

# Simultaneous Space-Time Adaptive Solution of Partial Differential Equations \*

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McMaster University, Canada

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\*Adaptive wavelet and multiscale methods for partial differential equations June 3 - 5, 2004, Banff International Research Station

# Collaborators

- O. V. Vasilyev (University of Colorado at Boulder)
- D. Goldstein (University of Colorado at Boulder)

# Outline

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- Motivation

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- Conclusion and future direction

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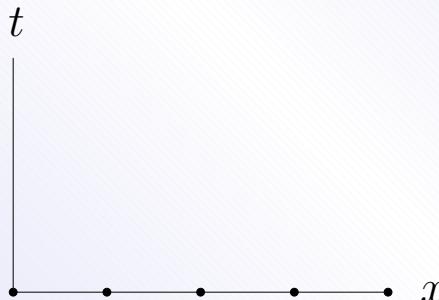
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- **Uniform grid for such a problem is not suitable**

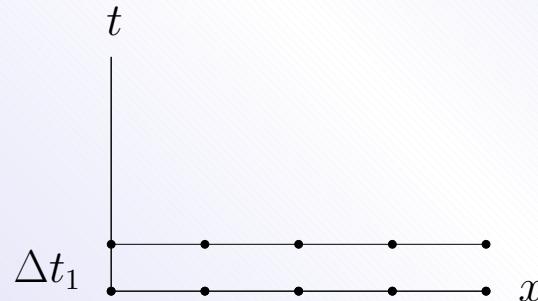
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- Grid should adapt in space and time



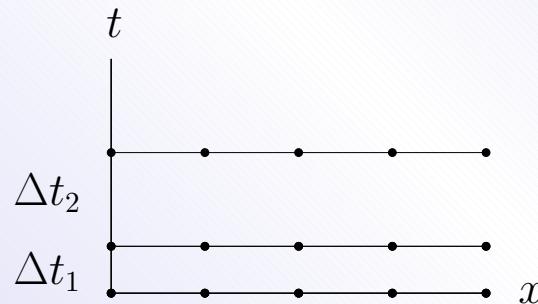
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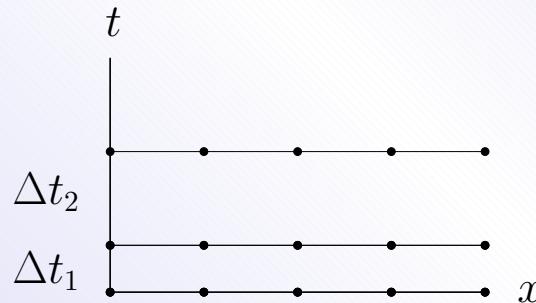
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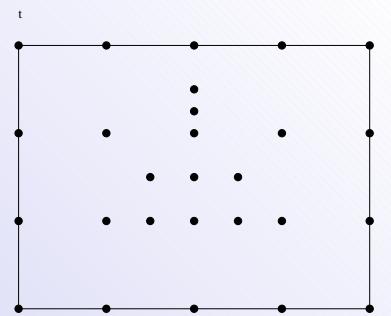
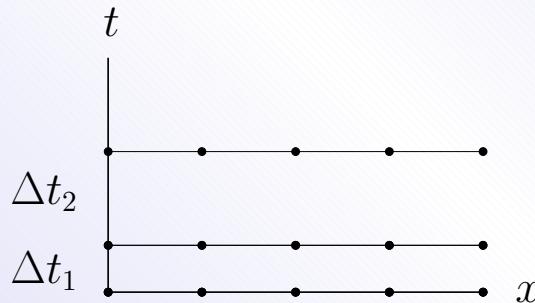
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← **Space-time adaptive grid**

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$$u(x) = \sum_{j=0}^{\infty} \sum_{k \in \mathcal{K}^j} d_k^j \psi_k^j(x)$$

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- **Represent a function in terms of wavelet basis:**

