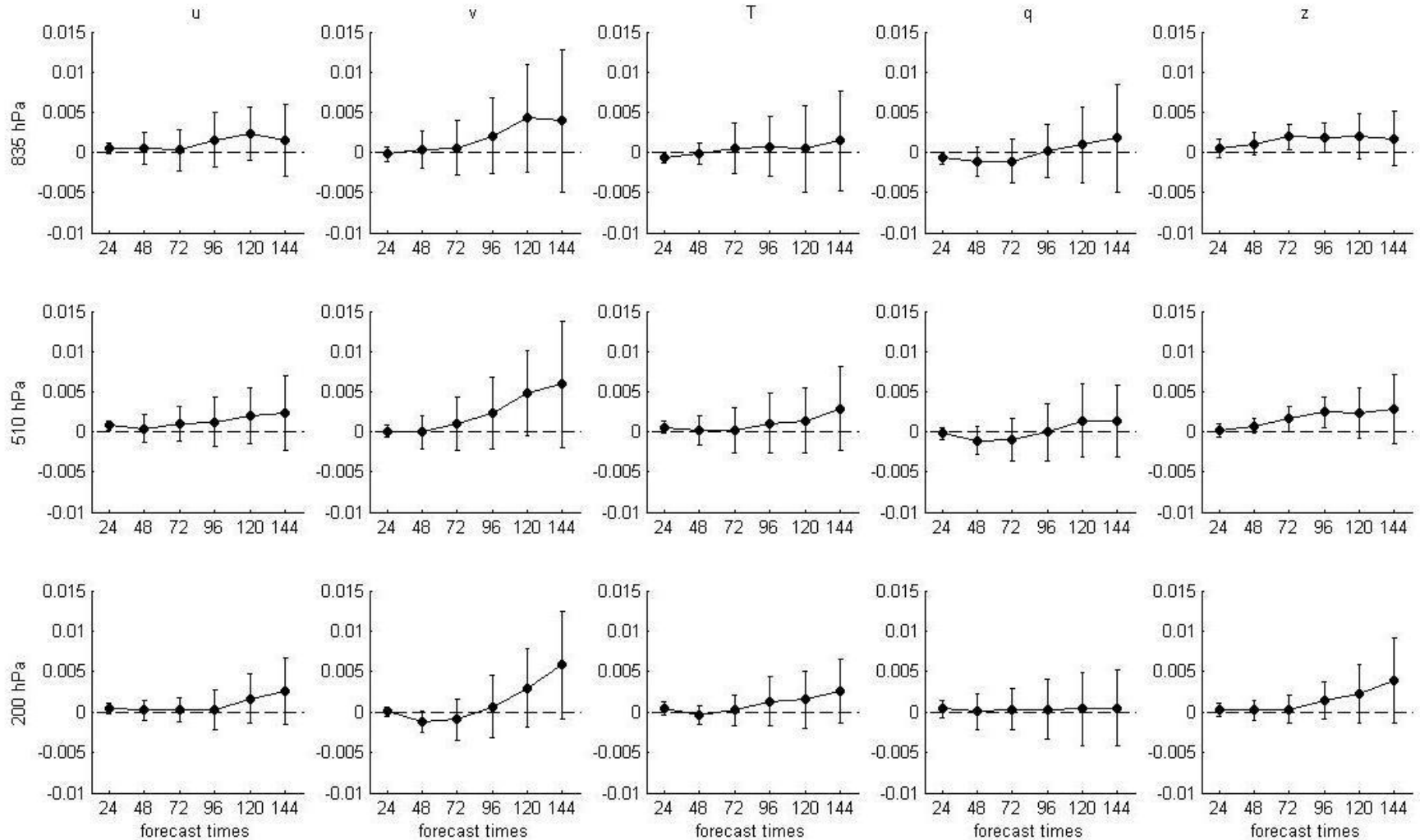


5-day forecasts  
made using the  
RAW filter have  
approximately  
the same skill as  
4-day forecasts  
made using the  
RA filter



# Implementation in SPEEDY

$ACC_{RAW} - ACC_{RA}$  for the 5 variables, computed globally and separated per pressure level



