

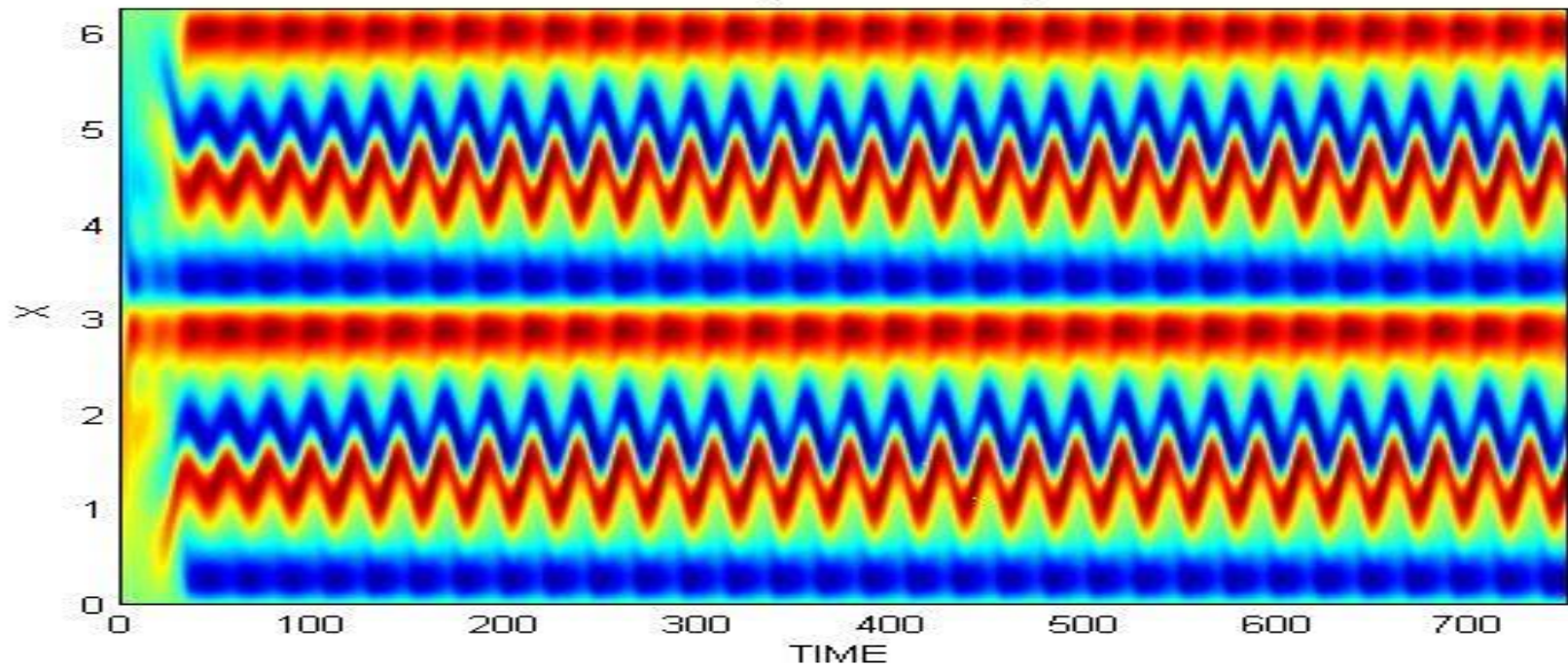


UNDERGRADUATE RESEARCH SCHOLAR

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Under the Direction of Dr. Lee Panetta

U EVOLUTION, $L^* = 5.4978$, $N = 128$



$$\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} + \frac{\partial^2 u}{\partial x^2} + \frac{\partial^4 u}{\partial x^4} = 0$$

