

# Sensitivity analysis for hyperbolic PDEs systems with discontinuous solution

**LMV**  
Laboratoire de mathématiques  
de Versailles - CNRS UMR 8100



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**École de recherche en mathématiques pour l'énergie nucléaire**

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# Outline of the talk

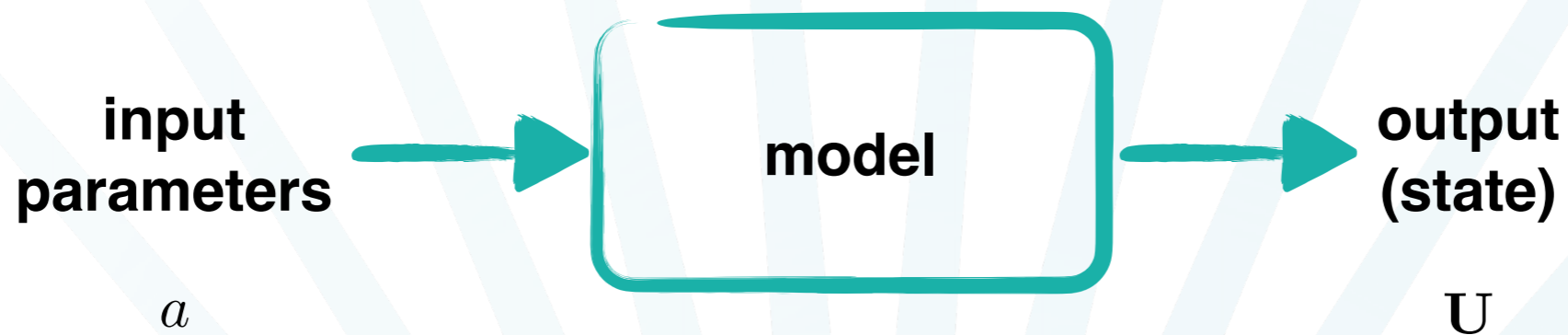
- ▶ Sensitivity analysis
- ▶ Sensitivity analysis for hyperbolic equations
- ▶ Riemann problem for the Euler equations and their sensitivity
- ▶ Classical numerical schemes
- ▶ Anti-diffusive numerical schemes
- ▶ Numerical results
- ▶ Uncertainty quantification



# **Sensitivity Analysis**

# Sensitivity Analysis

Sensitivity analysis: study of how changes in the **inputs** of a model affect the **outputs**



**Sensitivity:**  $\frac{\partial U}{\partial a} = U_a$

# Standard techniques of sensitivity analysis

Standard techniques of sensitivity analysis call for the differentiation of the state system:

$$\begin{cases} \partial_t \mathbf{U} + \partial_x \mathbf{F}(\mathbf{U}) = 0 & \Omega \times (0, T), \\ \mathbf{U}(x, 0) = \mathbf{g}(x) & \Omega, \end{cases}$$

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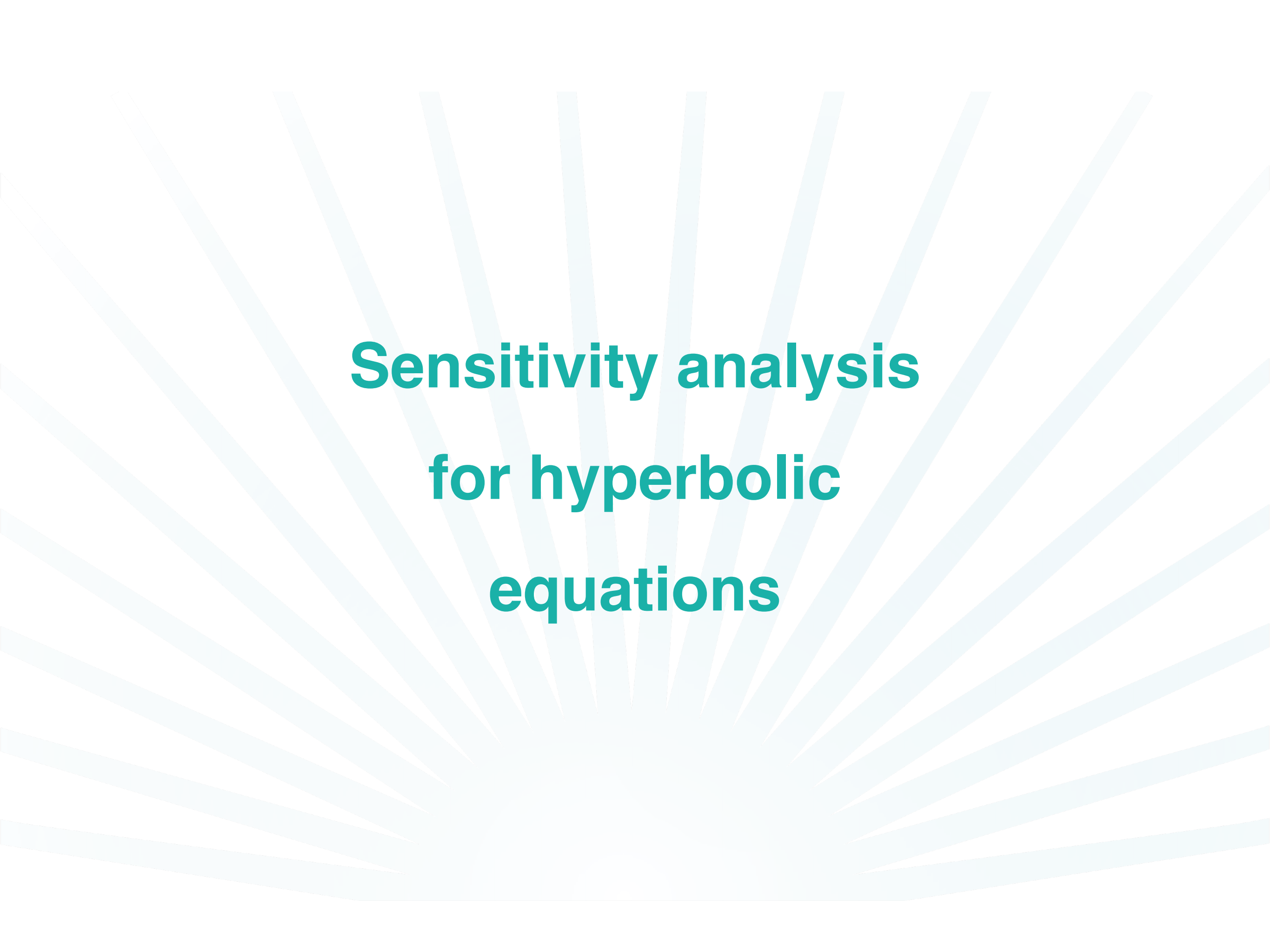
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This can be done under **hypotheses of regularity** of the state  $\mathbf{U}$ .