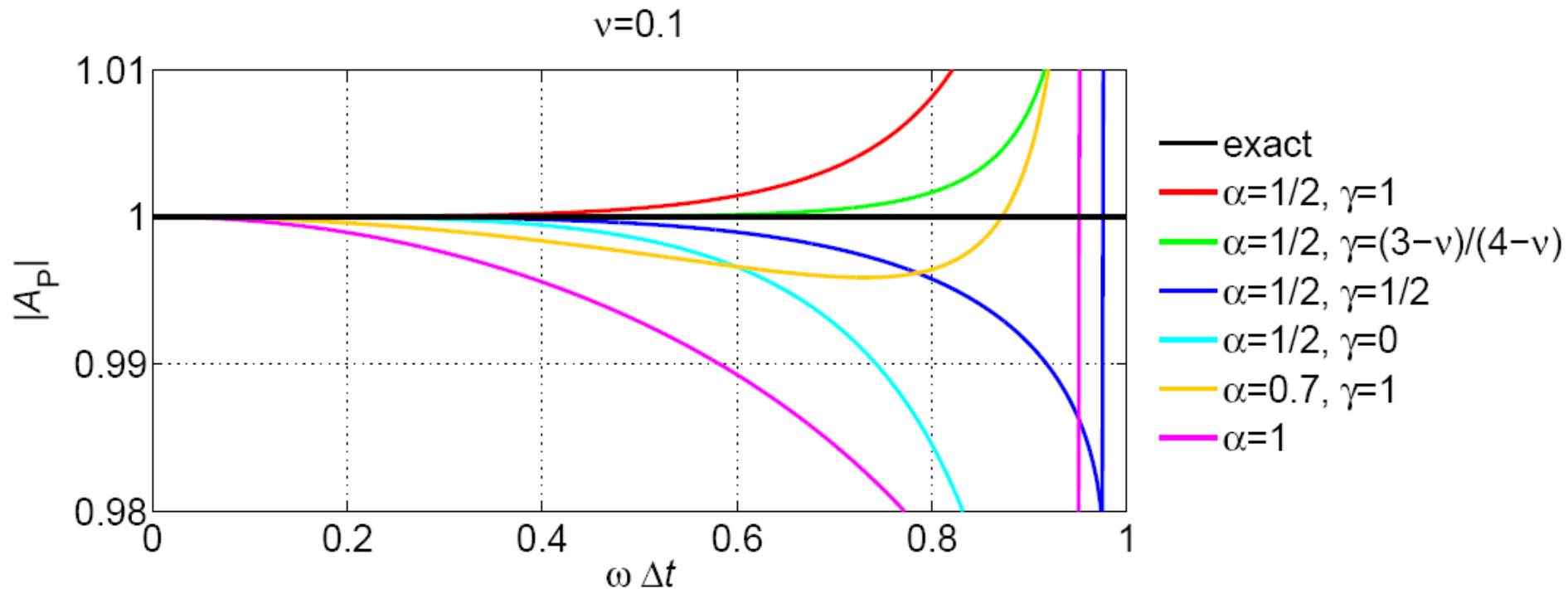
(Amezcuca, Kalnay & Williams 2011)

# Composite-tendency leapfrog with (1, -2, 1) filter

$$\frac{dx}{dt} = f(x) \left\{ \begin{array}{l} x_{n+1} = \bar{\bar{x}}_{n-1} + 2\Delta t [\gamma f(\bar{x}_n) + (1 - \gamma)f(x_n)] \\ \bar{\bar{x}}_n = \bar{x}_n + \frac{\nu\alpha}{2} [\bar{\bar{x}}_{n-1} - 2\bar{x}_n + x_{n+1}] \\ \bar{x}_{n+1} = x_{n+1} - \frac{\nu(1 - \alpha)}{2} [\bar{\bar{x}}_{n-1} - 2\bar{x}_n + x_{n+1}] \end{array} \right. \begin{array}{l} \text{leapfrog} \\ \text{filter} \end{array}$$

$$\frac{dx}{dt} = i\omega x \left\{ \begin{array}{ll} |A_P| - 1 = \frac{\nu(1 - 2\alpha)}{2(2 - \nu)} (\omega\Delta t)^2 + \mathcal{O}[(\omega\Delta t)^4] & \text{1st order} \\ \alpha = \frac{1}{2} \quad |A_P| - 1 = \frac{\nu(4\gamma - 3 + \nu - \nu\gamma)}{4(2 - \nu)^2} (\omega\Delta t)^4 + \mathcal{O}[(\omega\Delta t)^6] & \text{3rd order} \\ \gamma = \frac{3 - \nu}{4 - \nu} \quad |A_P| - 1 = \frac{\nu}{4(4 - \nu)(2 - \nu)^2} (\omega\Delta t)^6 + \mathcal{O}[(\omega\Delta t)^8] & \text{5th order} \end{array} \right.$$