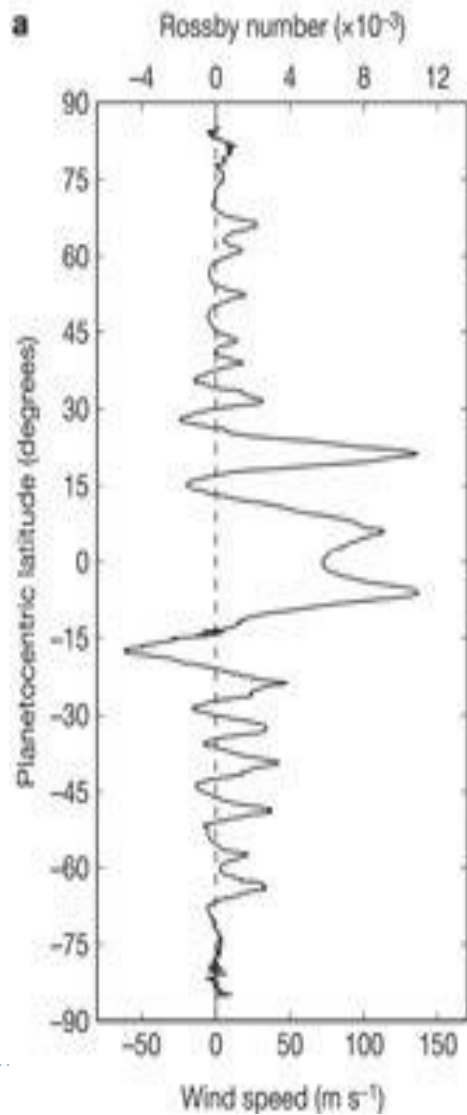
Kuramoto-Sivashinsky Equation

The Question

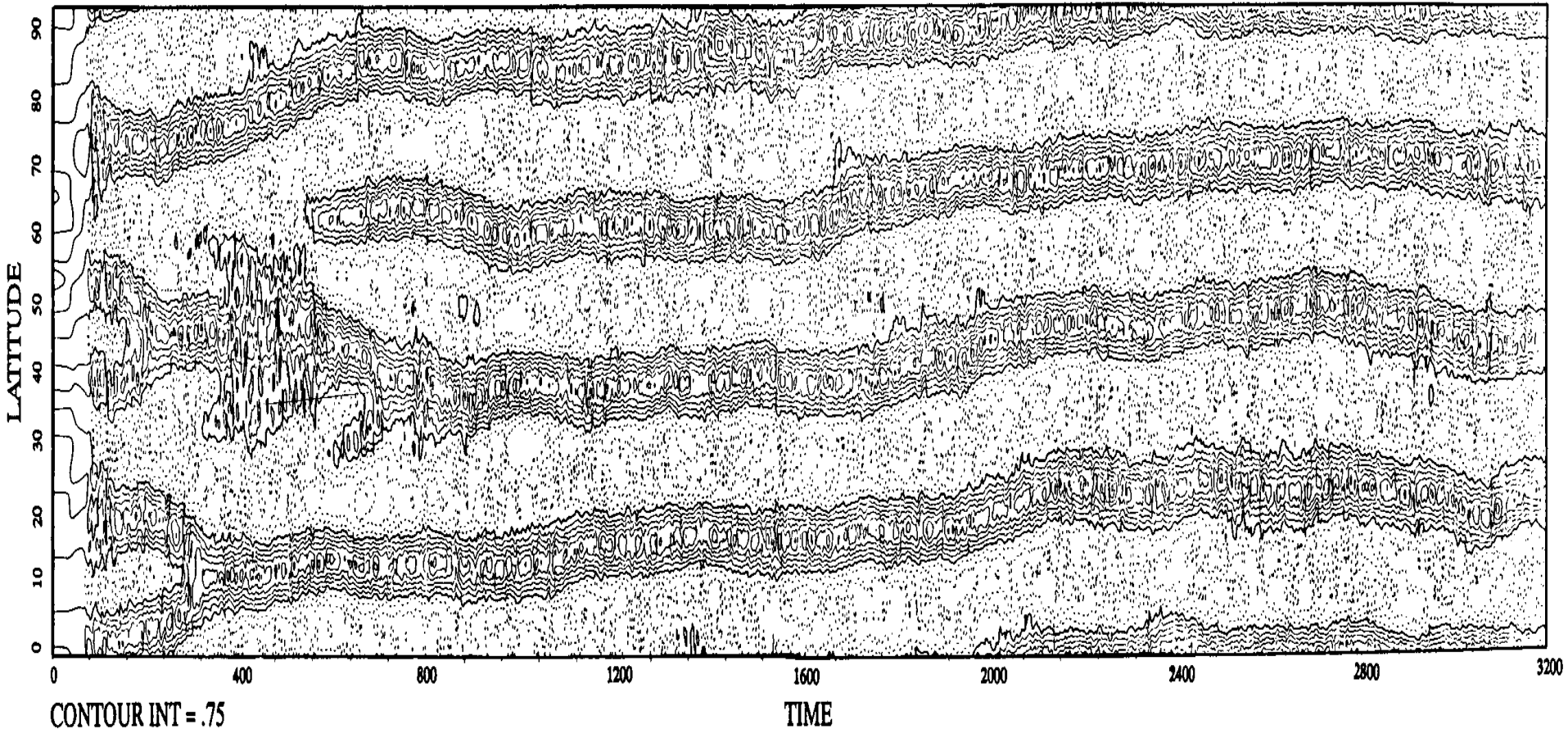
- ▶ Are numerical solutions to pattern-forming partial differential equations sensitive to time stepping methods?
- ▶ The Kuramoto-Sivashinsky Equation is a good model equation to study for this question



Has analogs in atmospheric science

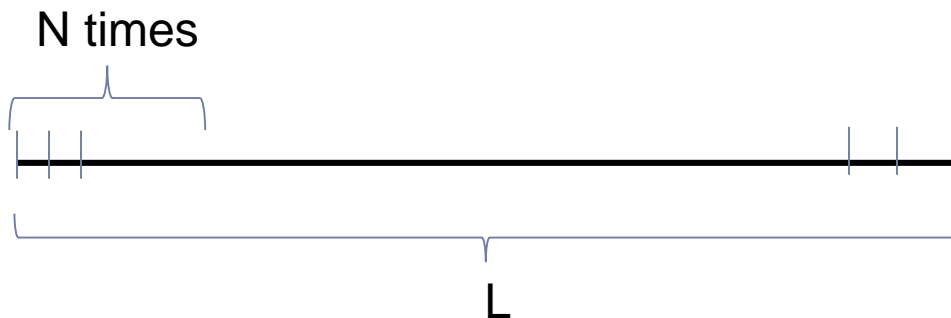


Has analogs in atmospheric science



What makes it so interesting?

- ▶ Assume solution is periodic on interval L (common assumption in atmospheric models)
- ▶ Given initial condition $u(x, 0) = u_0(x)$
- ▶ Divide L in to N parts (so N is the spatial resolution)