If these techniques are applied to hyperbolic equations, **Dirac delta functions** will appear in the sensitivity.



**Sensitivity analysis  
for hyperbolic  
equations**

# Definition of the source term

In order to have a sensitivity system which is valid also when the state is discontinuous, we add a correction term:

$$\begin{cases} \partial_t \mathbf{U}_a + \partial_x \mathbf{F}_a(\mathbf{U}, \mathbf{U}_a) = \mathbf{S} & \Omega \times (0, T), \\ \mathbf{U}_a(x, 0) = \mathbf{g}_a(x) & \Omega, \end{cases}$$

defined as follows:

$$\mathbf{S} = \sum_{k=1}^{N_s} \rho_k \delta(x - x_{k,s}(t)),$$

number of discontinuities

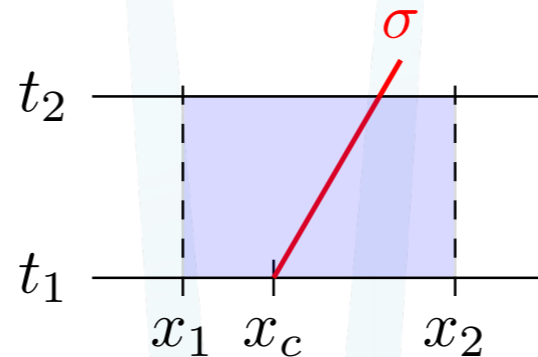
position of the k-th discontinuity

amplitude of the k-th correction  
(to be computed)

**Remark:** a **shock detector** is necessary to discretise such source term.

# Definition of the source term

To compute the amplitude of the correction, we consider an infinitesimal control volume containing a single discontinuity:



By integrating the sensitivity equations with the source term on the control volume, one has:

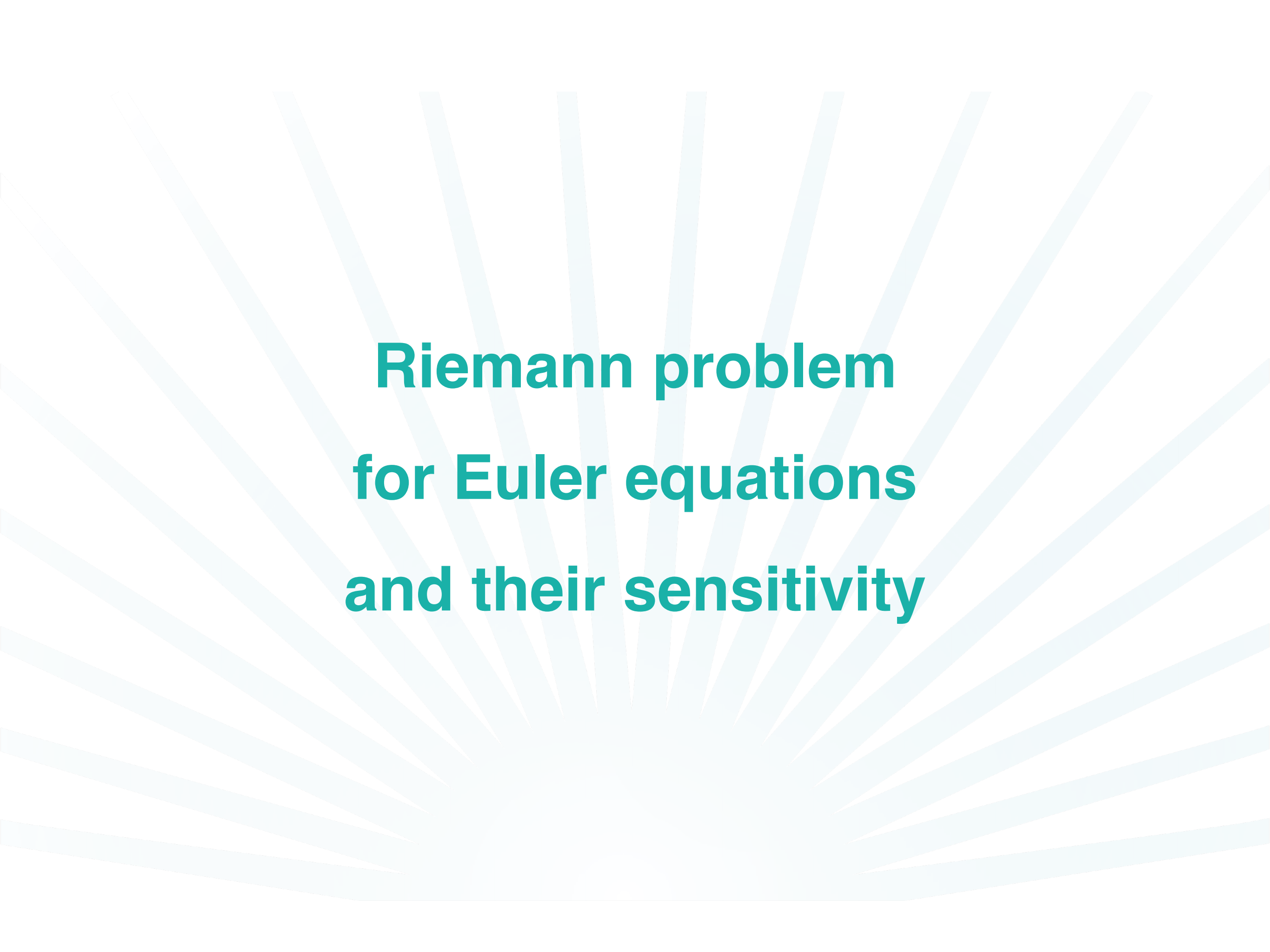
$$\rho_k = (\mathbf{U}_a^- - \mathbf{U}_a^+) \sigma_k + \mathbf{F}_a^+ - \mathbf{F}_a^-$$

