

Sensitivity analysis for nonlinear hyperbolic equations



21/6/2016 - Junior Seminar

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Where am I from?

Advisors: Régis Duvigneau (INRIA Sophia Antipolis),
Christophe Chalons (LMV UVSQ).

University: Université Paris Saclay -
Université de Versailles Saint-Quentin-en-Yvelines.

Lab: Laboratoire de Mathématiques de Versailles

- ▶ Analysis and PDEs
- ▶ Probability and Statistics
- ▶ Algebra
- ▶ Cryptography

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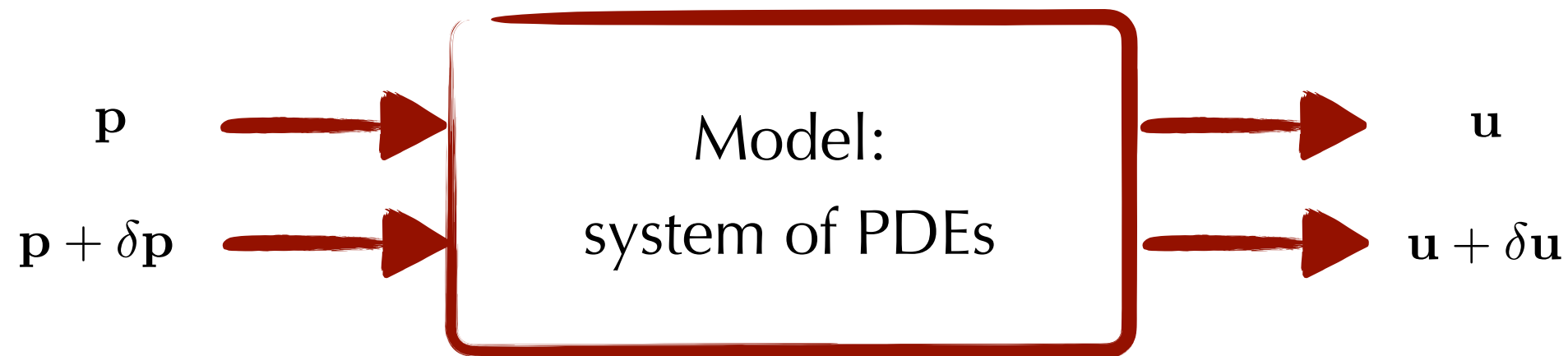
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- ▶ **Analysis and PDEs**
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Sensitivity Analysis

Sensitivity Analysis: the study of how **variations in the output** of a model can be attributed to different sources of **uncertainty in the model input**.



Therefore, we want to study the derivative of \mathbf{u} with respect to \mathbf{p} :

$$\mathbf{u}_{\mathbf{p}} = \frac{\partial \mathbf{u}}{\partial \mathbf{p}}$$

Applications

- ▶ Propagation of uncertainty or error: sensitivity can be used to study how uncertainty in a measurement of a parameter can affect the solution.
- ▶ Estimate of close solutions: using a first order Taylor expansion it is possible to estimate solution for different parameters values.

$$\mathbf{u}(\mathbf{p} + \delta\mathbf{p}) \simeq \mathbf{u}(\mathbf{p}) + \delta\mathbf{p}\mathbf{u}_{\mathbf{p}}(\mathbf{p})$$

- ▶ Optimisation: sensitivity can be useful to solve problems such as

$$\min_{\mathbf{p} \in \mathcal{P}} J(\mathbf{u}(\mathbf{p}))$$

for which it is necessary to compute the gradient of the cost functional:

$$\nabla_{\mathbf{p}} J = \frac{\partial J}{\partial \mathbf{u}} \mathbf{u}_{\mathbf{p}}$$

State equations

We will consider **hyperbolic equations**:

$$\begin{cases} \partial_t \mathbf{u} + \partial_x f(\mathbf{u}) = 0 & x \in \mathbb{R}, t > 0 \\ \mathbf{u}(x, 0) = \mathbf{g}(x; \mathbf{p}) & x \in \mathbb{R}. \end{cases}$$

Hyperbolic equations are also known as **conservation laws**:

- ▶ \mathbf{u} is the **conserved variable**
- ▶ $f(\mathbf{u})$ is the **flux function**
- ▶ $\mathbf{g}(x; \mathbf{p})$ is the **initial condition**

Sensitivity equations

Under hypothesis of **regularity**, one can differentiate the state equations with respect to the parameter:

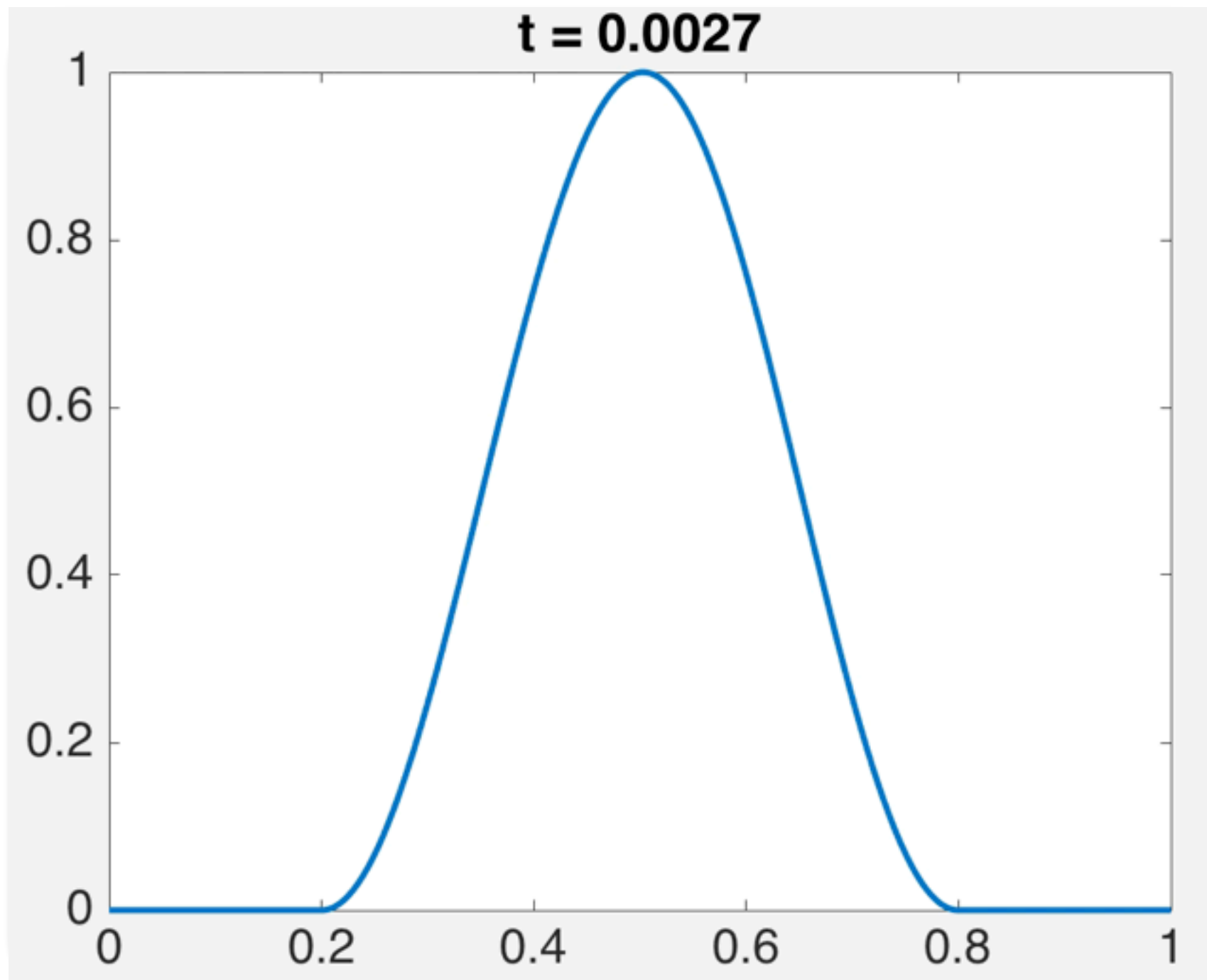
$$\begin{cases} \partial_{\mathbf{p}}(\partial_t \mathbf{u}) + \partial_{\mathbf{p}}(\partial_x f(\mathbf{u})) = 0 & x \in \mathbb{R}, t > 0 \\ \partial_{\mathbf{p}} \mathbf{u}(x, 0) = \partial_{\mathbf{p}} \mathbf{g}(x; \mathbf{p}) & x \in \mathbb{R}. \end{cases}$$

Exchanging the derivatives in space and time with the ones with respect to the parameter one has:

$$\begin{cases} \partial_t \mathbf{u}_{\mathbf{p}} + \partial_x (f'(\mathbf{u}) \mathbf{u}_{\mathbf{p}}) = 0 & x \in \mathbb{R}, t > 0 \\ \mathbf{u}_{\mathbf{p}}(x, 0) = \mathbf{g}_{\mathbf{p}}(x; \mathbf{p}) & x \in \mathbb{R}. \end{cases}$$