$$u(x) = \sum_{j=0}^{\infty} \sum_{k \in \mathcal{K}^j} d_k^j \psi_k^j(x)$$

- **Wavelets:**

- follow intermittency in position and scale
- provide automatic grid adaptation

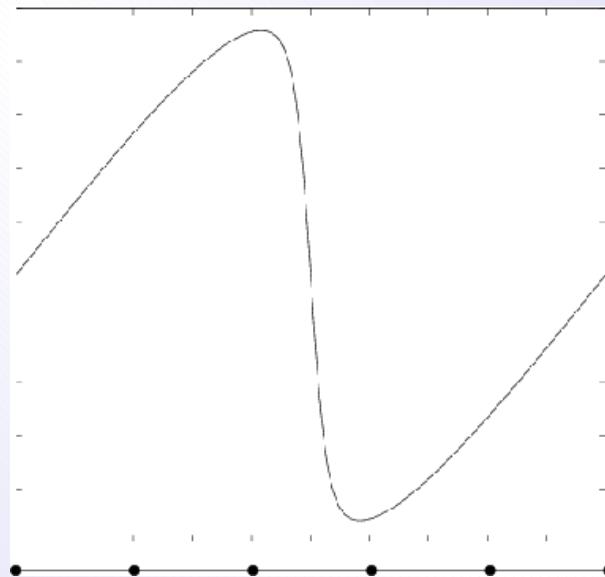
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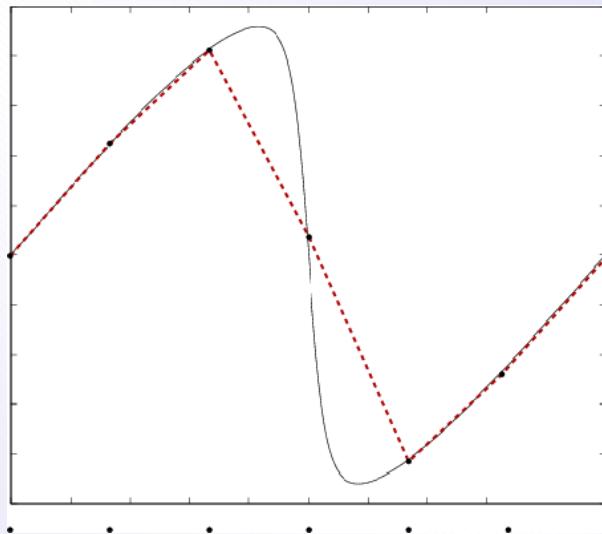
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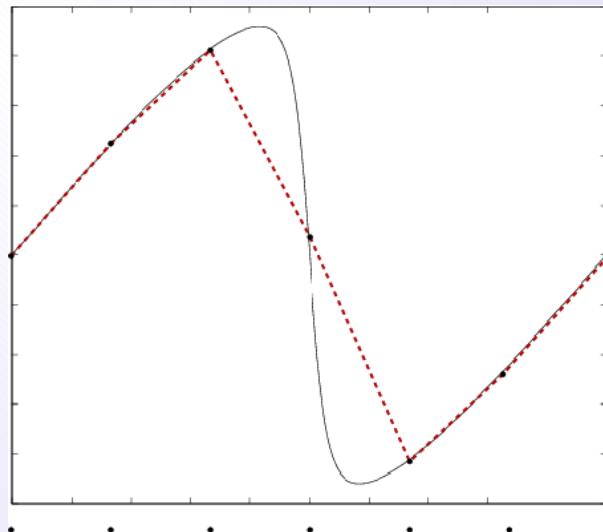
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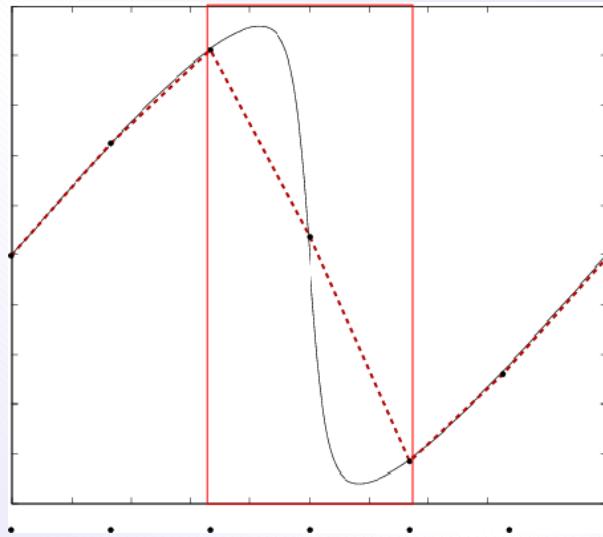
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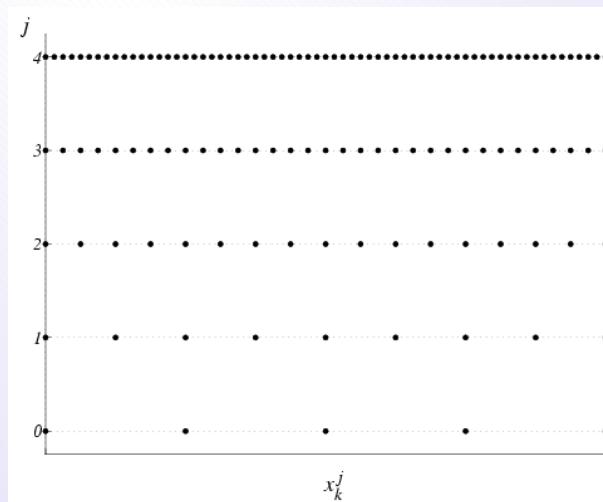
- Nested dyadic grid:

$$G^j = \{x_k^j \in \mathbb{R} : x_k^j = 2^{-j}k, k \in \mathcal{Z}, j \in \mathcal{Z}\}$$

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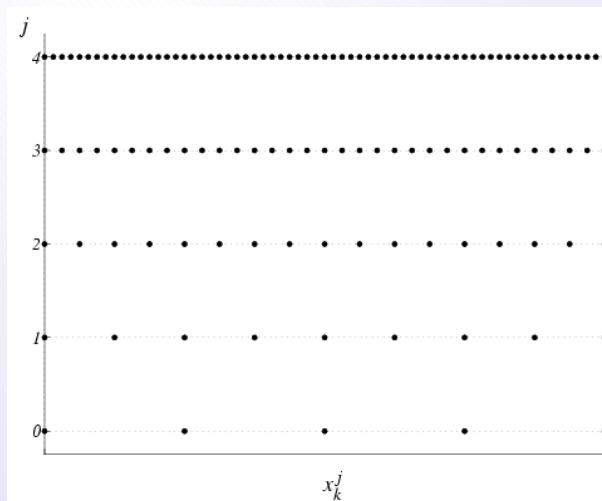
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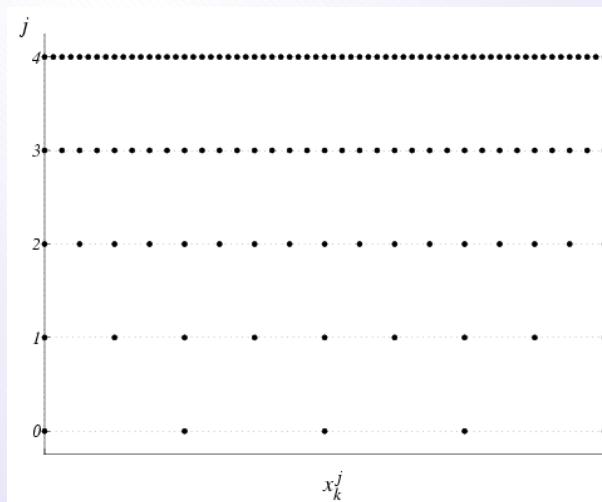