# Composite-tendency leapfrog with $(1, -2, 1)$ filter

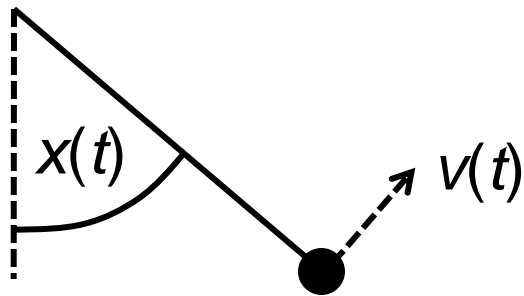


# Composite-tendency leapfrog with (1, -4, 6, -4, 1) filter

$$\frac{dx}{dt} = f(x) \left\{ \begin{array}{l} x_{n+1} = \bar{\bar{x}}_{n-1} + 2\Delta t [\gamma f(\bar{x}_n) + (1 - \gamma)f(x_n)] \\ \bar{\bar{x}}_n = \bar{x}_n + \nu\alpha [\bar{x}_{n-3} - 4\bar{x}_{n-2} + 6\bar{x}_{n-1} - 4\bar{x}_n + x_{n+1}] \\ \bar{x}_{n+1} = x_{n+1} - \nu(1 - \alpha) [\bar{x}_{n-3} - 4\bar{x}_{n-2} + 6\bar{x}_{n-1} - 4\bar{x}_n + x_{n+1}] \end{array} \right. \begin{array}{l} \text{leapfrog} \\ \text{filter} \end{array}$$

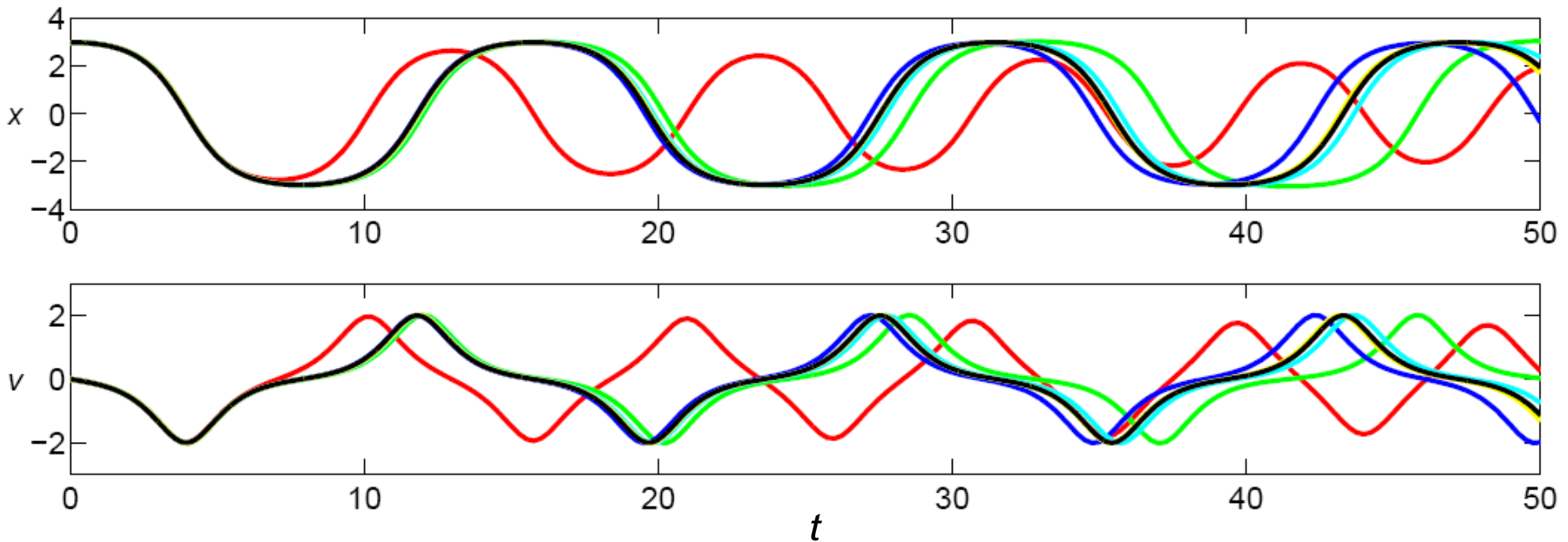
$$\frac{dx}{dt} = i\omega x \left\{ \begin{array}{l} |A_P| - 1 = -\frac{\nu(1 - 2\alpha)}{2(1 - \nu - 2\alpha\nu)}(\omega\Delta t)^4 + \mathcal{O}[(\omega\Delta t)^6] \quad \text{3rd order} \\ \alpha = \frac{1}{2} \quad |A_P| - 1 = \frac{\nu(5 - 8\gamma - 9\nu + 14\nu\gamma)}{8(1 - 2\nu)^2}(\omega\Delta t)^6 + \mathcal{O}[(\omega\Delta t)^8] \quad \text{5th order} \\ \gamma = \frac{5 - 9\nu}{2(4 - 7\nu)} \quad |A_P| - 1 = -\frac{5\nu(4 - 13\nu + 11\nu^2)}{32(1 - 2\nu)^2(4 - 7\nu)}(\omega\Delta t)^8 + \mathcal{O}[(\omega\Delta t)^{10}] \quad \text{7th order} \end{array} \right.$$