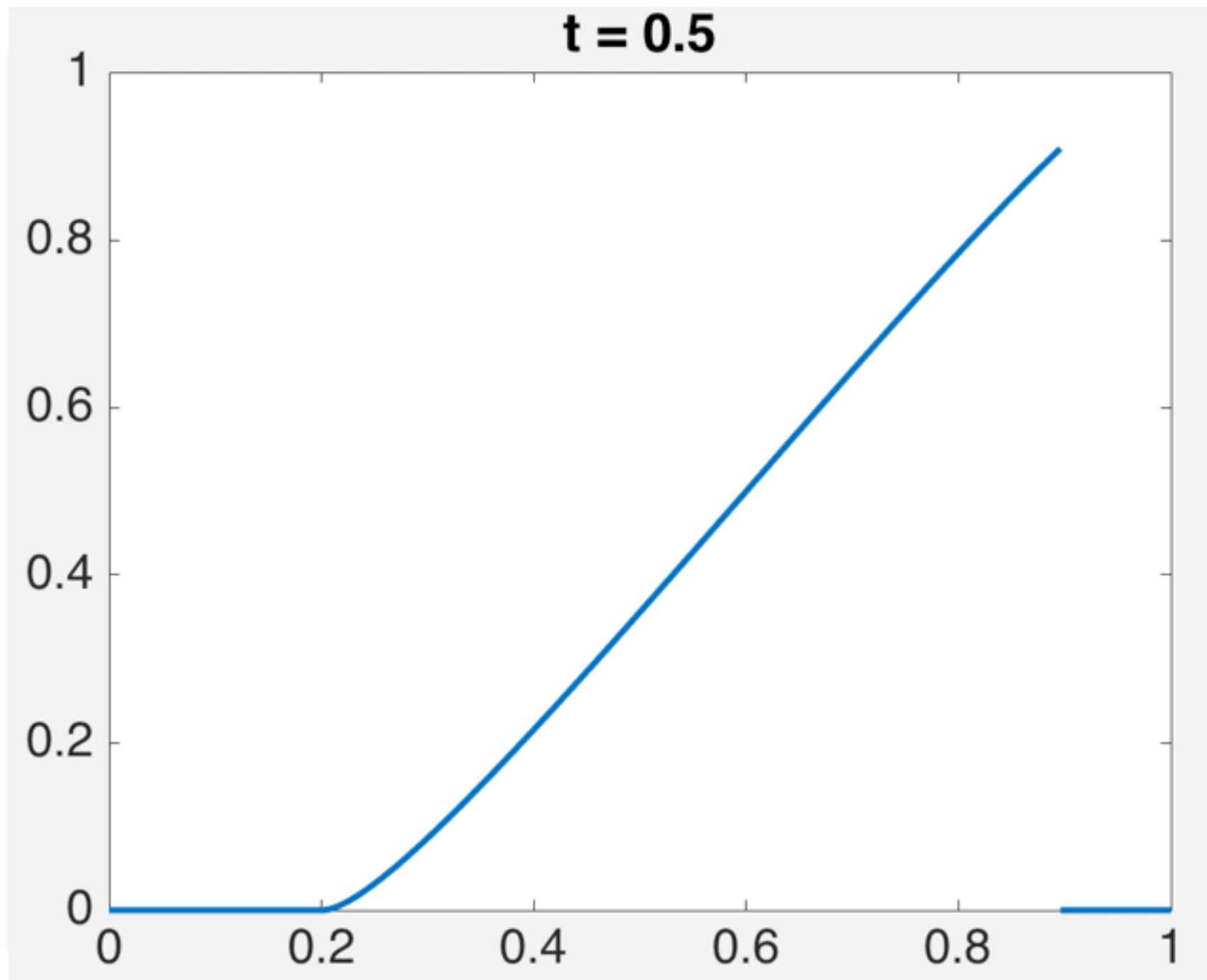
Example: the Burger's equation

$$f(u) = \frac{u^2}{2} \quad g(x; \overset{p}{A, x_c, L}) = \begin{cases} A \sin^2\left(\frac{\pi}{L}(x - x_c) + \frac{\pi}{2}\right) & x \in (x_c - \frac{L}{2}, x_c + \frac{L}{2}) \\ 0 & \text{otherwise.} \end{cases}$$



Example: the Burger's equation

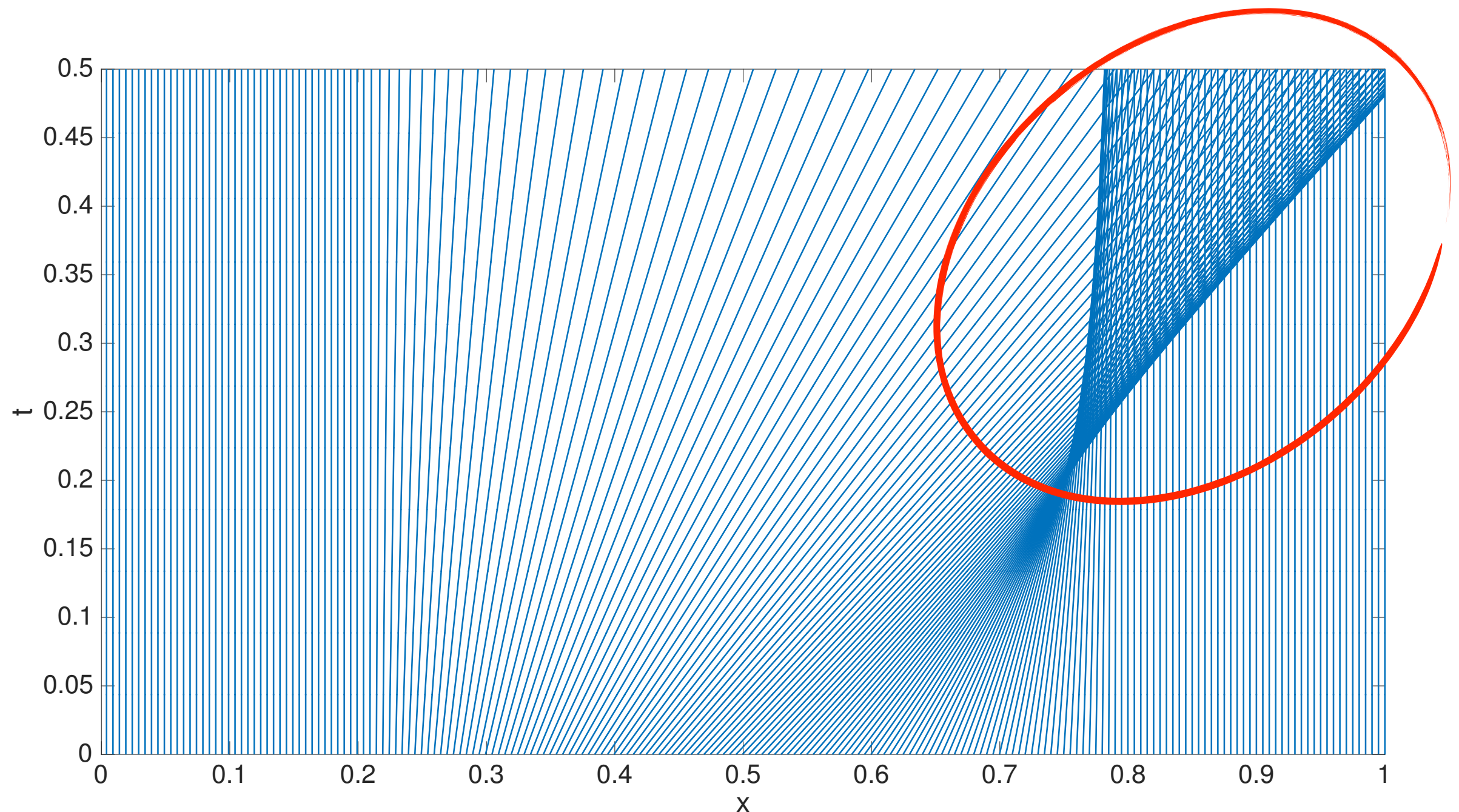
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Solution of the state equations