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- Easy to see the nestedness property:

$$G^j \subset G^{j+1} \quad \text{i.e.} \quad x_{2k}^{j+1} = x_k^j$$

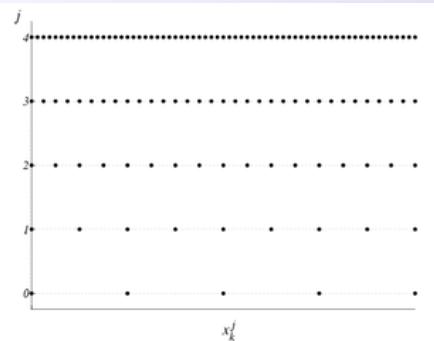
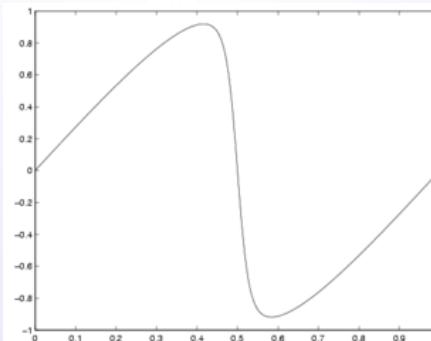
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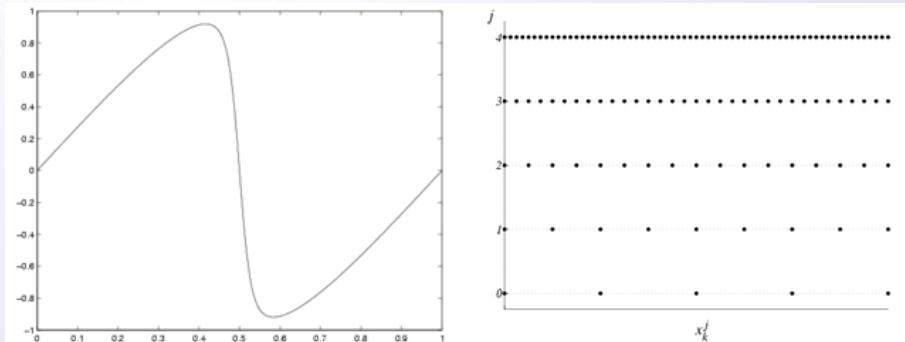
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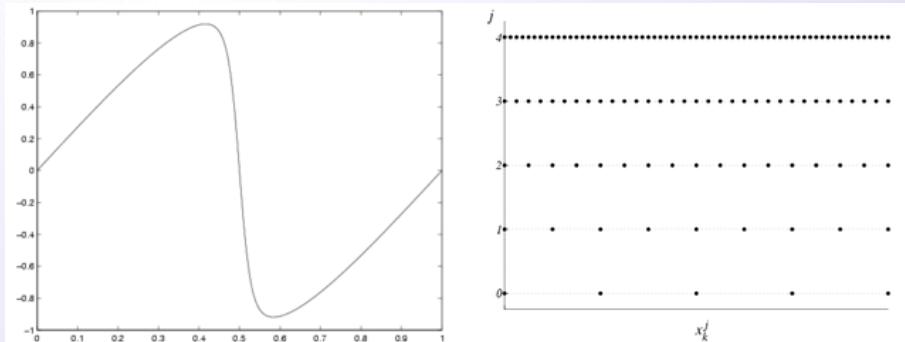
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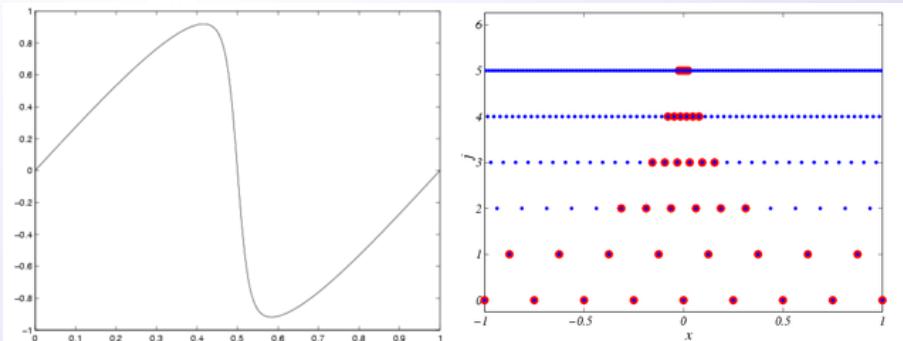


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- Classical solution procedure

Sequence of algebraic problem (via ODE solver)