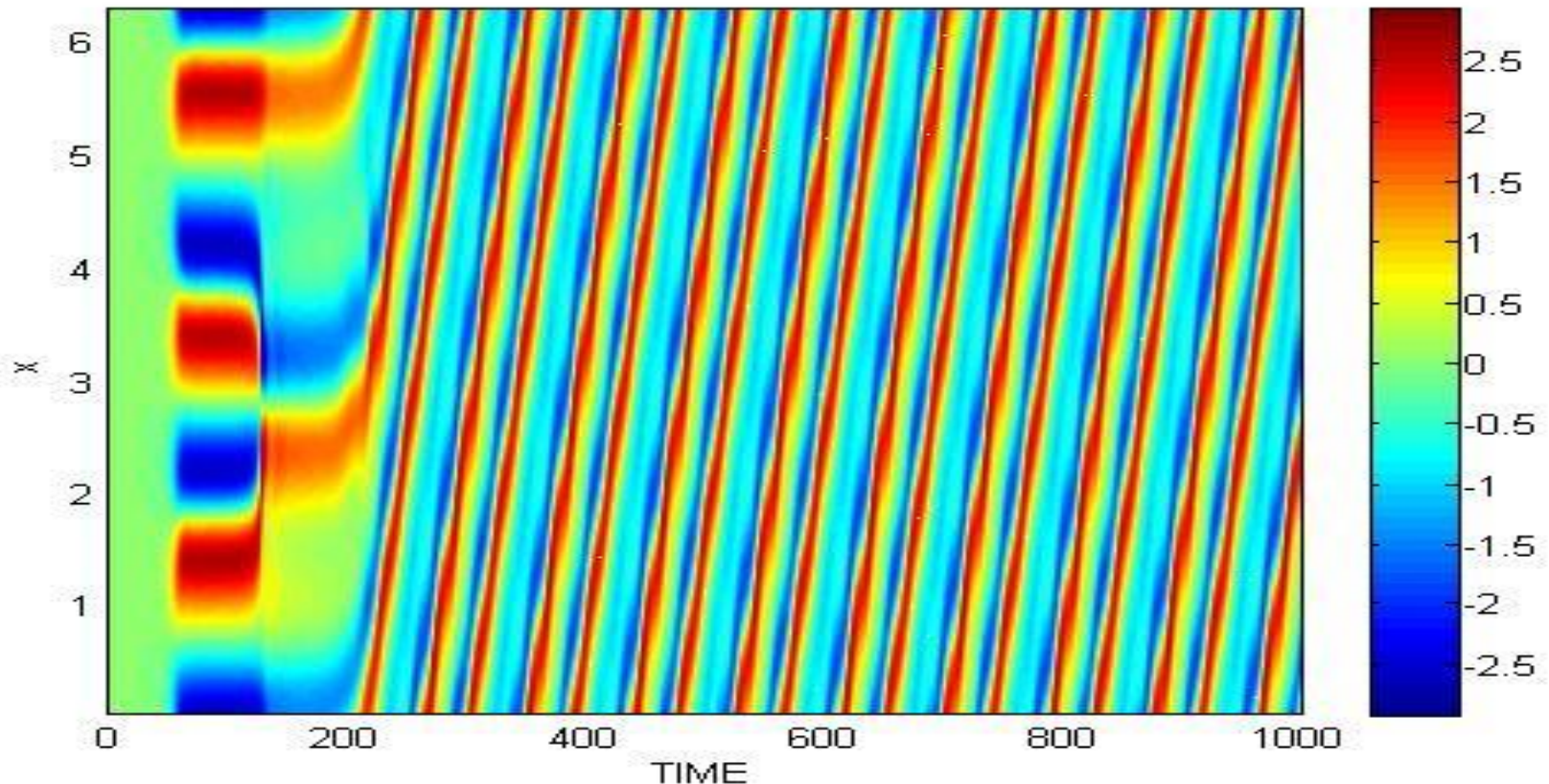


- ▶ For large values of L ($> 12\pi$), equation produces chaotic solutions
- ▶ For smaller values of L solutions have a wide array of structure
- ▶ Define $\hat{L} = \frac{L}{2\pi}$



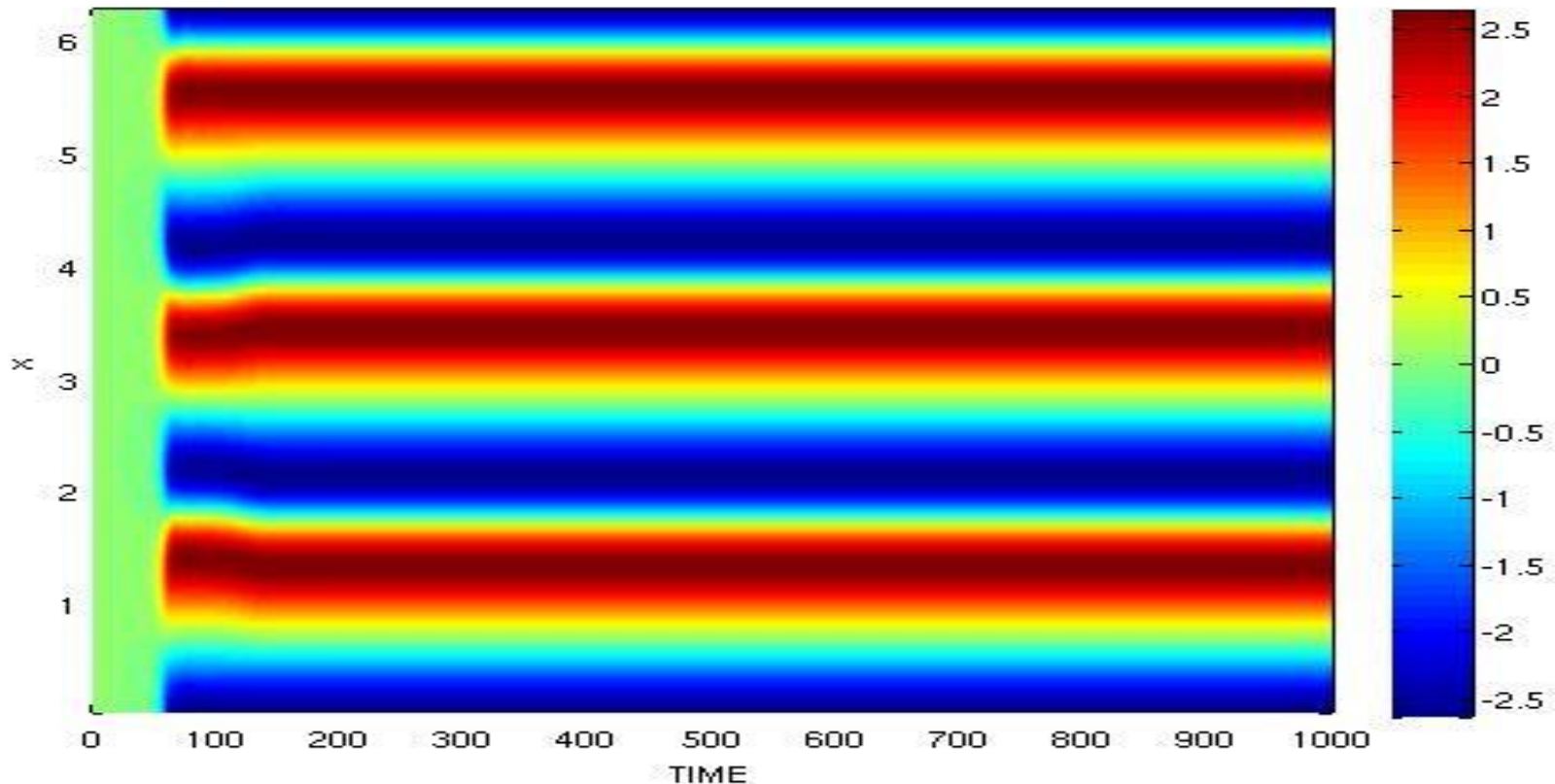
Solutions are very sensitive to \hat{L}

- ▶ Initial condition a randomized wave with small (10^{-5}) amplitude, $\hat{L} = 3.6398, N = 128$



Solutions are very sensitive to \hat{L}

- ▶ Increase \hat{L} by 0.0001 with exact same initial condition



Numerical Methods

- ▶ Separate the spatial (x) and temporal (t) derivatives so it looks like

$$\frac{\partial u}{\partial t} = -u \frac{\partial u}{\partial x} - \frac{\partial^2 u}{\partial x^2} - \frac{\partial^4 u}{\partial x^4}$$

- ▶ Use “pseudo-spectral” method for the spatial derivatives
- ▶ Time stepping method for time derivative