The Burger's equation can be rewritten as: $\partial_t u + u \partial_x u = 0$

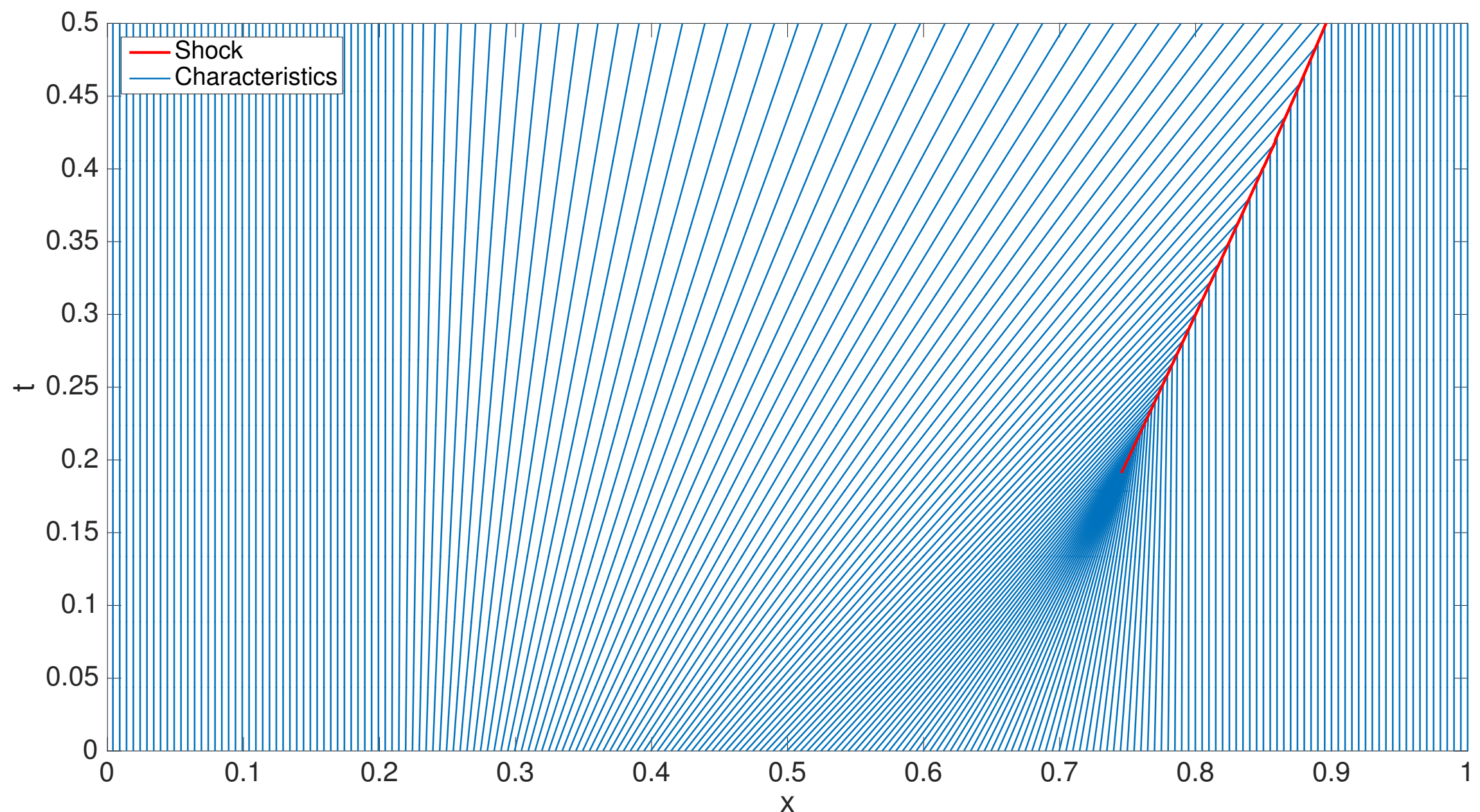
speed at which the initial condition is transported



Solution of the state equations

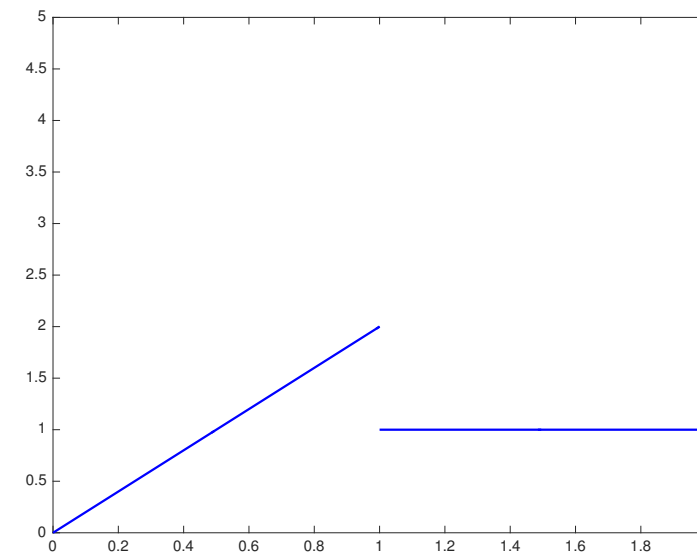
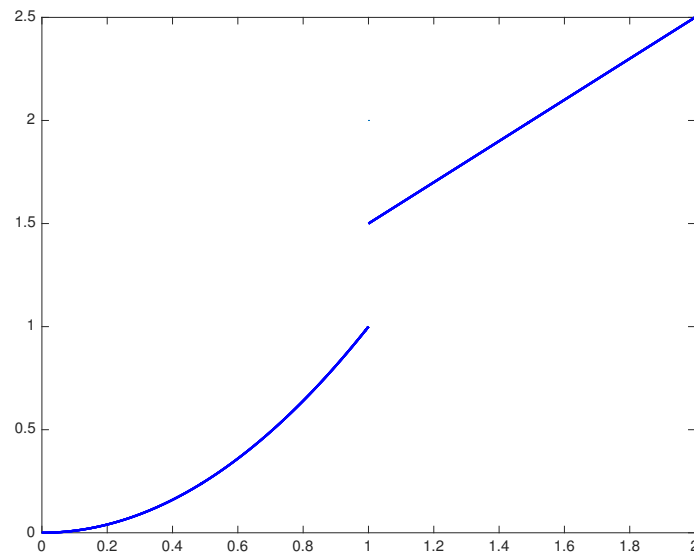
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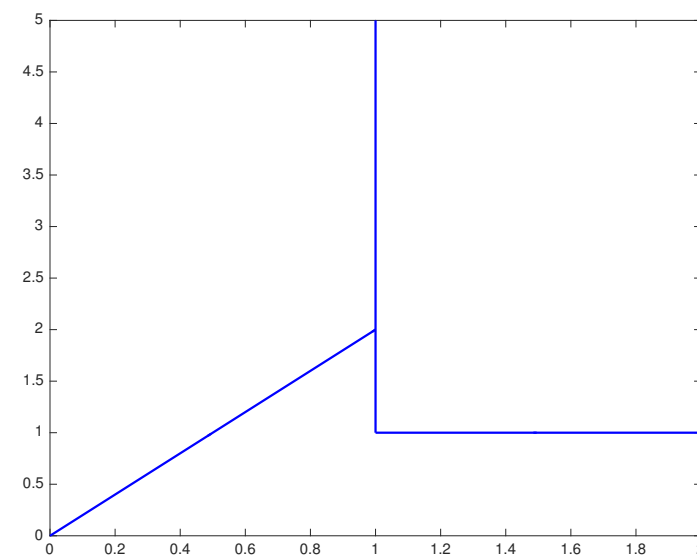
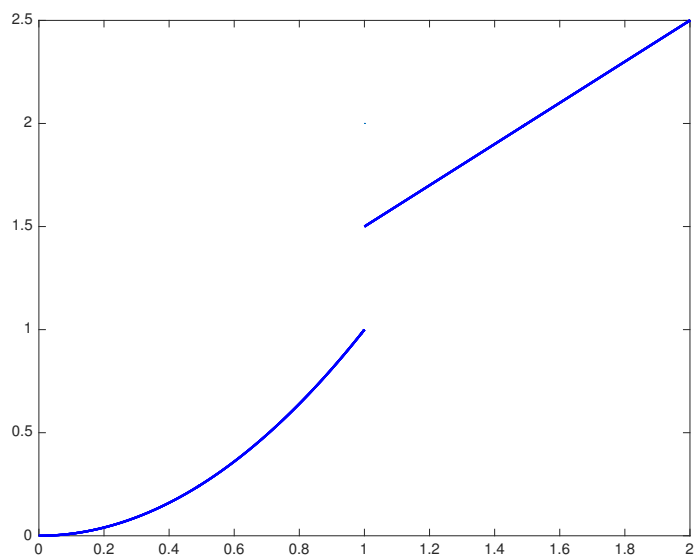