- Our goal

A single algebraic problem

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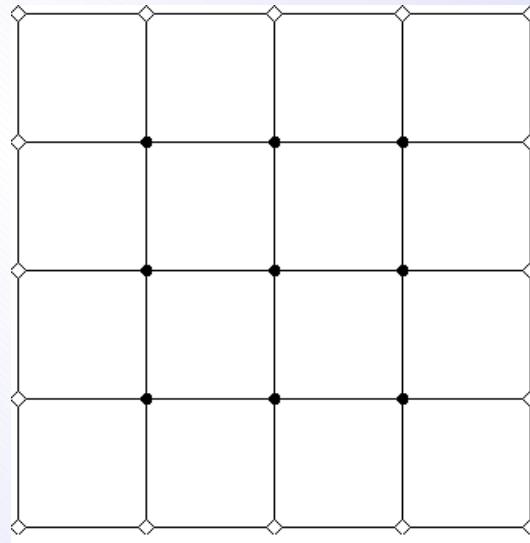
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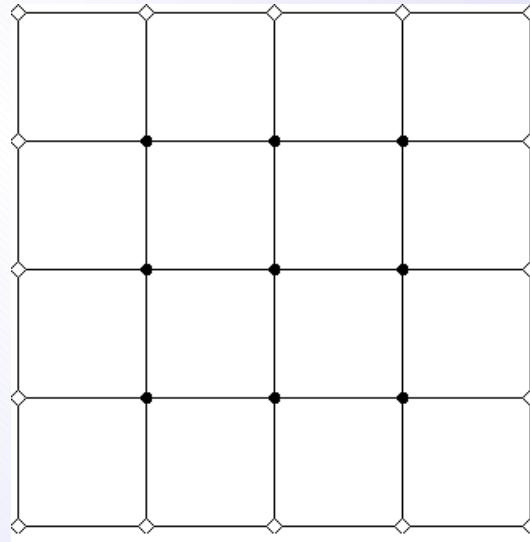
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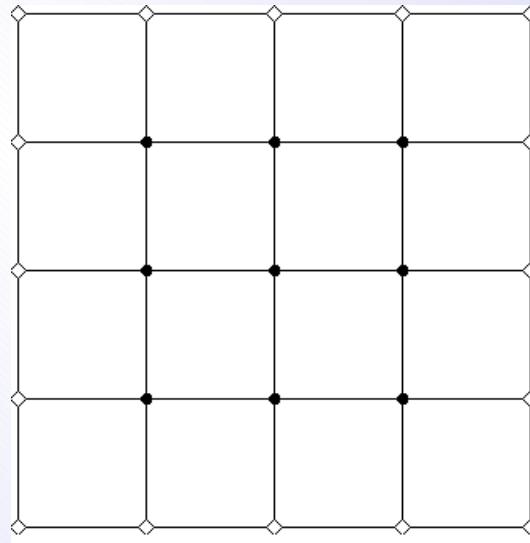
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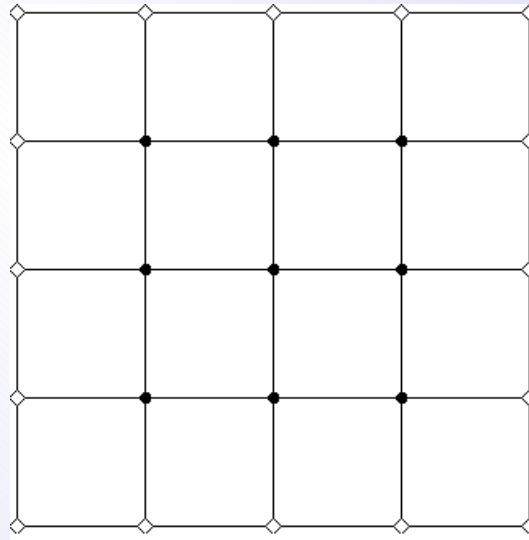
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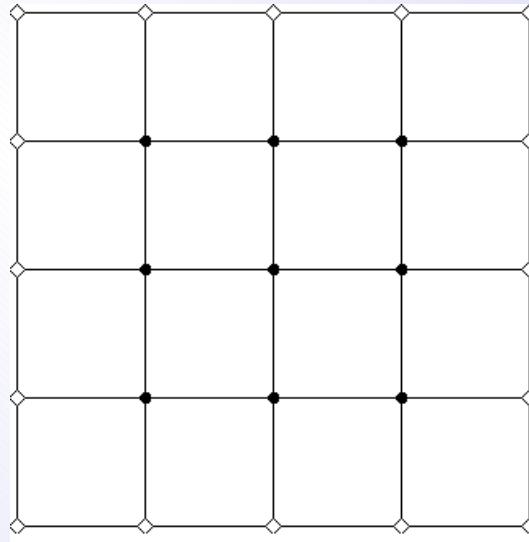


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Solve DE on internal points

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Solve DE on internal points  
Implement BC on Boundary points

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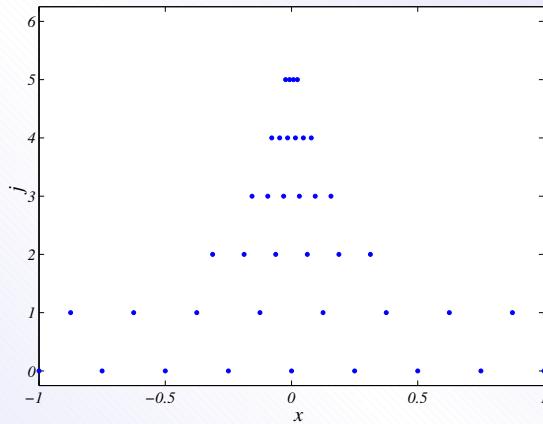
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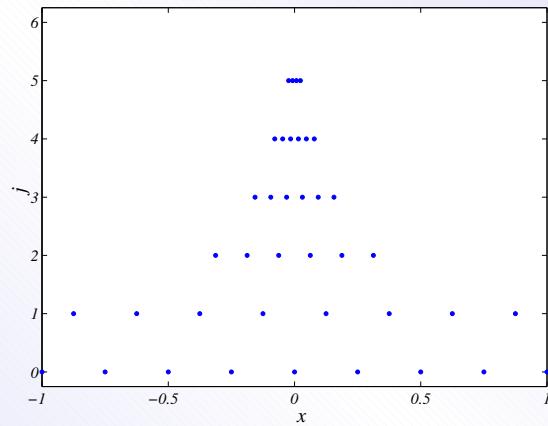
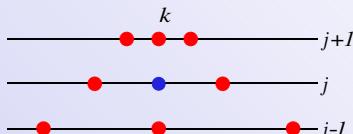


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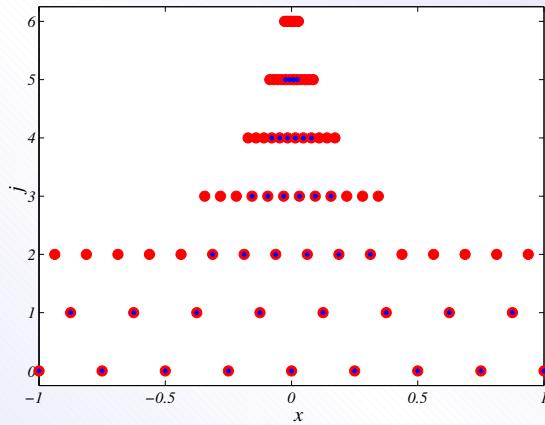
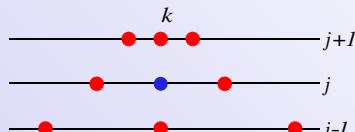


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We propose:

$$\mathcal{L}u - f = 0 \quad \text{for } t = t_{\max}$$

# Nonlinear evolution problem

- Example:

Navier-Stokes

$$\partial_t \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} = -\frac{1}{\rho} \nabla P + \nu \nabla^2 \mathbf{u}$$

Kuramoto-Sivashinsky

$$\partial_t u + \partial_{xxxx} u + \partial_{xx} u + u \partial_x u = 0$$

- Can we reduce to an algebraic problem?