Rankine-Hugoniot conditions for the state:  $(\mathbf{U}^+ - \mathbf{U}^-) \sigma_k = \mathbf{F}^+ - \mathbf{F}^-$

Differentiating them w.r.t. the parameter:

$$\begin{aligned} & (\mathbf{U}_a^- - \mathbf{U}_a^+) \sigma_k + (\mathbf{U}^- - \mathbf{U}^+) \sigma_{k,a} + \sigma_k (\partial_x \mathbf{U}^+ - \partial_x \mathbf{U}^-) \partial_a x_{k,s}(t) = \\ & = \mathbf{F}_a^- - \mathbf{F}_a^+ + \left( \frac{\partial \mathbf{F}(\mathbf{U}^+)}{\partial \mathbf{U}} \partial_x \mathbf{U}^+ - \frac{\partial \mathbf{F}(\mathbf{U}^-)}{\partial \mathbf{U}} \partial_x \mathbf{U}^- \right) \partial_a x_{k,s}(t). \end{aligned}$$

Finally, we obtain the following amplitude:  $\rho_k = (\mathbf{U}^+ - \mathbf{U}^-) \sigma_{a,k}$



**Riemann problem  
for Euler equations  
and their sensitivity**

# The Riemann problem for Euler equations

The Euler equations write:

$$\begin{cases} \partial_t \rho + \partial_x(\rho u) = 0, \\ \partial_t(\rho u) + \partial_x(\rho u^2 + p) = 0, \\ \partial_t(\rho E) + \partial_x(u(\rho E + p)) = 0, \end{cases}$$

Eigenvalues:

$$\lambda_1(\mathbf{U}) = u - c,$$

$$\lambda_2(\mathbf{U}) = u,$$

$$\lambda_3(\mathbf{U}) = u + c.$$

Eigenvectors:

$$\mathbf{r}_1(\mathbf{U}) = (1, u - c, H - uc)^t,$$

$$\mathbf{r}_2(\mathbf{U}) = (1, u, \frac{u^2}{2})^t,$$

$$\mathbf{r}_3(\mathbf{U}) = (1, u + c, H + uc)^t.$$

**Genuinely nonlinear**

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**Linearly degenerate**

# The Riemann problem for Euler equations

The Euler equations write:

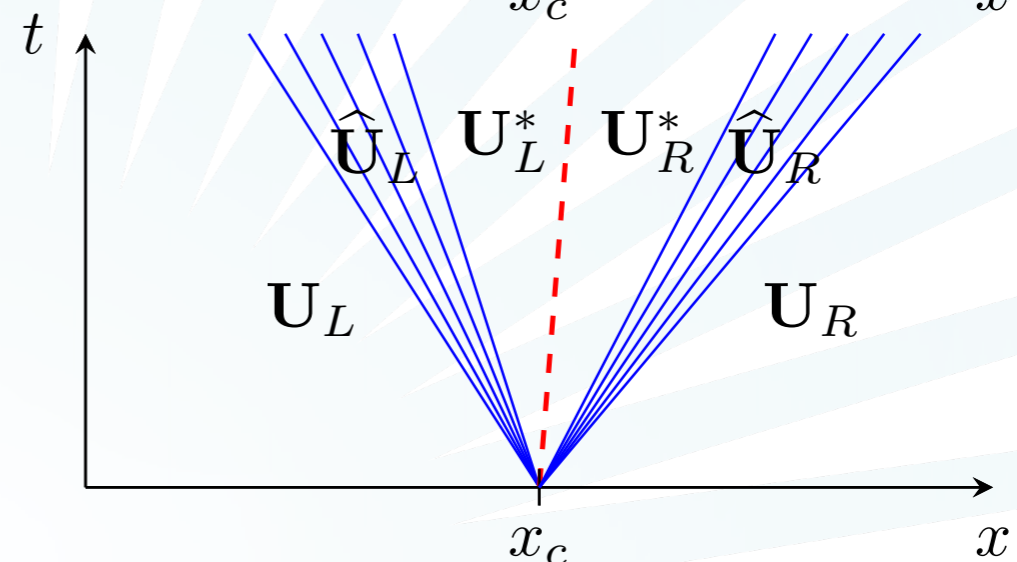
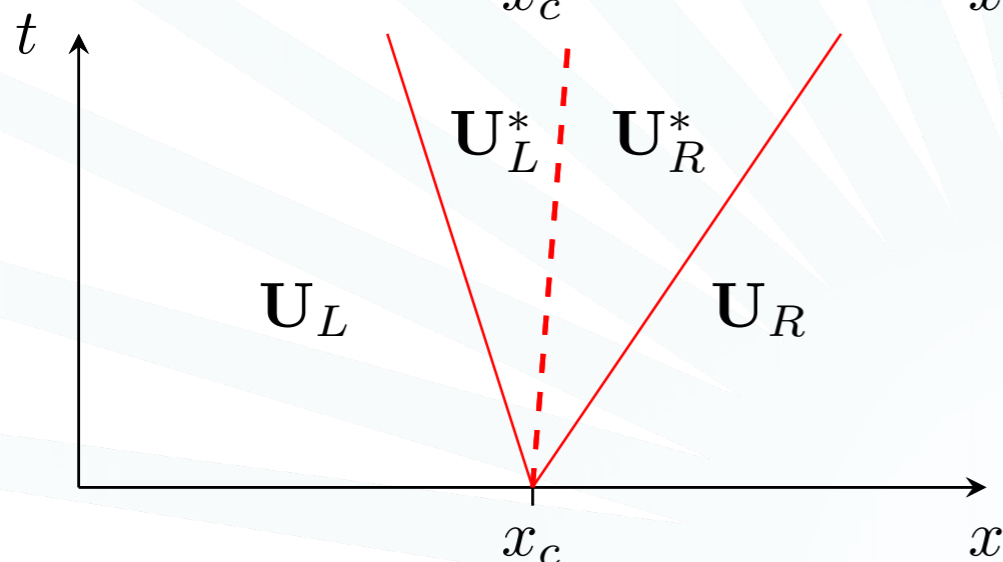
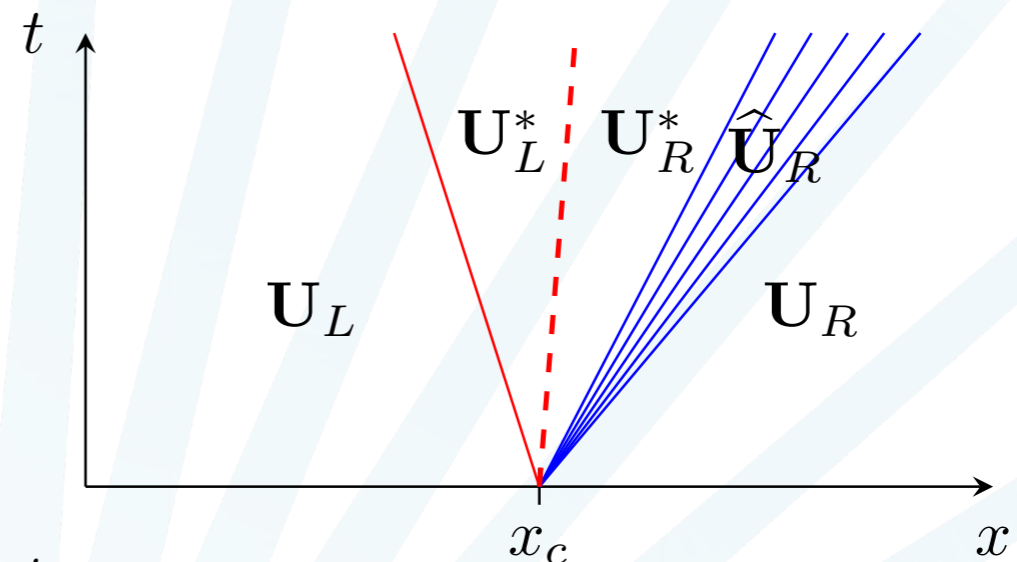
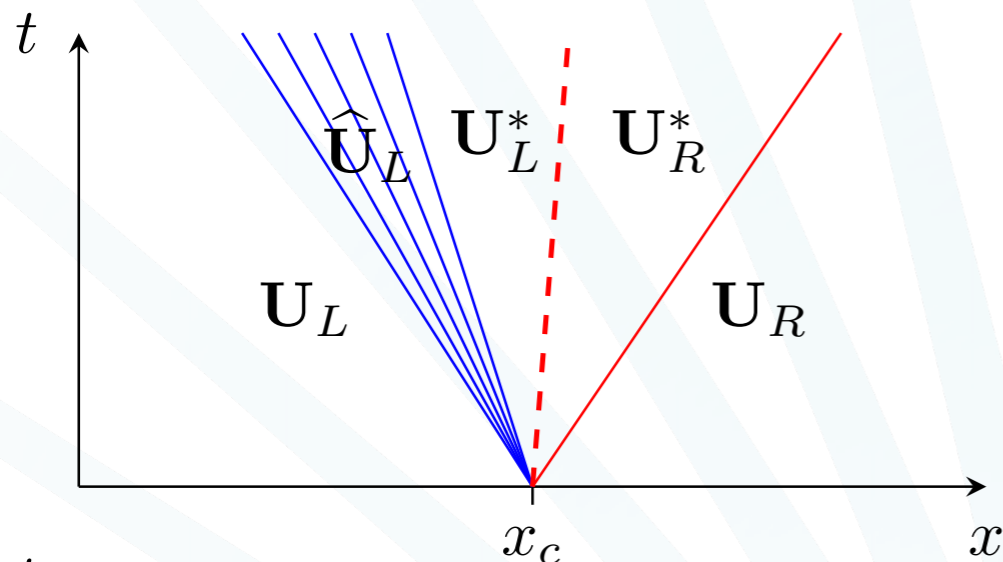
$$\begin{cases} \partial_t \rho + \partial_x(\rho u) = 0, \\ \partial_t(\rho u) + \partial_x(\rho u^2 + p) = 0, \\ \partial_t(\rho E) + \partial_x(u(\rho E + p)) = 0, \end{cases}$$