# Nonlinear simple pendulum



- RK5(4)
- (1,-2,1),  $\alpha=1$
- (1,-2,1),  $\alpha=0.5, \gamma=1$
- (1,-2,1),  $\alpha=0.5, \gamma=0.5$
- (1,-2,1),  $\alpha=0.5, \gamma=0.74026$
- (1,-4,6,-4,1),  $\alpha=0.5, \gamma=0.61864$

$x(0) = 0.95\pi, v(0) = 0, \Delta t = 0.25, \nu = 0.15$



# Summary table

filter	$\alpha$	$\gamma$	no. computational modes	no. function evaluations per time step	order	amplitude order of accuracy	phase order of accuracy	maximum $\omega\Delta t$ for stability ( $\nu = 0.1$ )	maximum $\omega\Delta t$ for accuracy ( $\nu = 0.1$ )
(1, -2, 1)	1	-	1	1	1	1	2	0.951	0.030
	1/2	1	1	1	2	(3)	2	0	(0.448)
	1/2	$(3 - \nu)/(4 - \nu)$	2	2	2	(5)	2	0	(0.712)
	1/2	1/2	2	2	2	3	2	0.975	0.475
	1/2	0	2	1	2	3	2	0.832	0.333
	0	1	1	1	1	(1)	2	0	(0.030)
(1, -4, 6, -4, 1)	1	-	3	1	2	(3)	2	0	(0.228)
	1/2	1	3	1	2	5	2	0.690	0.424
	1/2	$(5 - 9\nu)/[2(4 - 7\nu)]$	4	2	2	7	2	0.616	0.586
	1/2	0	4	1	2	(5)	2	0	(0.394)
	0	1	3	1	2	3	2	0.777	0.240

# Summary

- Time stepping is an **important contributor** to model error
- The **Robert–Asselin filter** is widely used but is dissipative and reduces accuracy
- The **RAW filter** has approximately the same stability but much greater accuracy
- Implementation in an existing code is **trivial** and there is no extra computational cost
- **5th-order** and even **7th-order** amplitude accuracy may be achieved, by using a composite tendency and/or a more discriminating filter

# Further information

Williams PD. Achieving seventh-order amplitude accuracy in leapfrog integrations. *Monthly Weather Review*, in preparation.

Williams PD (2011) The RAW filter: An improvement to the Robert–Asselin filter in semi-implicit integrations. *Monthly Weather Review* **139**(6), 1996-2007.

Williams PD (2009) A proposed modification to the Robert–Asselin time filter. *Monthly Weather Review* **137**(8), 2538-2546.

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