YES

- What is the boundary condition at fixed time?

We propose:

$$\mathcal{L}u - f = 0 \quad \text{for } t = t_{\max}$$

evolution type boundary condition.

# Multilevel elliptic solver

**V-cycle:**

$$\mathbf{r}^J = \mathbf{f}^J - \mathbf{L}\mathbf{u}^J$$

**for** all levels  $j = J : -1 : j_{\min} + 1$

**do**  $\nu_1$  steps of **approximate** solver for  $\mathbf{L}\mathbf{v}^j = \mathbf{r}^j$

$$\mathbf{r}^{j-1} = \mathbf{I}_w^{j-1}(\mathbf{r}^j - \mathbf{L}\mathbf{v}^j)$$

**enddo**

**end**

**Solve** for  $j = j_{\min}$  level:  $\mathbf{L}\mathbf{v}^j = \mathbf{r}^j$

**for** all levels  $j = j_{\min} + 1 : +1 : J$

$$\mathbf{v}^j = \mathbf{v}^j + \omega_0 \mathbf{I}_w^j \mathbf{v}^{j-1}$$

**do**  $\nu_2$  steps of **approximate** solver for  $\mathbf{L}\mathbf{v}^j = \mathbf{r}^j$  **enddo**

**end**

$$\mathbf{u}^J = \mathbf{u}^J + \omega_1 \mathbf{v}^J$$

**do**  $\nu_3$  steps of **exact** solver for  $\mathbf{L}\mathbf{u}^J = \mathbf{f}^J$  **enddo**

# Adaptive nonlinear solver

**V-cycle:**

$$\mathbf{r}^J = \mathbf{f}^J - \mathbf{L}\mathbf{u}^J$$

**for** all levels  $j = J : -1 : j_{\min} + 1$

**do**  $\nu_1$  steps of **approximate** solver for  $\mathbf{J}(u)\mathbf{v}^j = \mathbf{r}^j$

$$\mathbf{r}^{j-1} = \mathbf{I}_w^{j-1}(\mathbf{r}^j - \mathbf{J}(u)\mathbf{v}^j)$$

**enddo**

**end**

**Solve** for  $j = j_{\min}$  level:  $\mathbf{J}(u)\mathbf{v}^j = \mathbf{r}^j$

**for** all levels  $j = j_{\min} + 1 : +1 : J$

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**end**

$$\mathbf{u}^J = \mathbf{u}^J + \omega_1 \mathbf{v}^J$$

**enddo**

# Result and discussion

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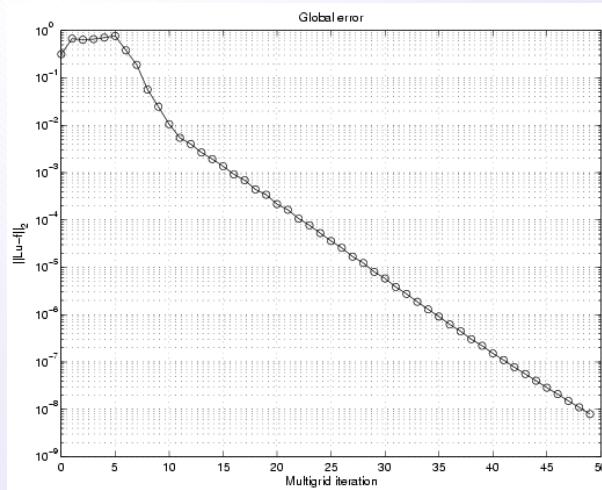
- Adaptive nonlinear solver

$$\partial_t u + \partial_{xxxx} u + \partial_{xx} u + u \partial_x u = 0$$

# Result and discussion

- Adaptive nonlinear solver

$$\partial_t u + \partial_{xxxx} u + \partial_{xx} u + u \partial_x u = 0$$



$L_2$  norm of residual as a function of multigrid iteration

# Result and discussion: Cont'd...

- Burgers equation

# Result and discussion: Cont'd...

- Burgers equation

$$\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} = \nu \frac{\partial^2 u}{\partial x^2}, \quad x \in \Omega \subset \mathbb{R} \times [0, t_{\max}], \Omega = [0, 1]$$

$$u(0, t) = u(1, t), \quad u(x, 0) = \sin(2\pi x)$$

$$\nu = 10^{-2}$$

# Result and discussion: Cont'd...

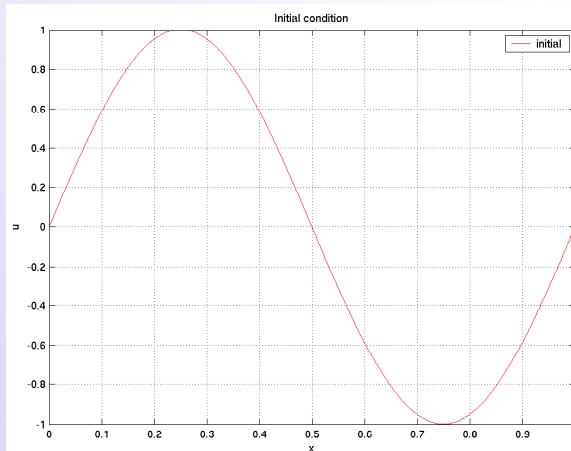
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$$u(0, t) = u(1, t), \quad u(x, 0) = \sin(2\pi x)$$