Eigenvalues:

$$\begin{aligned} \lambda_1(\mathbf{U}) &= u - c, \\ \lambda_2(\mathbf{U}) &= u, \\ \lambda_3(\mathbf{U}) &= u + c. \end{aligned}$$

Eigenvectors:

$$\begin{aligned} \mathbf{r}_1(\mathbf{U}) &= (1, u - c, H - uc)^t, \\ \mathbf{r}_2(\mathbf{U}) &= (1, u, \frac{u^2}{2})^t, \\ \mathbf{r}_3(\mathbf{U}) &= (1, u + c, H + uc)^t. \end{aligned}$$





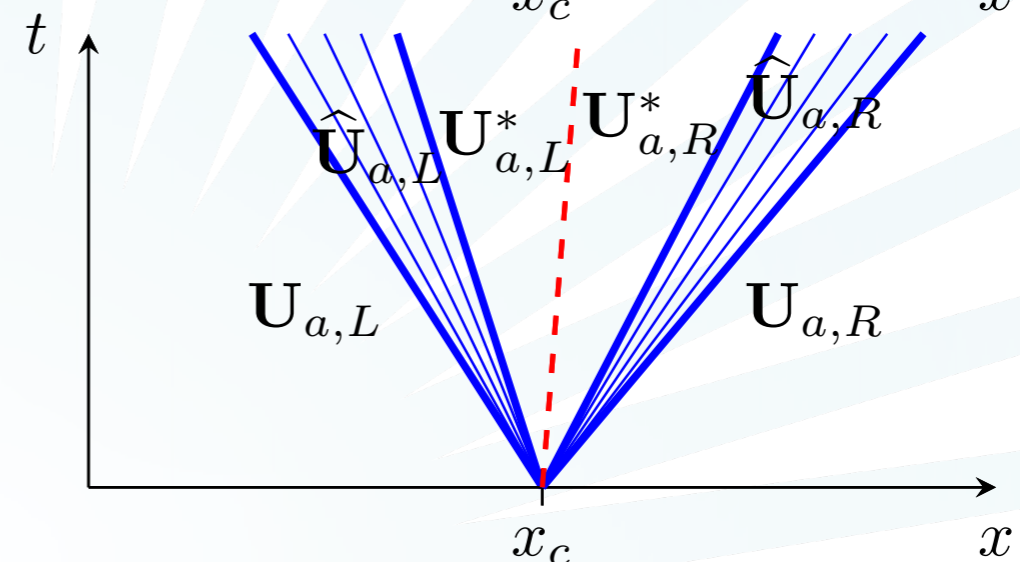
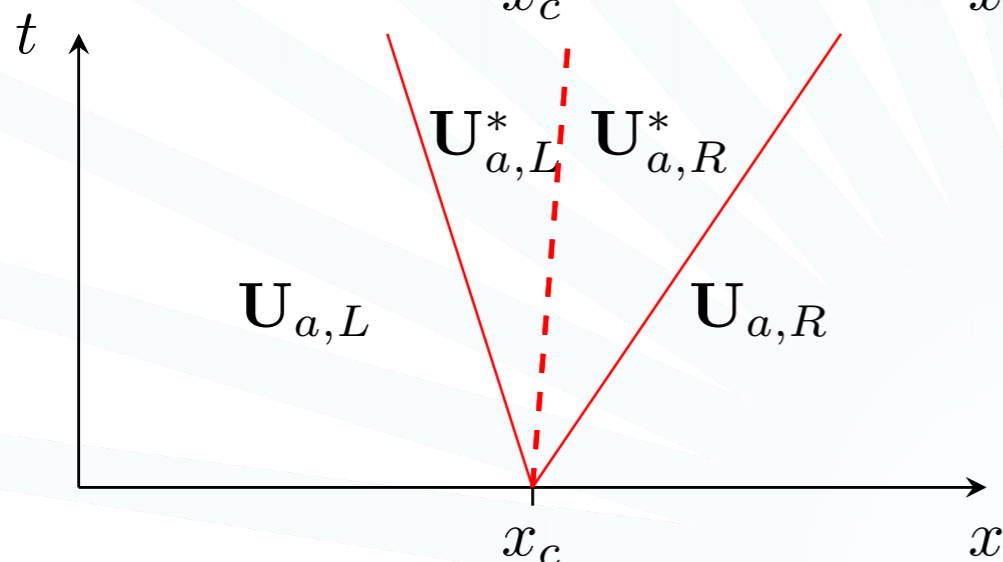
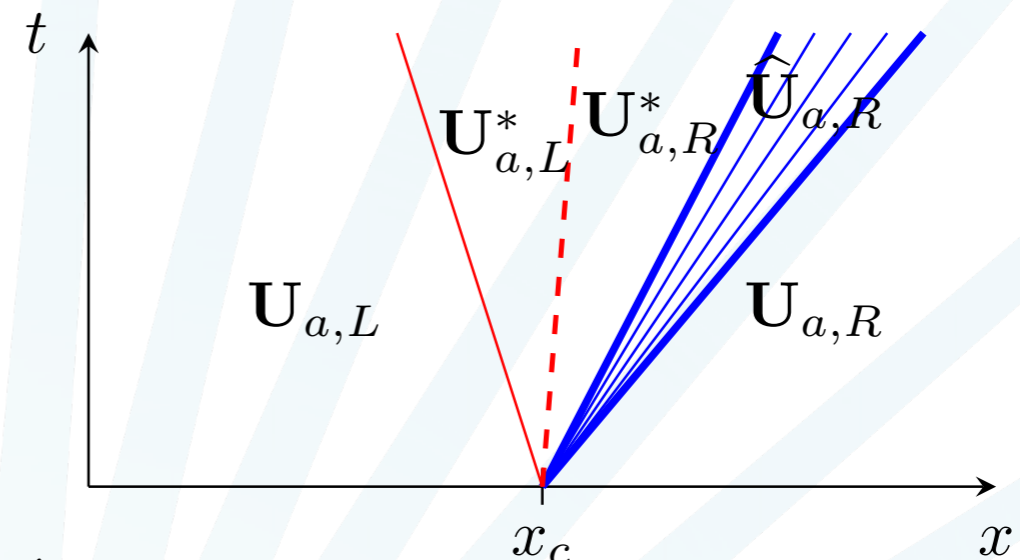
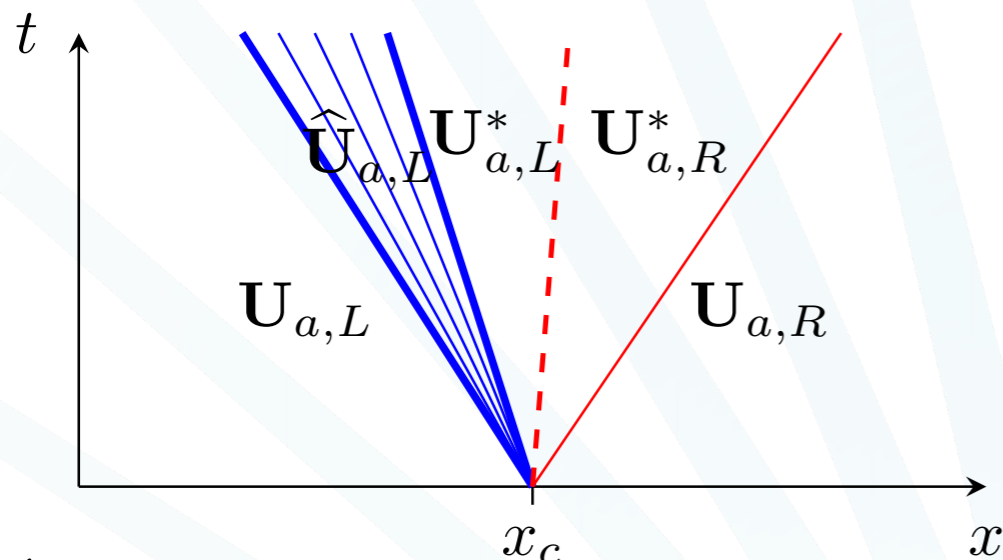
# The Riemann problem for the sensitivity equations