Solving the spatial derivatives

- ▶ If $u(x, t_n)$ is known, then we can use the Discrete Fourier Transform to approximate $u(x)$ as,

$$u(x, t_n) \approx \sum_{k=-\frac{N}{2}}^{\frac{N}{2}-1} \tilde{u}(k) e^{i\hat{k}x}$$

- ▶ Where $\hat{k} = \frac{2\pi k}{L}$
- ▶ So differentiation becomes simple multiplication

$$\frac{\partial^2 u}{\partial x^2} \approx - \sum_{k=-\frac{N}{2}}^{\frac{N}{2}-1} \hat{k}^2 \tilde{u}(k) e^{i\hat{k}x}$$

$$\frac{\partial^4 u}{\partial x^4} \approx \sum_{k=-\frac{N}{2}}^{\frac{N}{2}-1} \hat{k}^4 \tilde{u}(k) e^{i\hat{k}x}$$



Solving the spatial derivatives

- ▶ For the nonlinear term, since $u(x, t_n)$ is known and

$$u \frac{\partial u}{\partial x} = \frac{1}{2} \frac{\partial u^2}{\partial x},$$

- ▶ Can calculate $\frac{u^2}{2}$ and then proceed as with the linear terms.

- ▶ Let $v = \frac{u^2}{2}$

- ▶ So from the original equation,

$$\frac{\partial u}{\partial t} = - \frac{\partial v}{\partial x} - \frac{\partial^2 u}{\partial x^2} - \frac{\partial^4 u}{\partial x^4}$$



Solving the spatial derivatives

- ▶ For the nonlinear terms, since $u(x, t_n)$ is known and

$$u \frac{\partial u}{\partial x} = \frac{1}{2} \frac{\partial u^2}{\partial x},$$

- ▶ Can calculate $\frac{u^2}{2}$ and then proceed as with the linear terms.

- ▶ Let $v = \frac{u^2}{2}$

- ▶ We have

$$\frac{\partial \tilde{u}}{\partial t} \approx - \sum_{k=-\frac{N}{2}}^{\frac{N}{2}-1} i\hat{k} \tilde{v}(k) e^{i\hat{k}x} + \sum_{k=-\frac{N}{2}}^{\frac{N}{2}-1} (\hat{k}^2 - \hat{k}^4) \tilde{u}(k) e^{i\hat{k}x}$$



Spectral view of the equation

- ▶ From the spectral view of the equation,

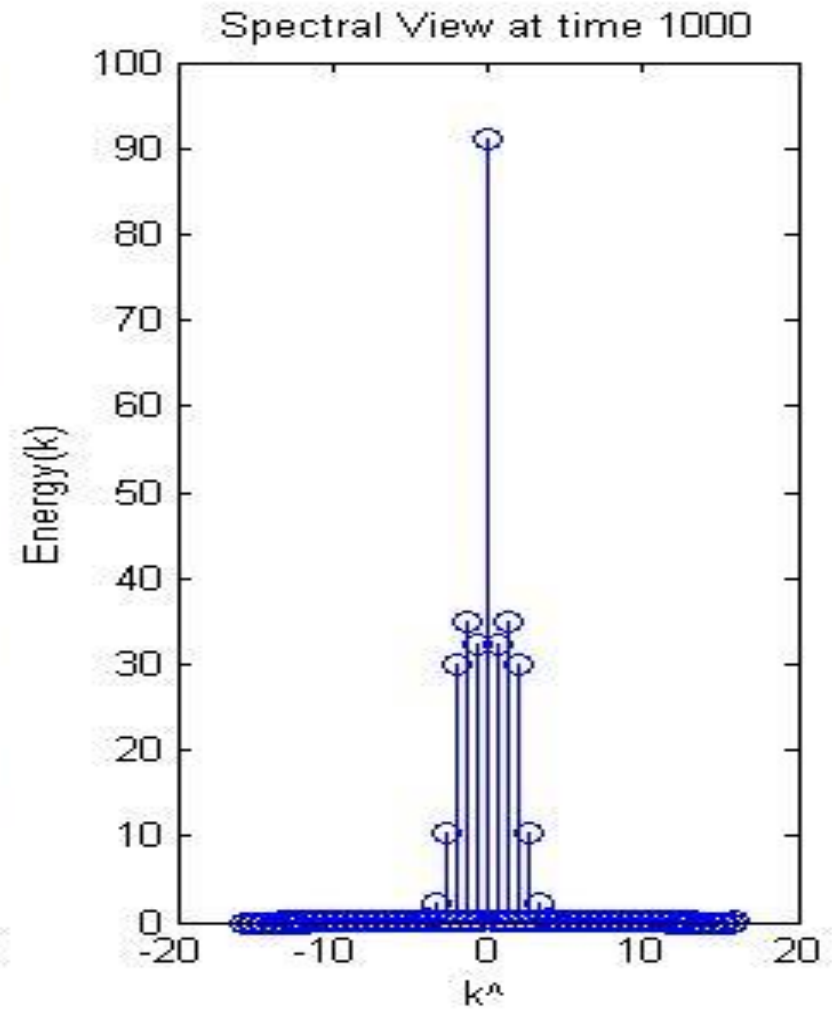
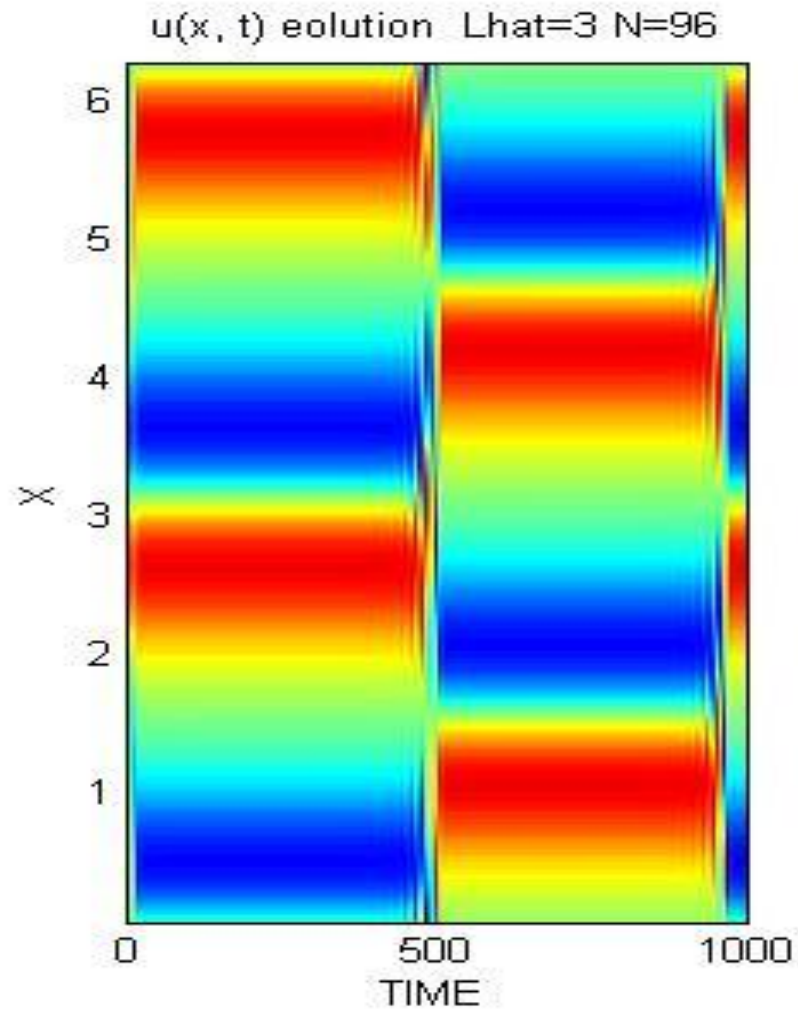
$$\frac{\partial \tilde{u}}{\partial t} = - \sum_{k=-\frac{N}{2}}^{\frac{N}{2}-1} i\hat{k} \tilde{v}(k) e^{i\hat{k}x} + \sum_{k=-\frac{N}{2}}^{\frac{N}{2}-1} (\hat{k}^2 - \hat{k}^4) \tilde{u}(k) e^{i\hat{k}x}$$