$u$   
↑  
 $x$

$$\nu = 10^{-2}$$



Initial condition

# Result and discussion: Cont'd...

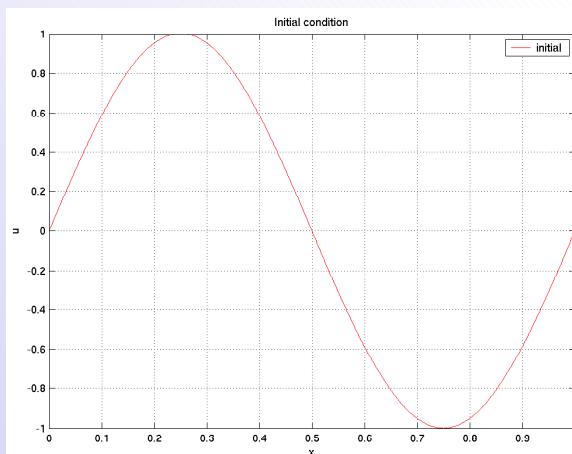
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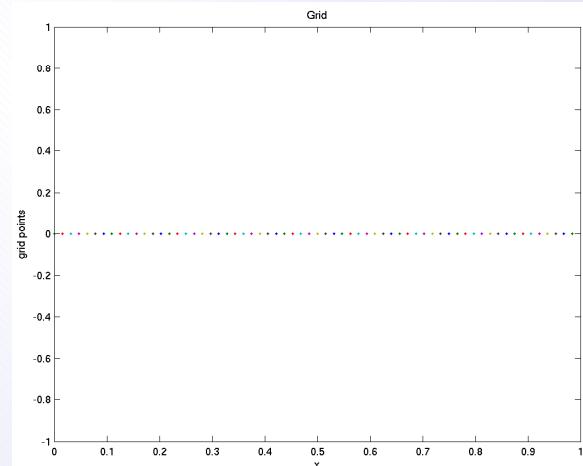
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Initial condition



Grid

# Result and discussion: Cont'd...

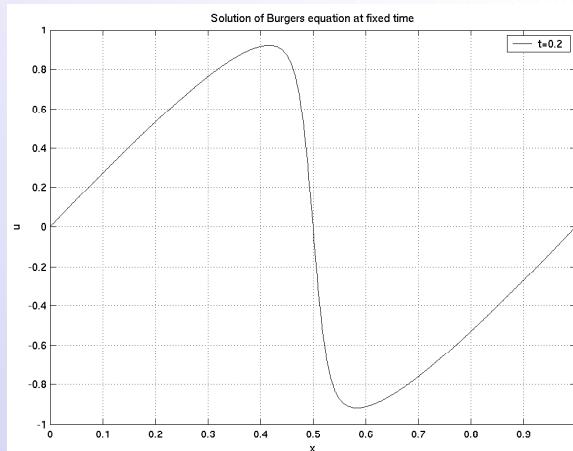
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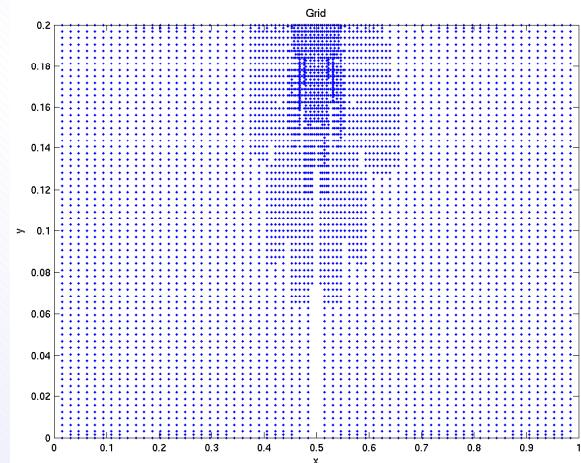
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 $x$

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Computed solution



Adapted grid

# Result and discussion: Cont'd...

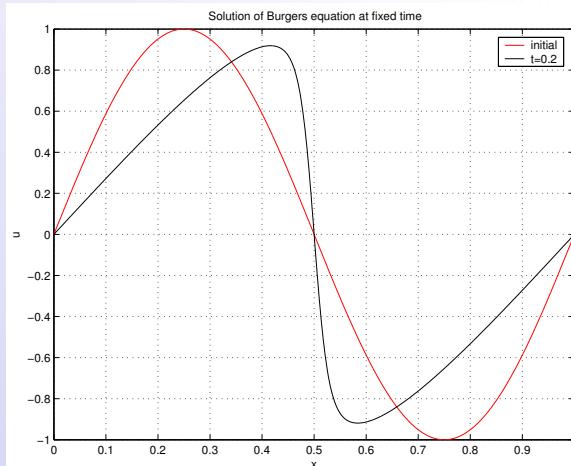
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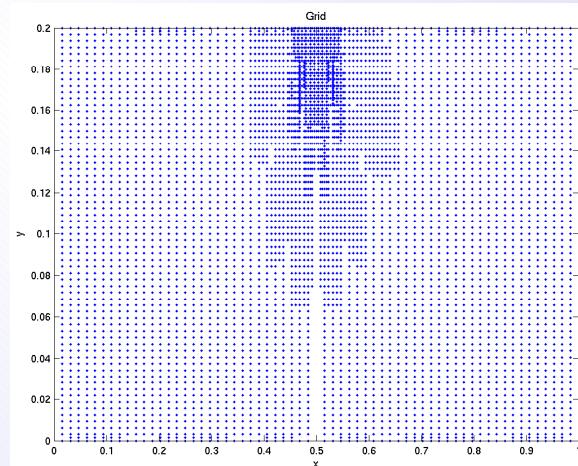
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Computed solution