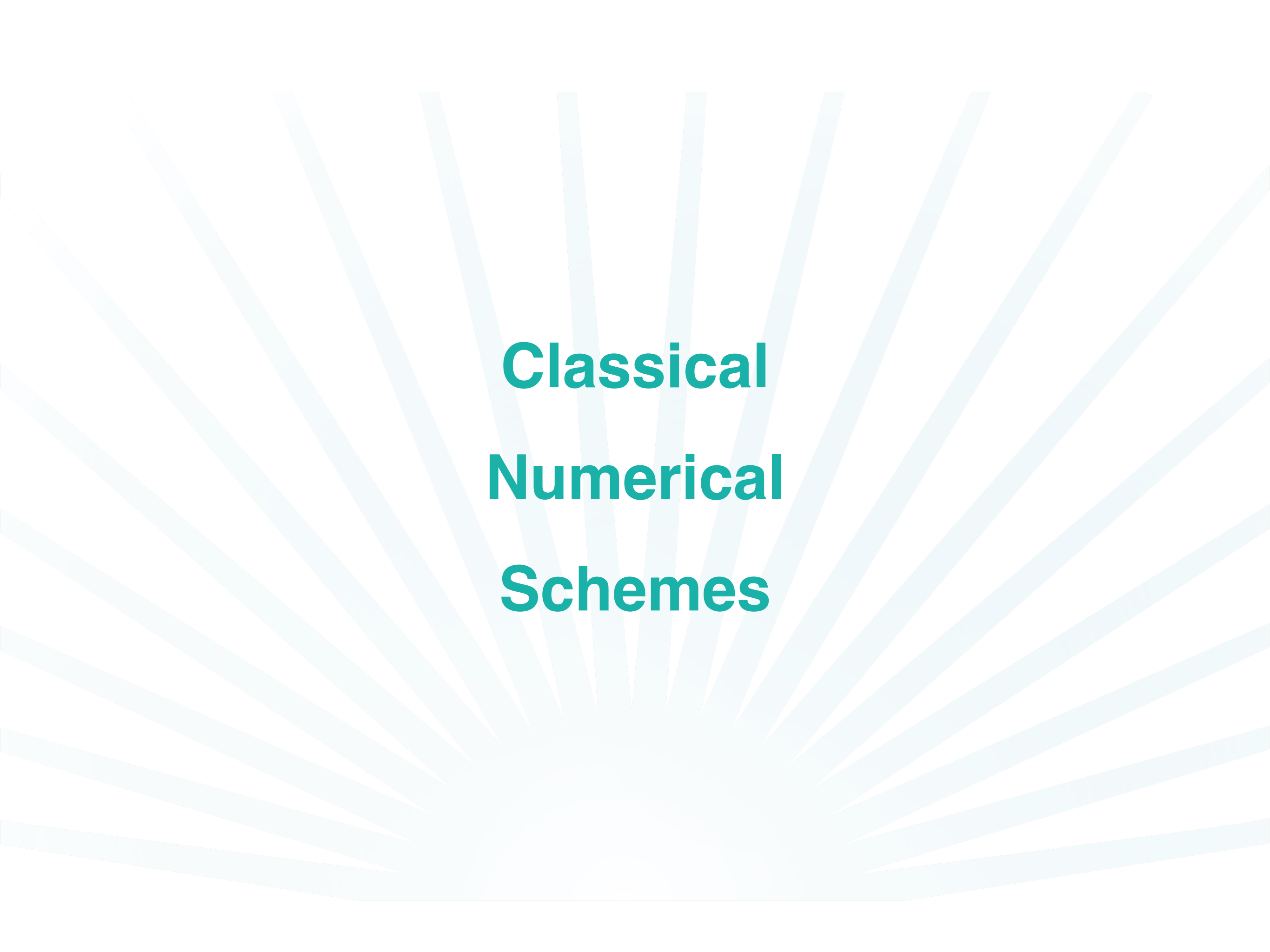
The sensitivity system writes:

$$\begin{cases} \partial_t \rho_a + \partial_x (\rho u)_a = S_1, \\ \partial_t (\rho u)_a + \partial_x (\rho_a u^2 + 2\rho u u_a + p_a) = S_2, \\ \partial_t (\rho E)_a + \partial_x (u_a (\rho E + p) + u((\rho E)_a + p_a)) = S_3, \end{cases}$$

Eigenvalues:

$$\begin{aligned} \lambda_1(\mathbf{U}) &= u - c, \\ \lambda_2(\mathbf{U}) &= u, \\ \lambda_3(\mathbf{U}) &= u + c. \end{aligned}$$





**Classical  
Numerical  
Schemes**

# Classical numerical schemes

## Godunov-type schemes

Step 0 : initial data discretisation

approximate Riemann solvers are used

Step 1 : solution of a Riemann problem for each interface  $x_{j-1/2}$  obtaining  $\mathbf{v}(x, t^{n+1})$

Step 2 : average 
$$\mathbf{v}_j^{n+1} = \frac{1}{\Delta x} \int_{x_{j-1/2}}^{x_{j+1/2}} \mathbf{v}(x, t^{n+1}) dx$$

**Remark:** the state and the sensitivity systems are solved **separately**.

# Classical numerical schemes

$$\begin{cases} \partial_t \mathbf{U} + \partial_x \mathbf{F}(\mathbf{U}) = 0 \\ \partial_t \mathbf{U}_a + \partial_x \mathbf{F}_a(\mathbf{U}, \mathbf{U}_a) = \mathbf{S}(\mathbf{U}) \end{cases} \quad \mathbf{S}(\mathbf{U}) = \sum_{k=1}^{N_s} \sigma_{a,k} (\mathbf{U}_k^+ - \mathbf{U}_k^-)$$

**Remark:** HLL-type schemes cannot be used for the state, two intermediate star states are necessary to have a well-defined the source term for the sensitivity.

## Approximate Riemann solver for the state

First order Roe-type scheme

$$\lambda_1^{ROE} = \tilde{u} - \tilde{c}_3 \quad \lambda_2^{ROE} = \tilde{u} \quad \lambda_3^{ROE} = \tilde{u} + \tilde{c} \quad \text{Roe-averaged eigenvalues}$$

$$\mathbf{U}_R - \mathbf{U}_L = \sum_{k=1}^3 \alpha_k \tilde{\mathbf{r}}_k \quad \text{decomposition along Roe-averaged eigenvectors}$$

$$\mathbf{U}_L^* = \mathbf{U}_L + \alpha_1 \tilde{\mathbf{r}}_1 \quad \mathbf{U}_R^* = \mathbf{U}_R - \alpha_3 \tilde{\mathbf{r}}_3$$

# Classical numerical schemes

## Approximate Riemann solvers for the sensitivity

► HLL-type scheme: simpler structure than the state solver.