Two kinds of derivatives

- **Classical** derivative: it is defined everywhere but in the discontinuity



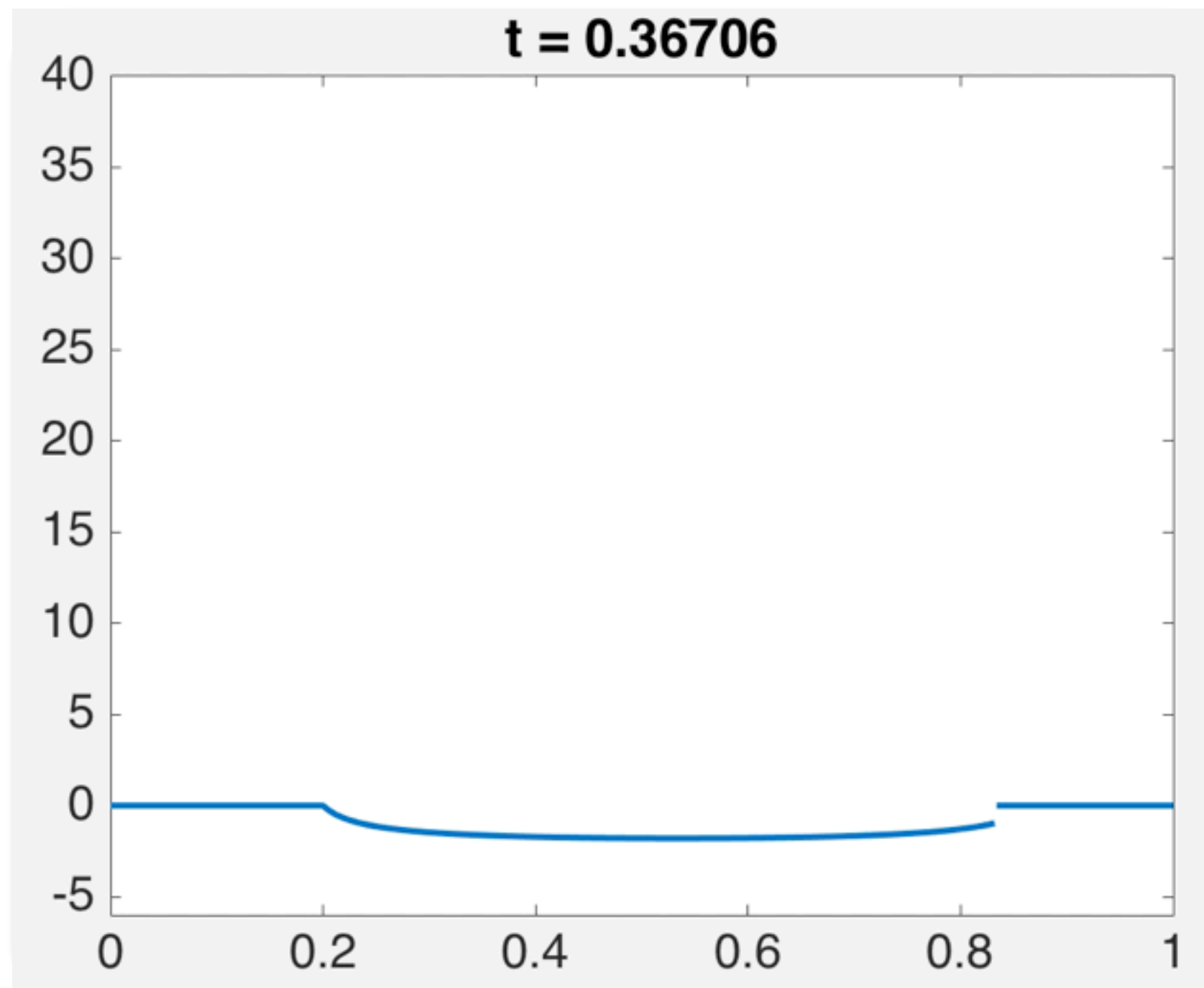
- **Weak** derivative: it is defined also in the discontinuity, where it is a **Dirac's distribution**



Example: the Burger's equation

$$f(u) = \frac{u^2}{2}$$

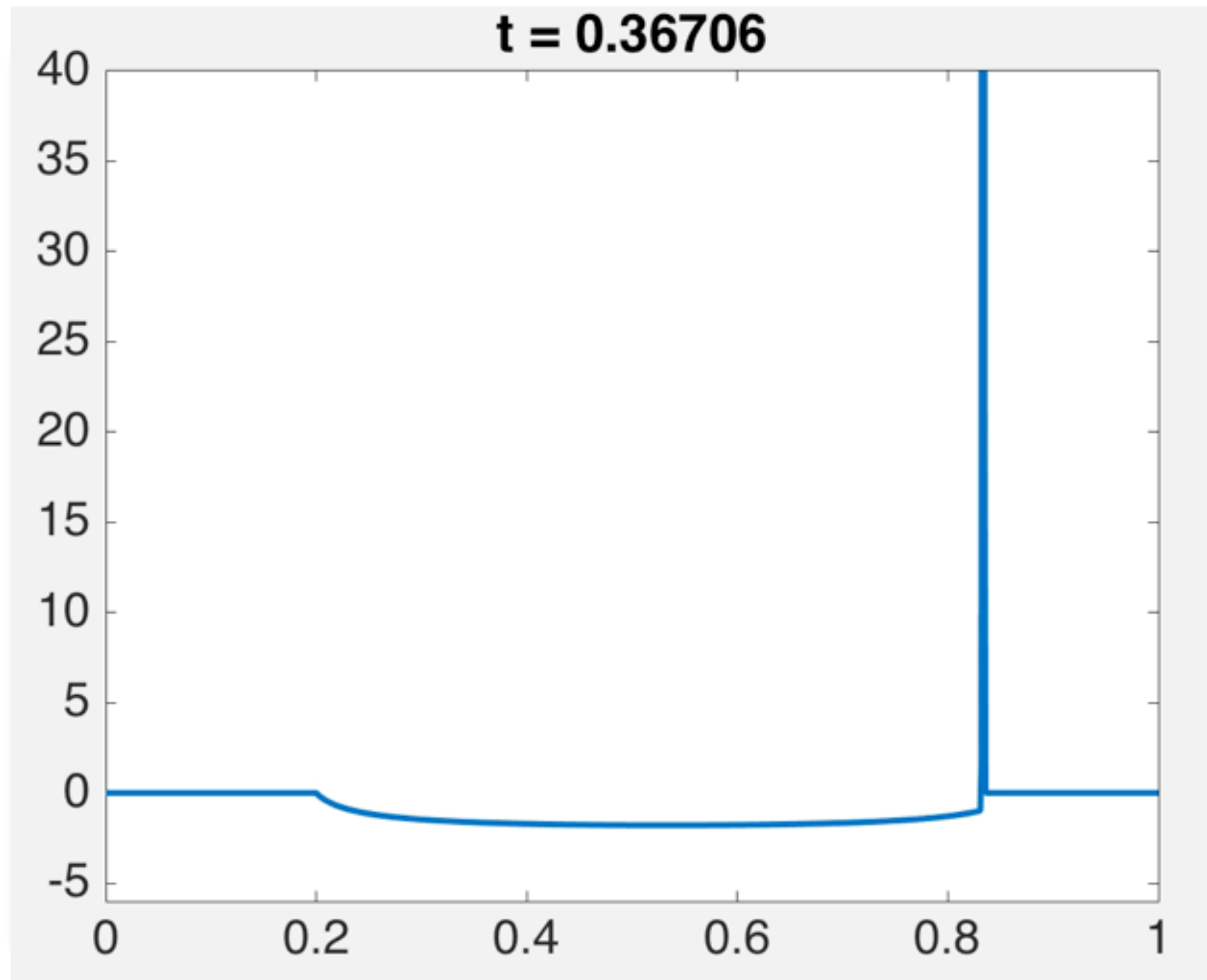
$$g(x; A, x_c, L) = \begin{cases} A \sin^2\left(\frac{\pi}{L}\left(x - \overset{p}{\textcircled{x_c}} + \frac{\pi}{2}\right)\right) & x \in \left(x_c - \frac{L}{2}, x_c + \frac{L}{2}\right) \\ 0 & \text{otherwise.} \end{cases}$$



Example: the Burger's equation

$$f(u) = \frac{u^2}{2}$$

$$g(x; A, x_c, L) = \begin{cases} A \sin^2\left(\frac{\pi}{L}\left(x - \overset{p}{\textcircled{x_c}} + \frac{\pi}{2}\right)\right) & x \in \left(x_c - \frac{L}{2}, x_c + \frac{L}{2}\right) \\ 0 & \text{otherwise.} \end{cases}$$



Choice of the derivative

Weak derivative:

- ▶ no correction to numerical schemes needed

Classical derivative:

- ▶ it does not corrupt the solution in the regular zones
- ▶ it is possible to estimate close solutions

Rankine-Hugoniot conditions

Across the shock, the **state** is governed by the Rankine-Hugoniot conditions:

$$\begin{cases} \partial_t \mathbf{u} + \partial_x f(\mathbf{u}) = 0 & x \in \mathbb{R}, t > 0 \\ \mathbf{u}(x, 0) = \mathbf{g}(x; \mathbf{p}) & x \in \mathbb{R}. \end{cases} \quad f(\mathbf{u}^+) - f(\mathbf{u}^-) = \sigma(\mathbf{u}^+ - \mathbf{u}^-)$$