Adapted grid

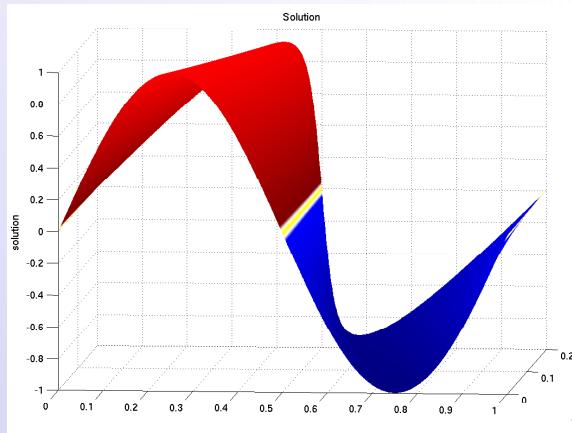
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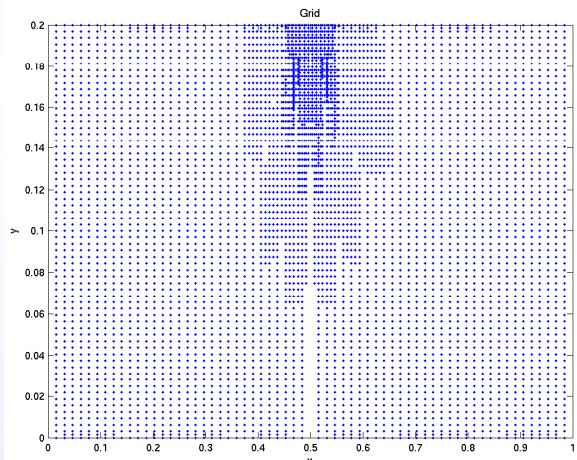
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**Solution surface**  
(Space-time domain)

$$\nu = 10^{-2}$$



**Adapted grid**

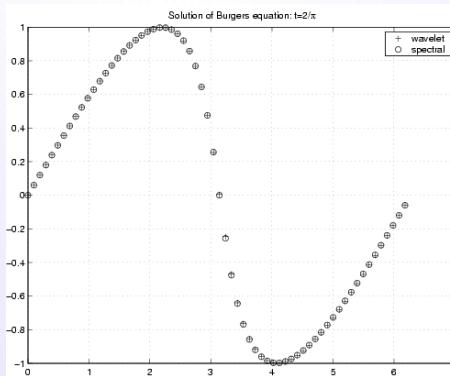
# Result and discussion: Cont'd...

- Burgers equation

$$\frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} = \nu \frac{\partial^2 u}{\partial x^2}, \quad x \in (-\pi, \pi)$$

$$u(-\pi, t) = u(\pi, t)$$

$$u(x, 0) = \sin(x)$$



Compare wavelet solution with a spectral code.

# Result and discussion: Cont'd...

- Moving shock

## Result and discussion: Cont'd...

- **Moving shock**

$$\frac{\partial u}{\partial t} + (u + v) \frac{\partial u}{\partial x} = \nu \frac{\partial^2 u}{\partial x^2}, \quad x \in \Omega \times [0, t_{\max}], \quad \Omega = (0, 2)$$

$$u(0, t) = 1, \quad u(2, t) = -1, \quad u(x, 0) = -\tanh\left(\frac{x - x_0}{2\nu}\right)$$

$$\nu = 10^{-2}, \quad x_0 = 0.5$$

# Result and discussion: Cont'd...

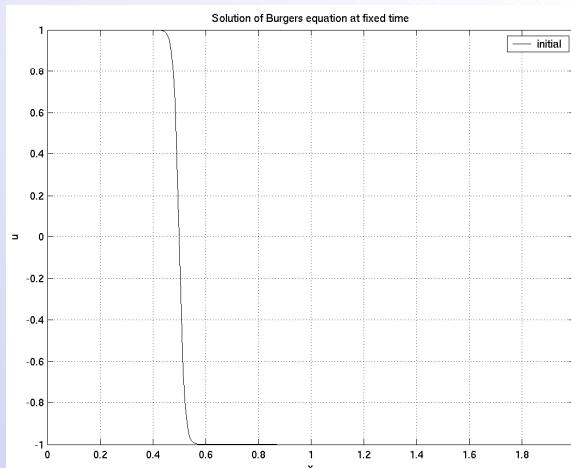
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**Initial condition**

# Result and discussion: Cont'd...

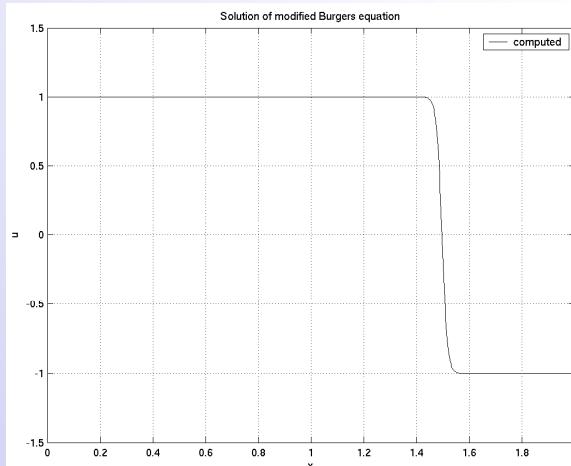
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**Solution at  $t = 1.0$**