- ▶ The 2nd derivative is a forcing term for low wave numbers ($|\hat{k}| > 1$)
- ▶ The 4th derivative is a source of dissipation in the high wave numbers
- ▶ The nonlinear term transfers energy from low to high wave numbers

$$\underbrace{-u \frac{\partial u}{\partial x}}_{\text{distributive}} \quad \underbrace{-\frac{\partial^2 u}{\partial x^2}}_{\text{forcing}} \quad \underbrace{-\frac{\partial^4 u}{\partial x^4}}_{\text{dissipative}}$$



Spectral view of the equation



Time-Stepping methods

- ▶ Basic method:
Leapfrog
- ▶ Two modifications
 1. Leapfrog + periodic Predictor-Corrector,
 2. Robert-Asselin-Williams (RAW) filter
- ▶ How do these two methods compare in the formation of structure in this equation?

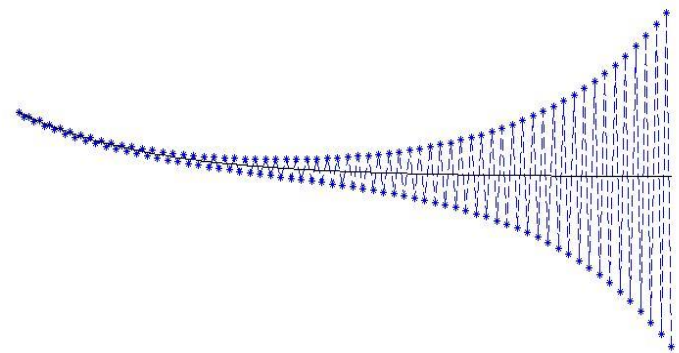
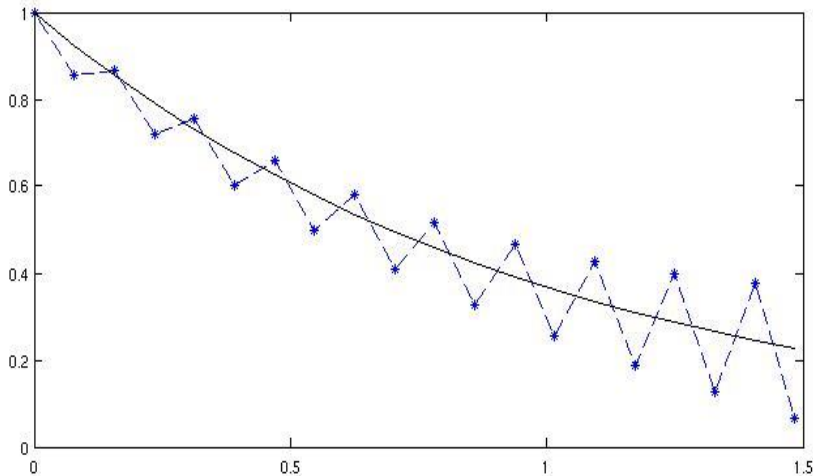


Leapfrog time-stepping scheme

- ▶ The leapfrog scheme (centered difference) is an approximation for the time derivative

- ▶
$$\frac{u(x, t_n + \Delta t) - u(x, t_n - \Delta t)}{2\Delta t} \approx \left. \frac{\partial u}{\partial t} \right|_{t_n} = f(u(x, t_n))$$

- ▶ Where the right hand side is treated as a function f
- ▶ This method is unstable



Predictor Corrector method

- ▶ Stability can be improved even more by restarting every 25 steps in time.
- ▶ When starting from a single initial condition $u(t_0)$, need $u(t_0 + \Delta t)$ to use leapfrog again.
- ▶ Use Forward Euler method* to calculate $u\left(t_0 + \frac{\Delta t}{2}\right)$
- ▶ Use leapfrog on $u(t_0)$ and $u\left(t_0 + \frac{\Delta t}{2}\right)$ to calculate $u(t_0 + \Delta t)$ then continue as before

- ▶ *Forward Euler:
$$\left. \frac{\partial u}{\partial t} \right|_{t_n} \approx \frac{u(t_n + \Delta t) - u(t_n)}{\Delta t}$$



Robert-Asselin-Williams Filter

- ▶ A separate improvement on the basic leapfrog scheme is the RAW filter

$$\frac{u(x, t_n + \Delta t) - \overline{\overline{u(x, t_n - \Delta t)}}}{2\Delta t} = f(\overline{u(x, t_n)})$$

- ▶ Where,

$$\overline{u(x, t_n)} = u(x, t_n) + \frac{v(1 - \alpha)}{2} [\overline{\overline{u(x, t_n - 2\Delta t)}} - 2\overline{u(x, t_n - \Delta t)} + u(x, t_n)]$$