If we wrote the same conditions for the **sensitivity**, we would have:

$$\begin{cases} \partial_t \mathbf{u}_{\mathbf{p}} + \partial_x (f'(\mathbf{u}) \mathbf{u}_{\mathbf{p}}) = 0 & x \in \mathbb{R}, t > 0 \\ \mathbf{u}_{\mathbf{p}}(x, 0) = \mathbf{g}_{\mathbf{p}}(x; \mathbf{p}) & x \in \mathbb{R}. \end{cases} \quad f'(\mathbf{u}^+) \mathbf{u}_{\mathbf{p}}^+ - f'(\mathbf{u}^-) \mathbf{u}_{\mathbf{p}}^- = \sigma(\mathbf{u}_{\mathbf{p}}^+ - \mathbf{u}_{\mathbf{p}}^-)$$

However, differentiating with respect to \mathbf{p} the conditions for the state we obtain:

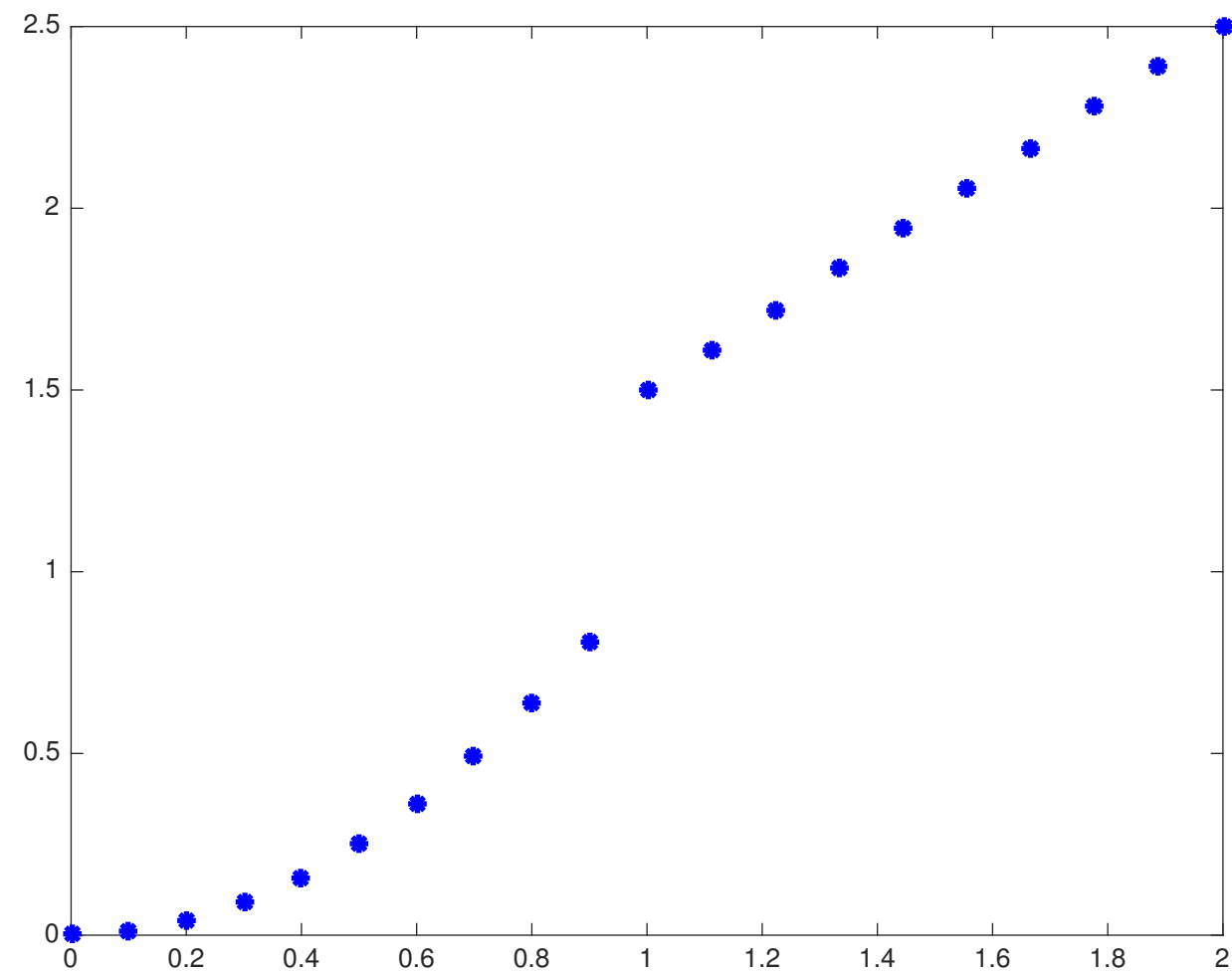
$$f'(\mathbf{u}^+) \mathbf{u}_{\mathbf{p}}^+ - f'(\mathbf{u}^-) \mathbf{u}_{\mathbf{p}}^- = \sigma(\mathbf{u}_{\mathbf{p}}^+ - \mathbf{u}_{\mathbf{p}}^-) + \partial_{\mathbf{p}} \sigma(\mathbf{u}^+ - \mathbf{u}^-)$$

Idea: add to the sensitivity equation a **source term** that balances it out.

$$\partial_t \mathbf{u}_{\mathbf{p}} + \partial_x (f'(\mathbf{u}) \mathbf{u}_{\mathbf{p}}) = \mathbf{s}(\mathbf{u}^+, \mathbf{u}^-) \quad x \in \mathbb{R}, t > 0$$

Shock detection

The term $\partial_{\mathbf{p}}\sigma(\mathbf{u}^+ - \mathbf{u}^-)$ is zero in the regular zones, however this is not true if we consider a **discretisation** of the equations.

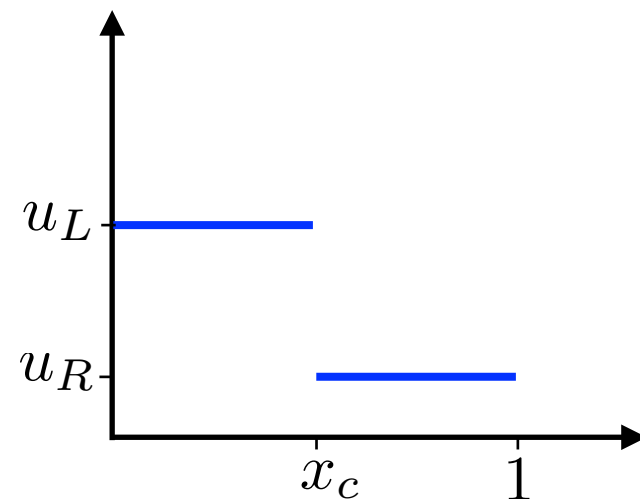


It is necessary to define a **shock detector**.

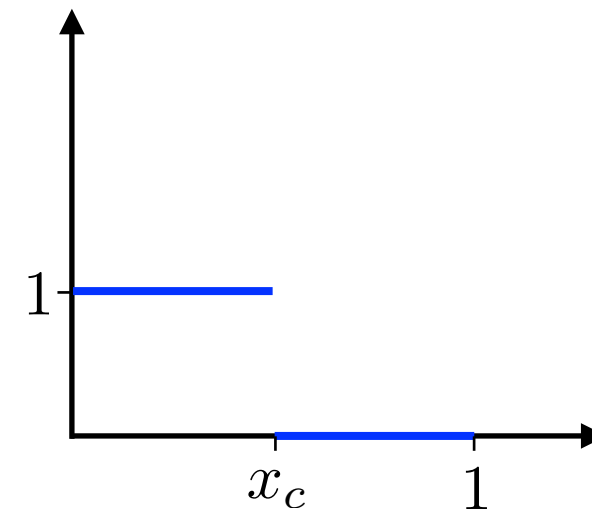
Numerical results

The **Riemann problem** for the Burger's equation:

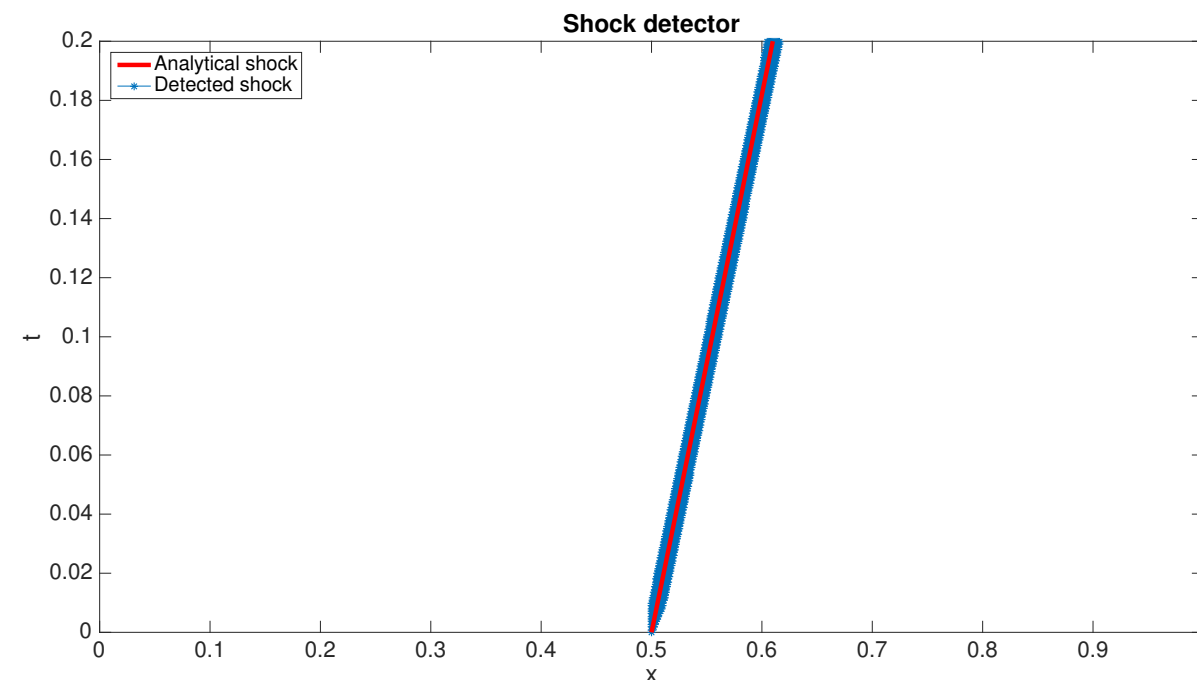
$$g(x; u_R, u_L, x_c)$$



$$g_{u_L}(x; u_R, u_L, x_c)$$

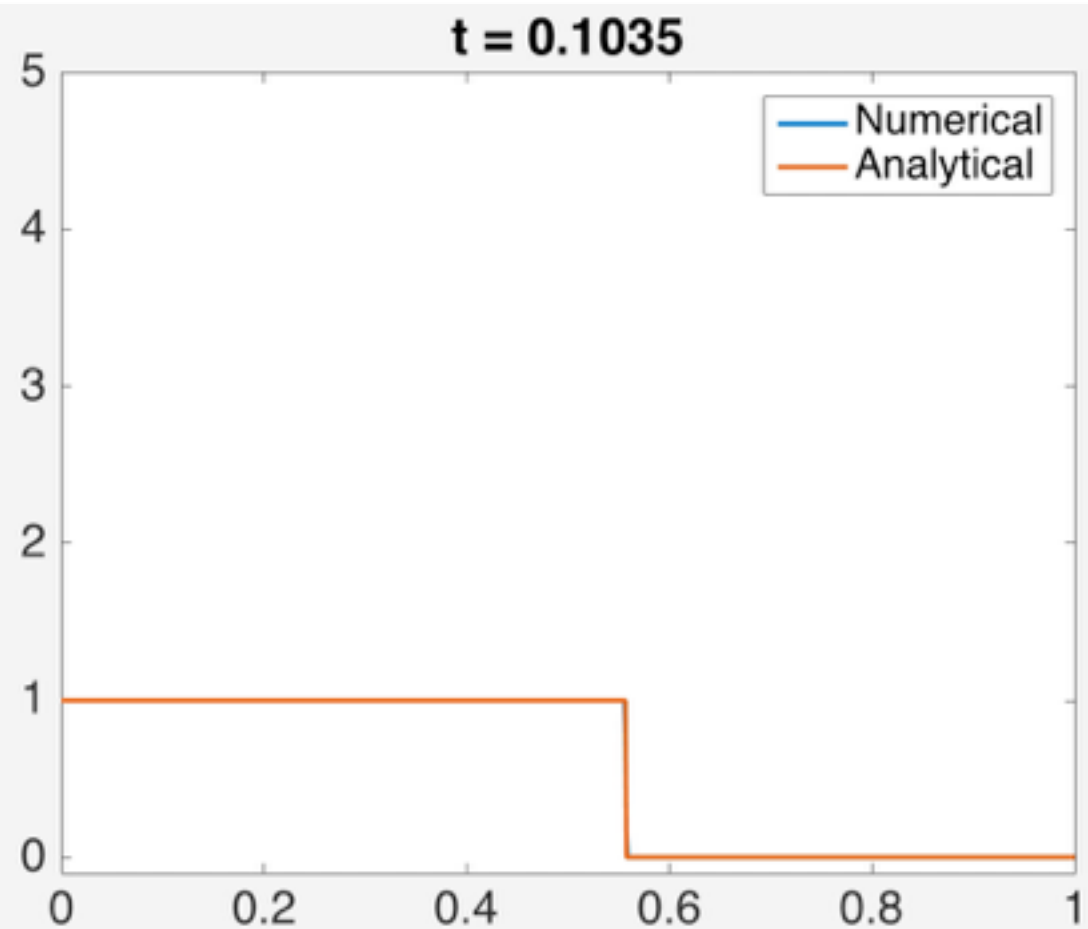


In this case it is easy to define a good shock detector.

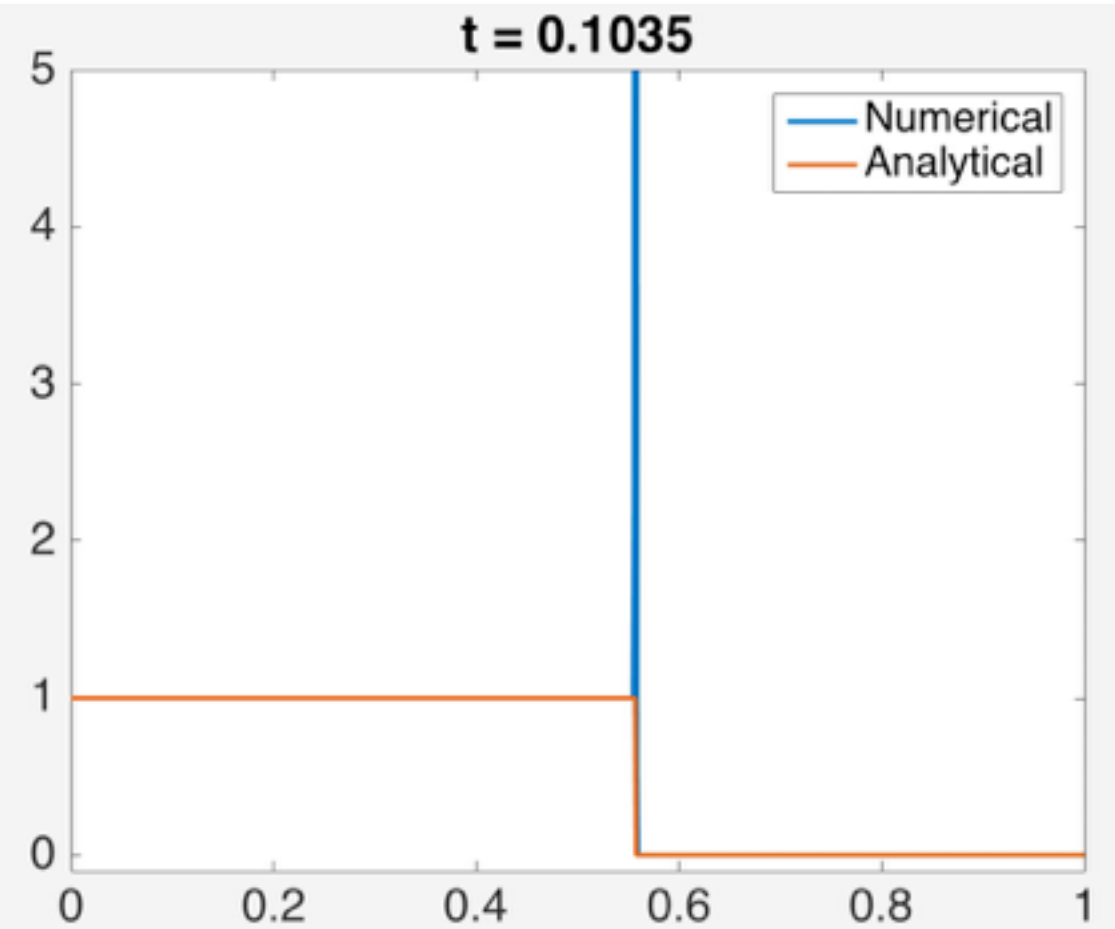


Numerical results

Sensitivity **with** source term:



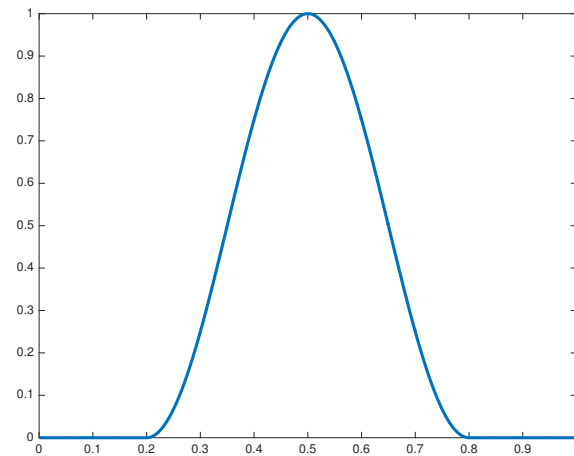
Sensitivity **without** source term:



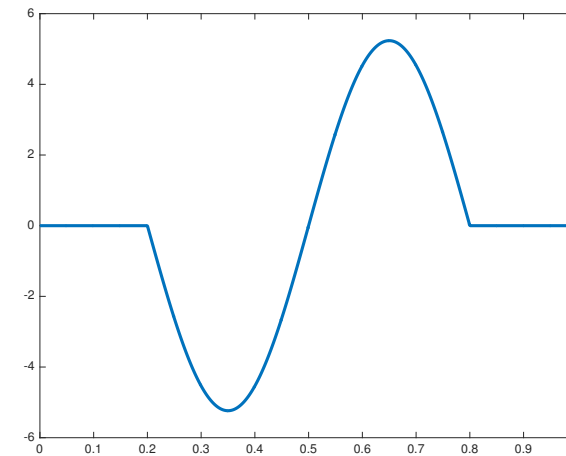
Numerical results

The same shock detector in a less simple case does not work:

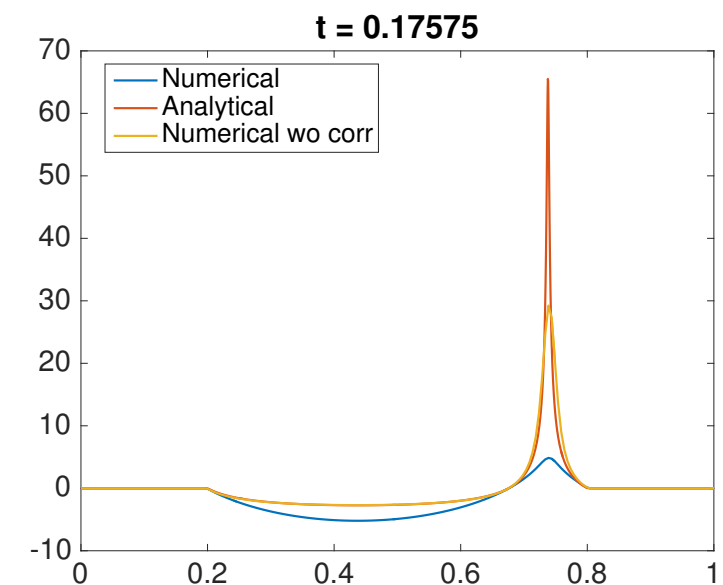
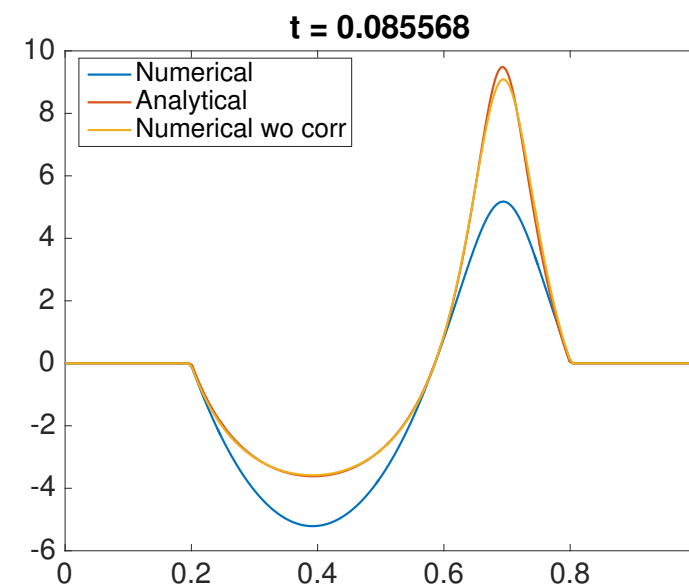
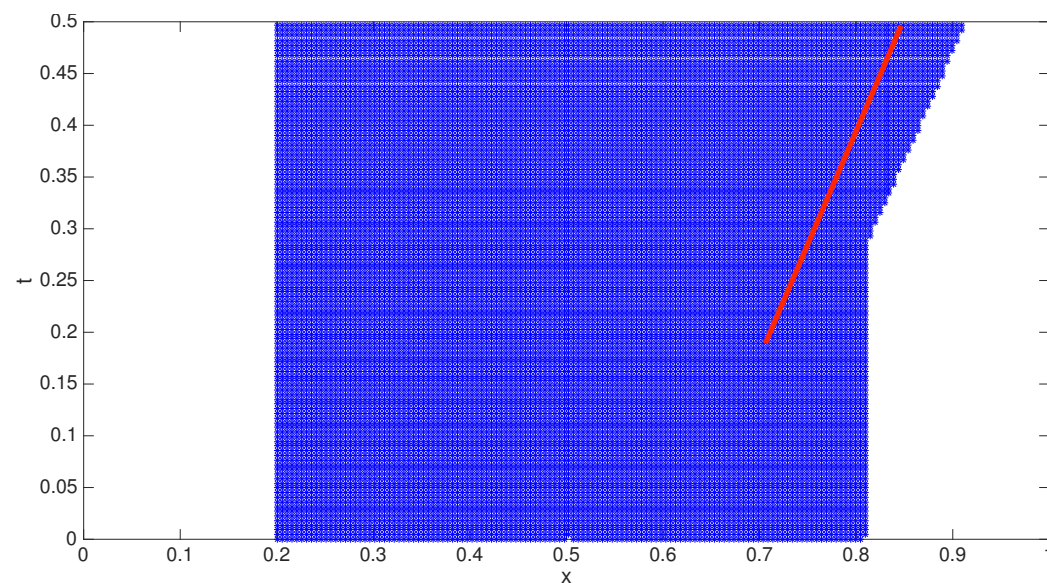
$$g(x; \mathbf{p})$$



$$g_{\mathbf{p}}(x; \mathbf{p})$$

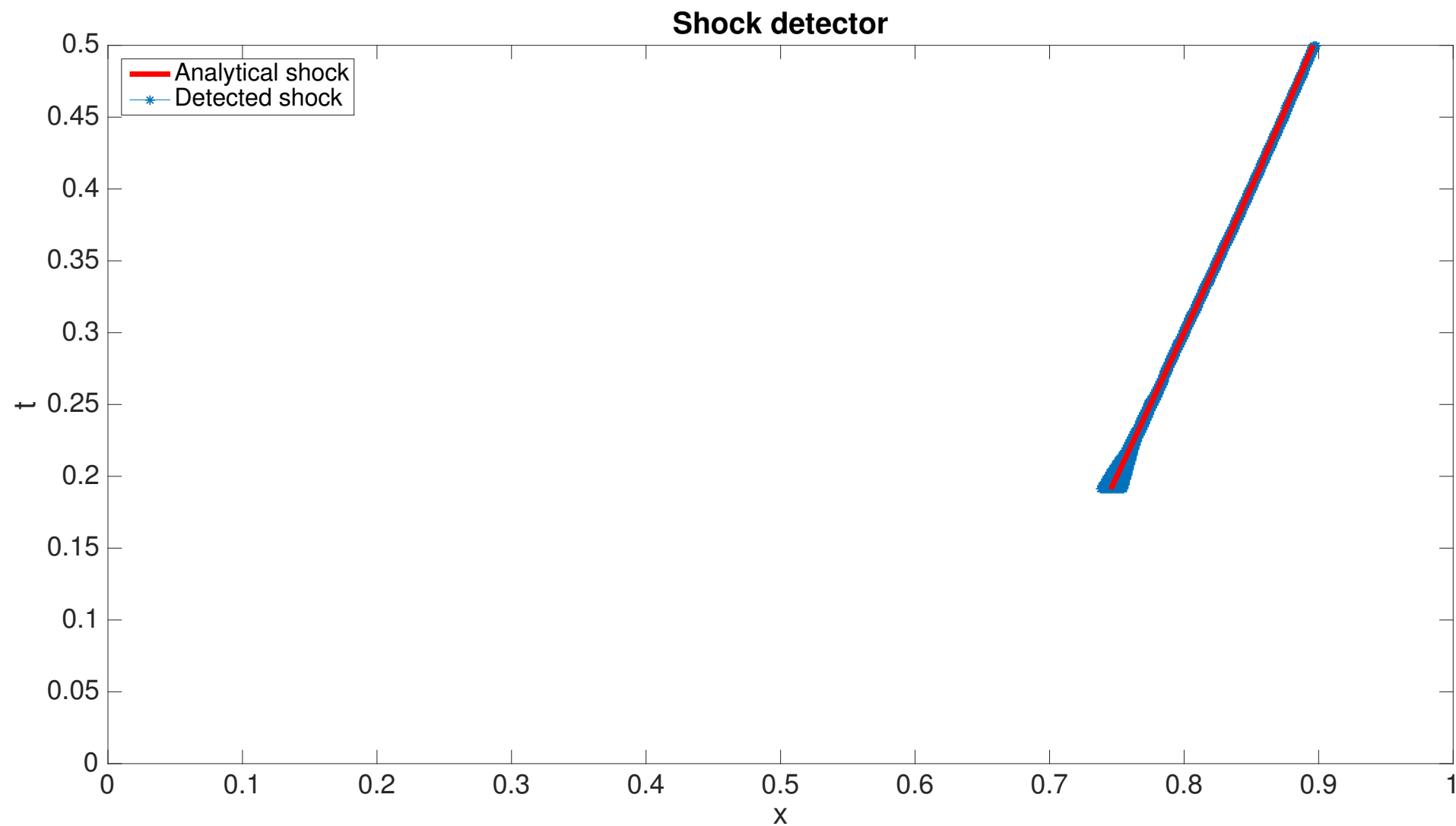


It leads to an **overcorrection**:



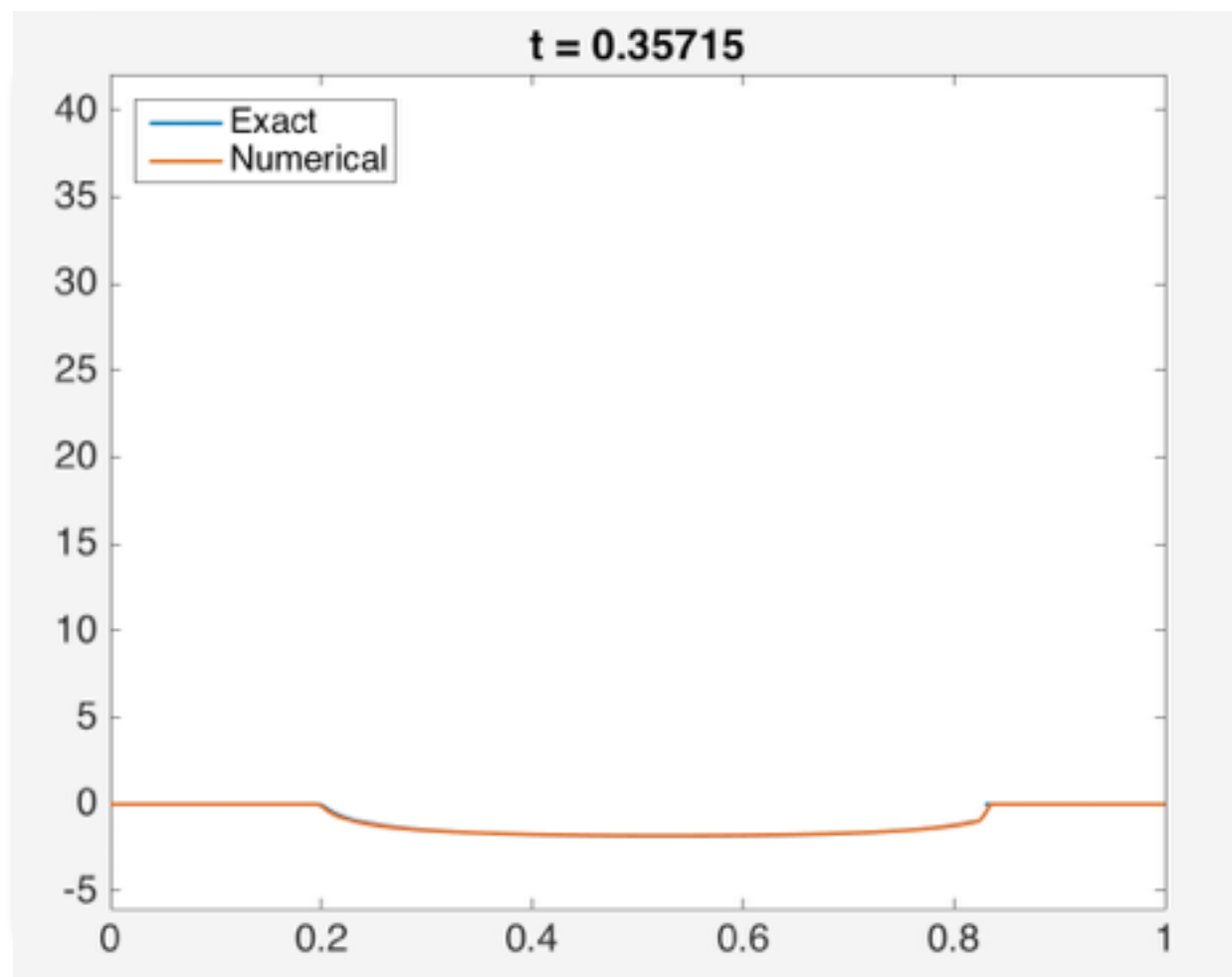
Numerical results

We defined a new shock detector based on the **second derivative** and on the **breaking time**.

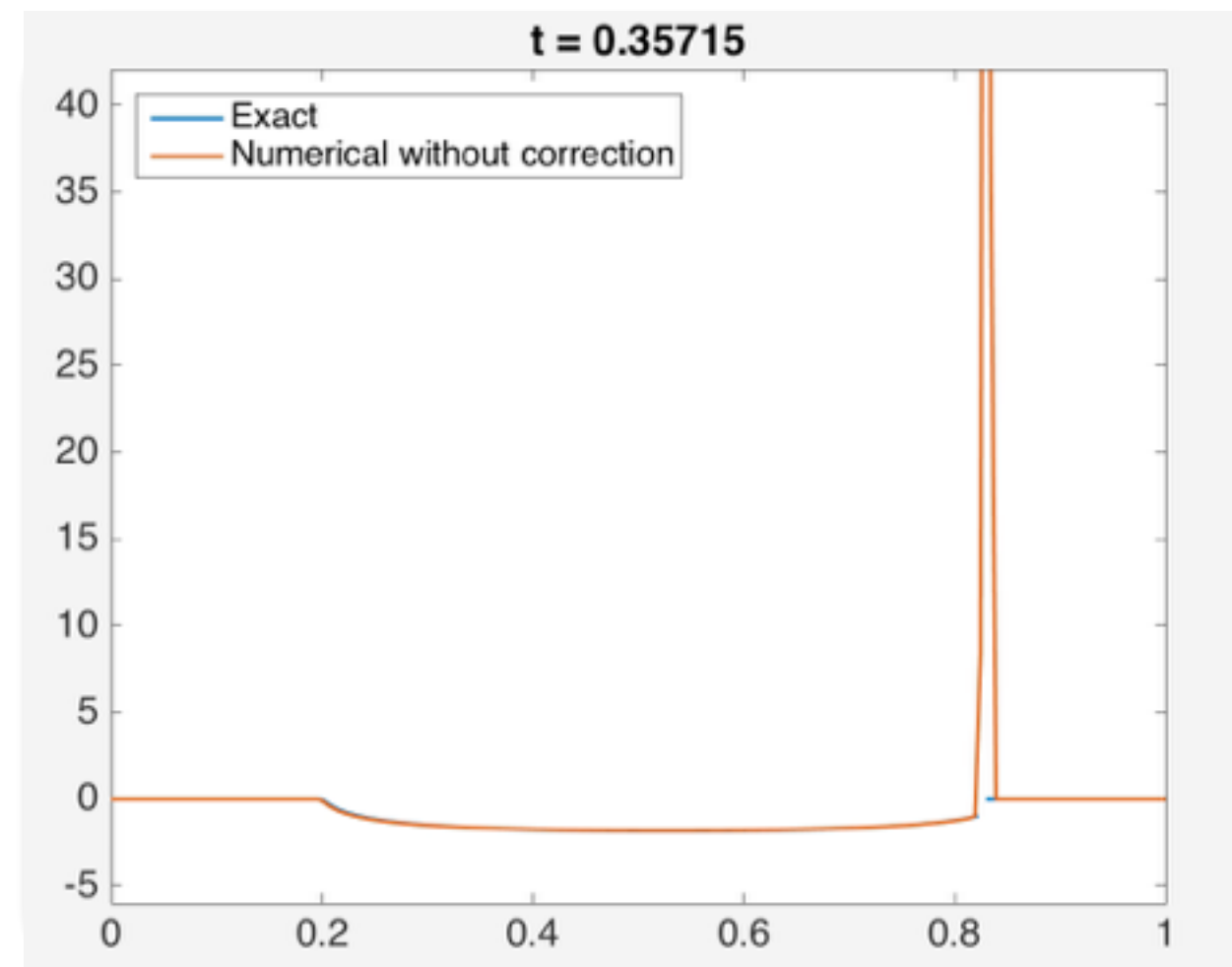


Numerical results

Sensitivity **with** source term:



Sensitivity **without** source term:



Conclusion and future developments

- ▶ A general method for sensitivity analysis in case of discontinuities has been developed ;
- ▶ Shock detectors are specific to each case;
- ▶ The method has been extended to systems (Euler 1D);
- ▶ We plan to increase the space dimension (2D or 3D).

Thank you
for your attention!