# Result and discussion: Cont'd...

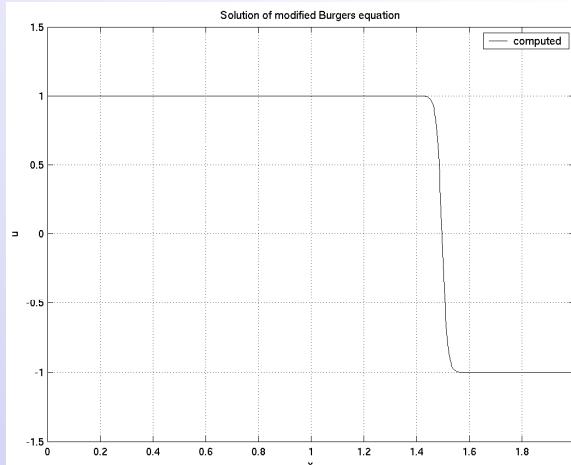
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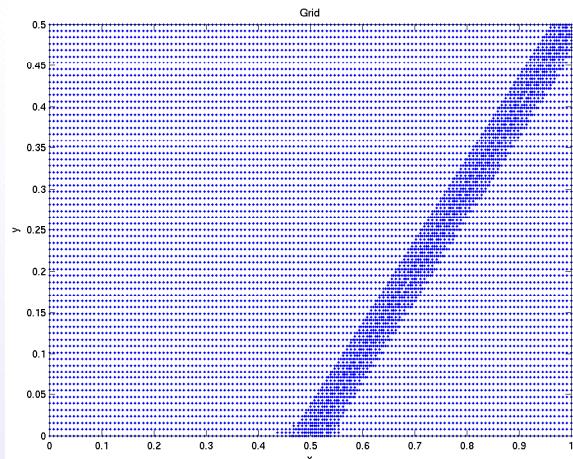
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**Solution at  $t = 1.0$**



**Adapted grid**

# Result and discussion: Cont'd...

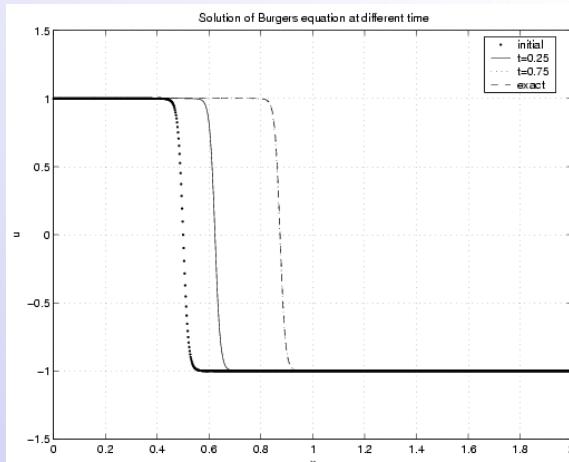
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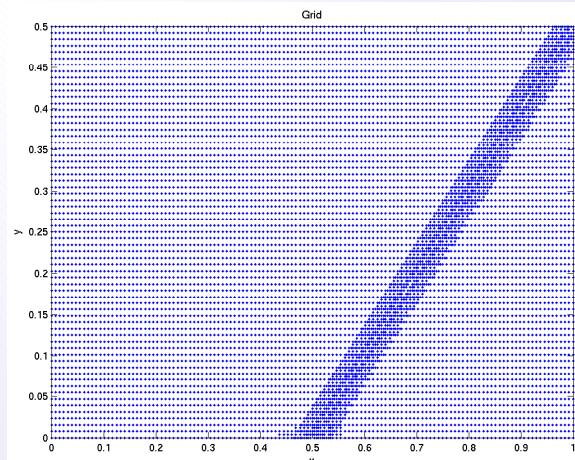
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Solution

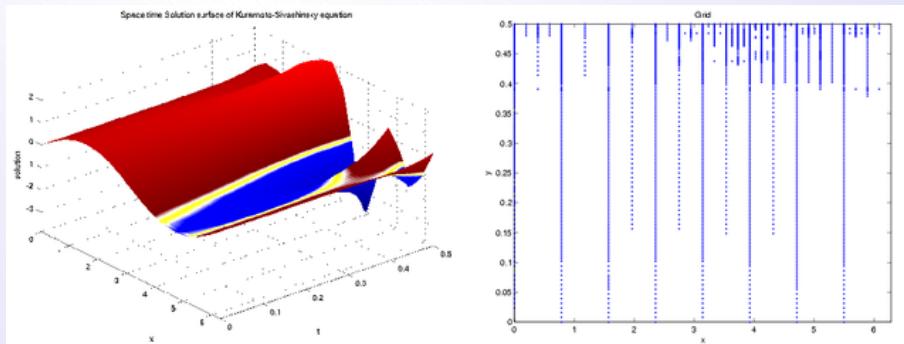


Adapted grid

# Result and discussion: Cont'd...

- Kuramoto-Sivashinsky equation

$$\frac{\partial u}{\partial t} + \nu_4 \partial_x^4 u + \partial_x^2 u + u \partial_x u = 0, \quad x \in \Omega \times [0, t_{\max}], \quad \Omega = [0, 2\pi]$$

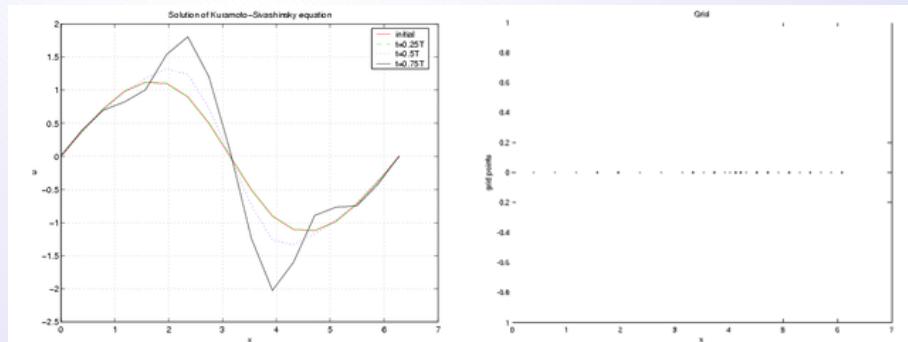


Space-time solution surface and corresponding grid

# Result and discussion: Cont'd...

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Fixed time solution and corresponding grid

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    - \* Better time stepping

- Solution

- \* flip and solve method
    - \* Lagrangian or variational idea

# Thank You