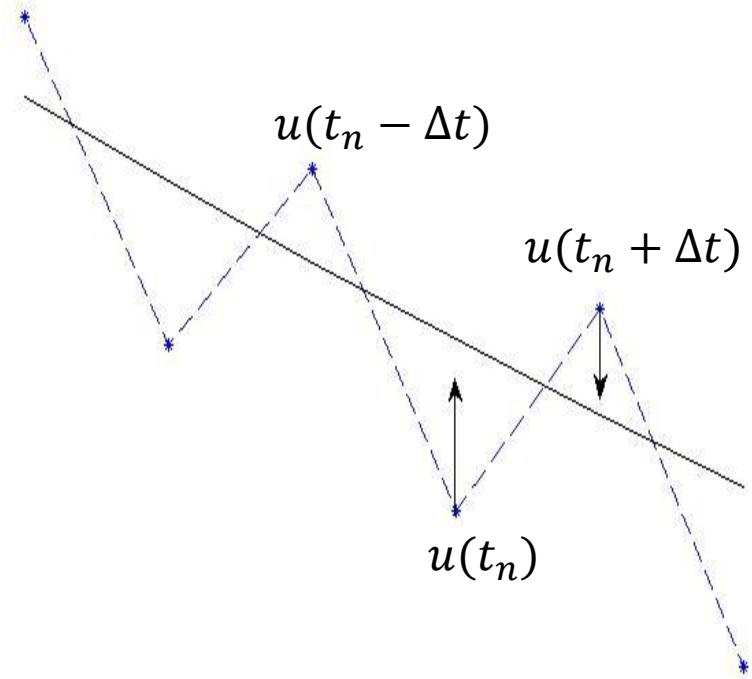
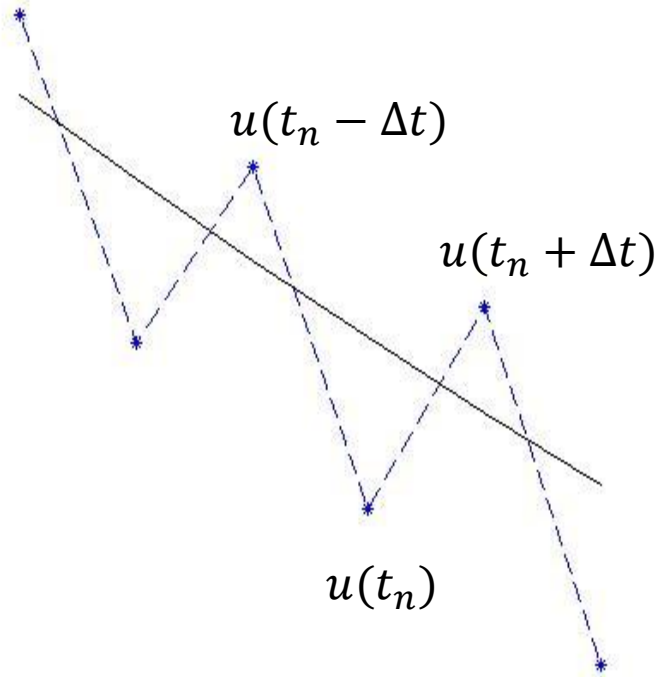
- ▶ And,

$$\overline{\overline{u(x, t_n - \Delta t)}} = \overline{u(x, t_n - \Delta t)} - \frac{v\alpha}{2} [\overline{\overline{u(x, t_n - 2\Delta t)}} - 2\overline{u(x, t_n - \Delta t)} + u(x, t_n)]$$

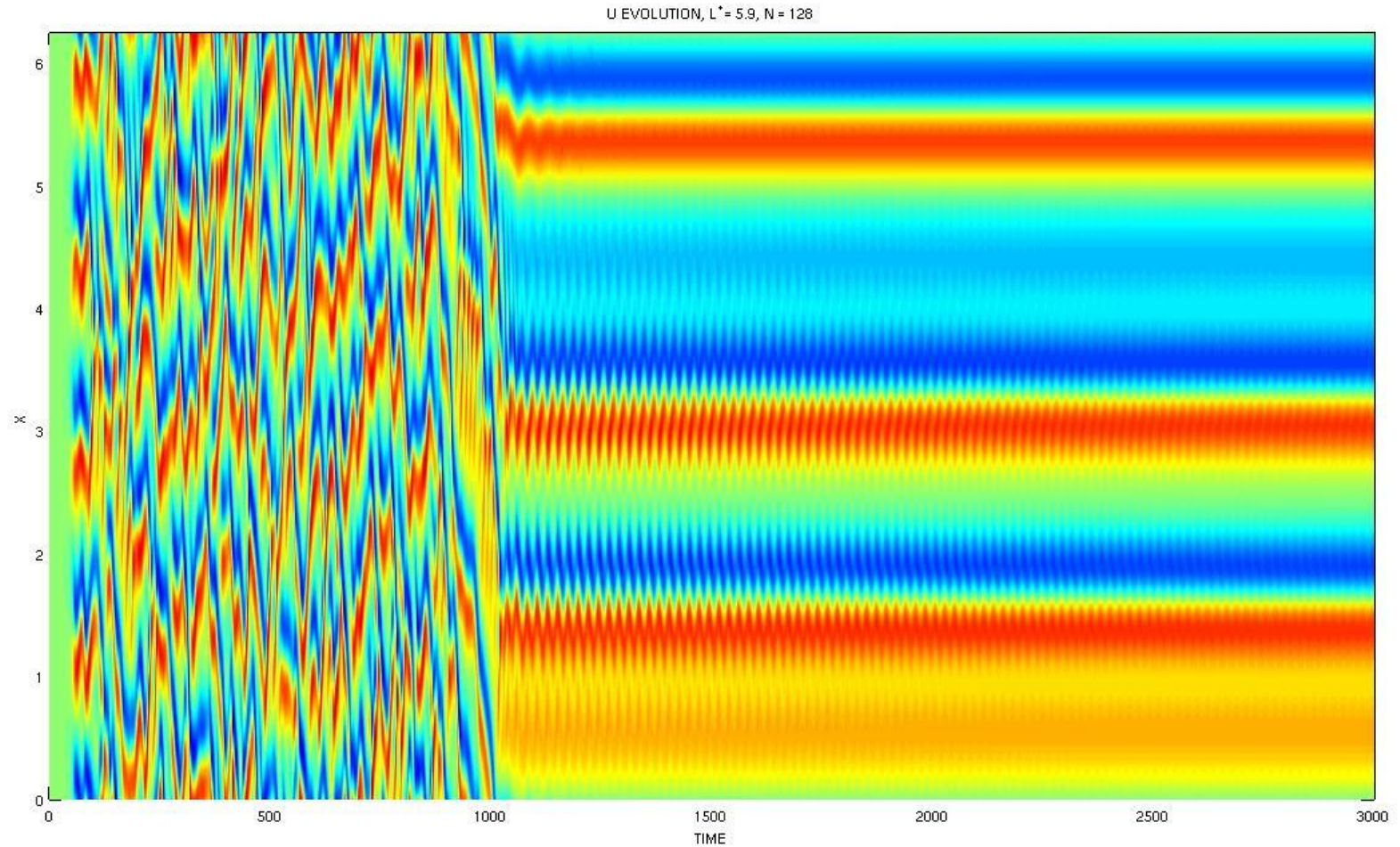


Robert-Asselin-Williams Filter

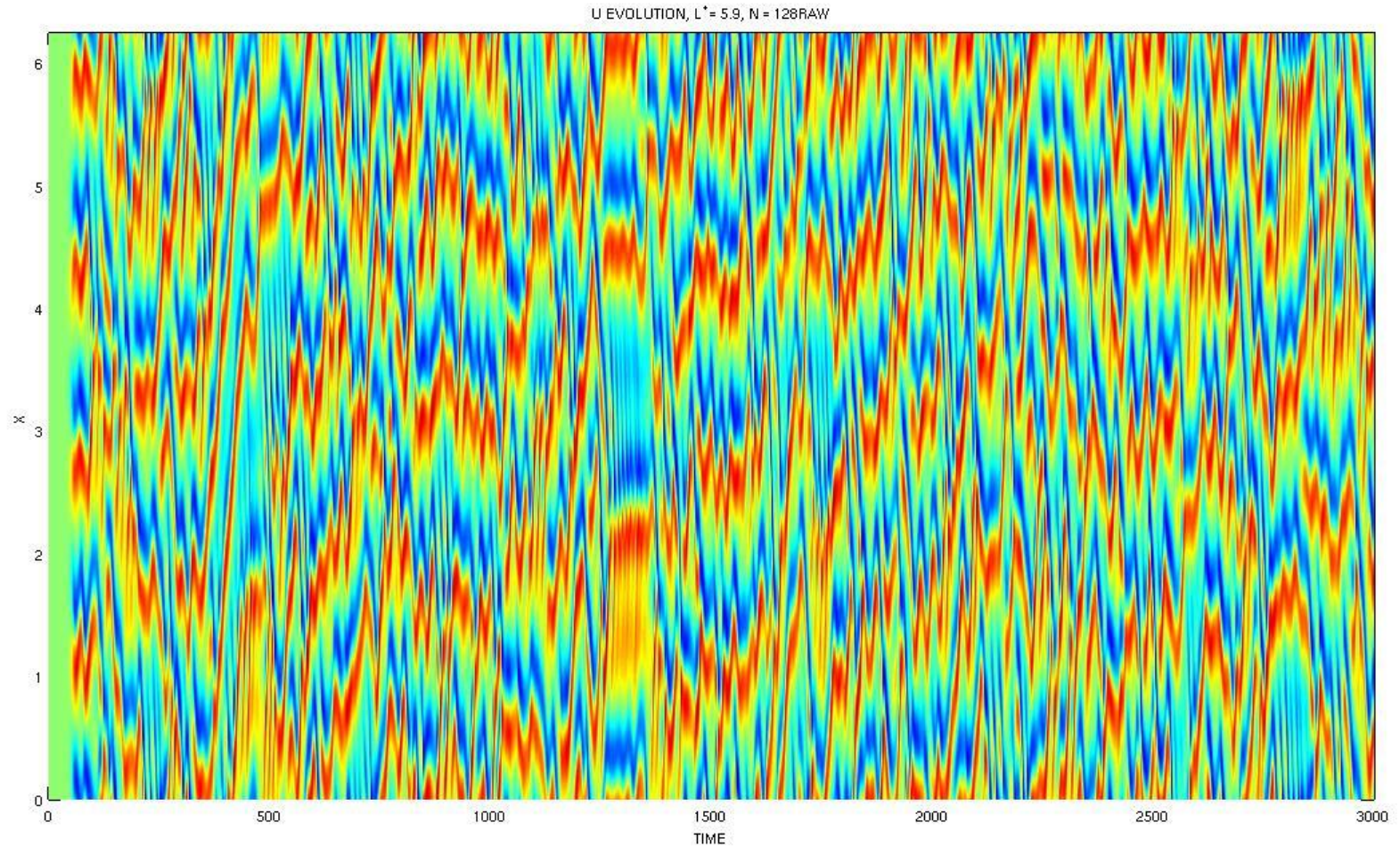
- ▶ What this is saying is we compute the next time step
- ▶ Then push $u(t_n)$ and $u(t_n + \Delta t)$ towards the



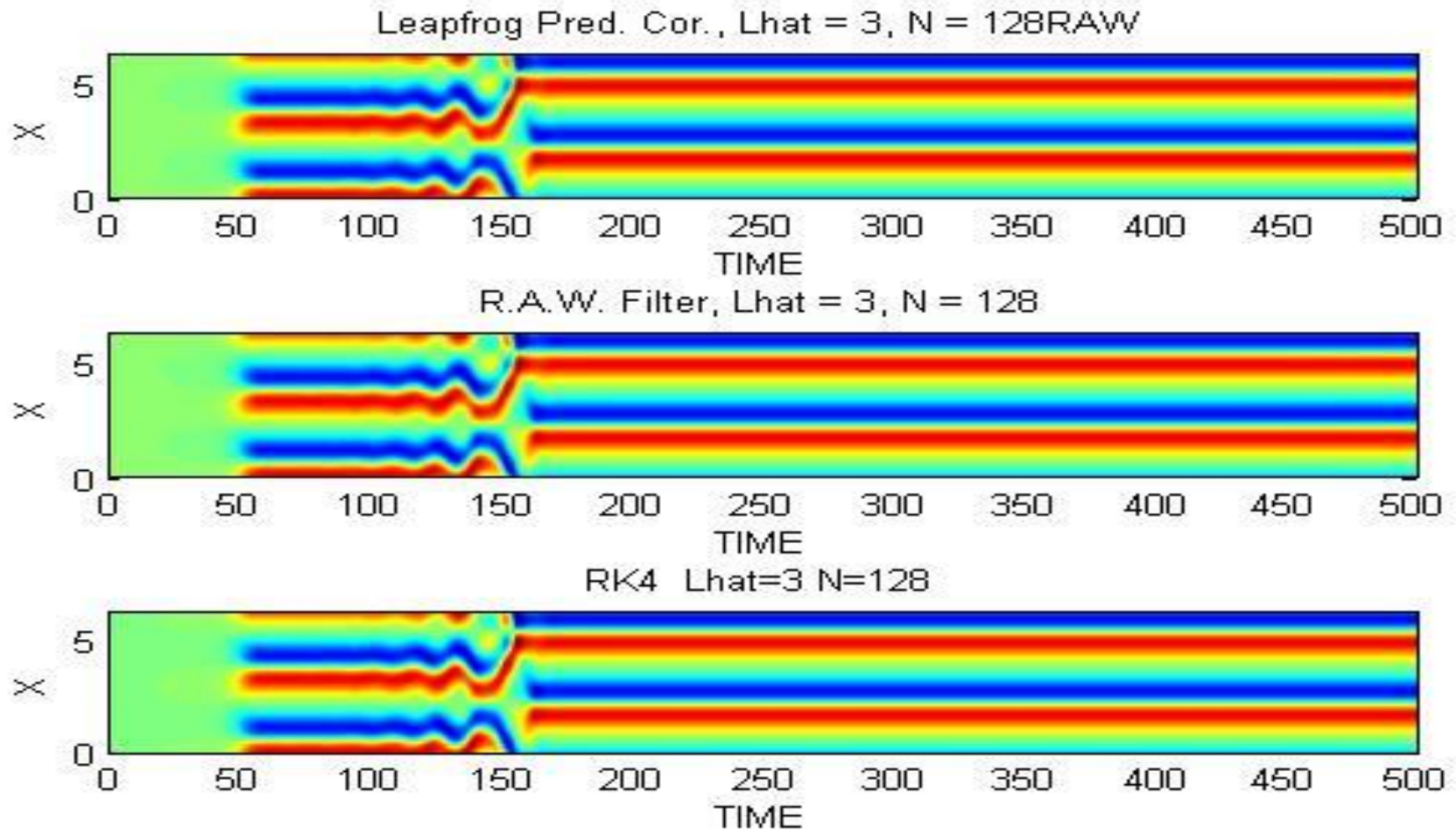
What is the problem?



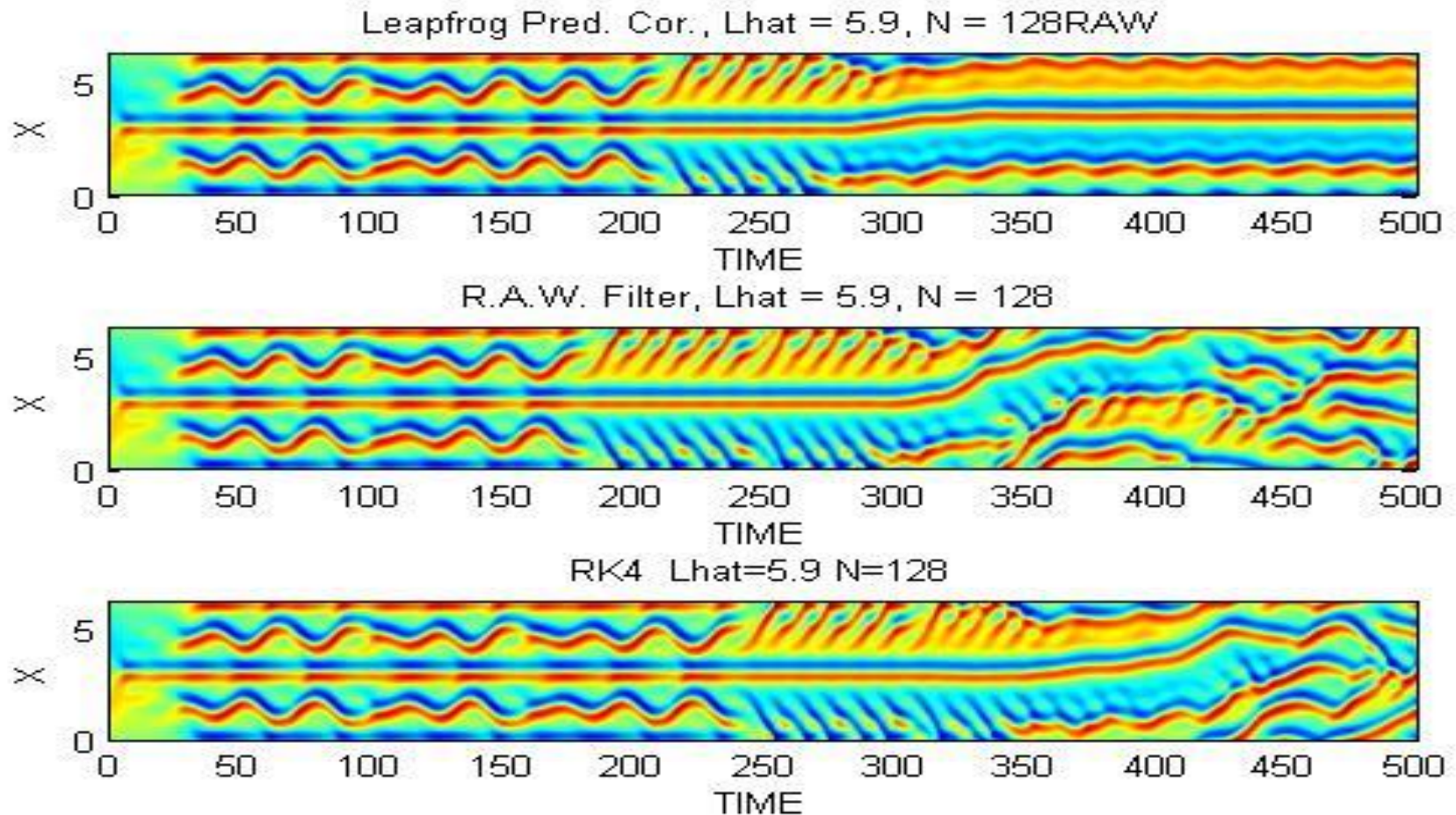
What is the problem?



Compared to more accurate method



Compared to more accurate method



Conclusion: Time-stepping method matters

- ▶ 4th order Runge-Kutta method takes time and memory
- ▶ Can also be difficult to implement in existing code
- ▶ In general RAW gives a better idea of the behavior of the solution than the Predictor Corrector method
- ▶ It is also very simple to update existing code
- ▶ Williams has produced a more general filter to give up to 7th order accuracy
- ▶ When looking at the development of structure time-stepping matters!