HLL consistency conditions yield:

$$\mathbf{U}_{a,j-1/2}^* = \frac{1}{\lambda_3^{ROE} - \lambda_1^{ROE}} \left( \lambda_3^{ROE} \mathbf{U}_{a,j}^n - \lambda_1^{ROE} \mathbf{U}_{a,j-1}^n - \mathbf{F}_a(\mathbf{U}_j, \mathbf{U}_{a,j}) + \mathbf{F}_a(\mathbf{U}_{j-1}, \mathbf{U}_{a,j-1}) + \mathbf{S}_{j-1/2} \right)$$

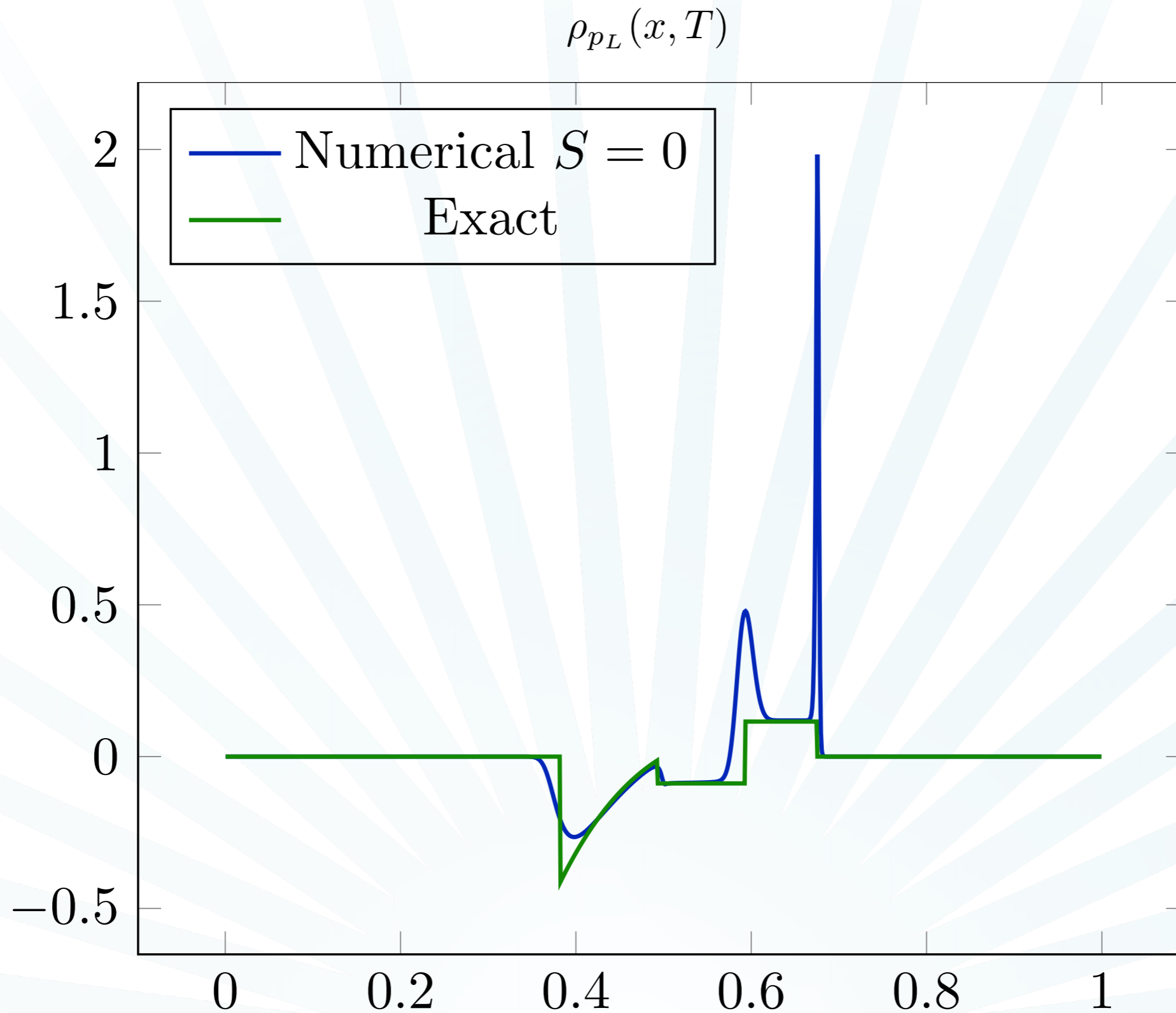
$$\begin{aligned} \mathbf{S}_{j-1/2} = & \partial_a \lambda_{1,j-1/2}^{ROE} (\mathbf{U}_{L,j-1/2}^* - \mathbf{U}_{j-1}) d_{1,j-1/2} \\ & + \partial_a \lambda_{2,j-1/2}^{ROE} (\mathbf{U}_{R,j-1/2}^* - \mathbf{U}_{L,j-1/2}^*) \\ & + \partial_a \lambda_{3,j-1/2}^{ROE} (\mathbf{U}_j - \mathbf{U}_{R,j-1/2}^*) d_{3,j-1/2} \end{aligned}$$

► HLLC-type scheme: same structure as the state.

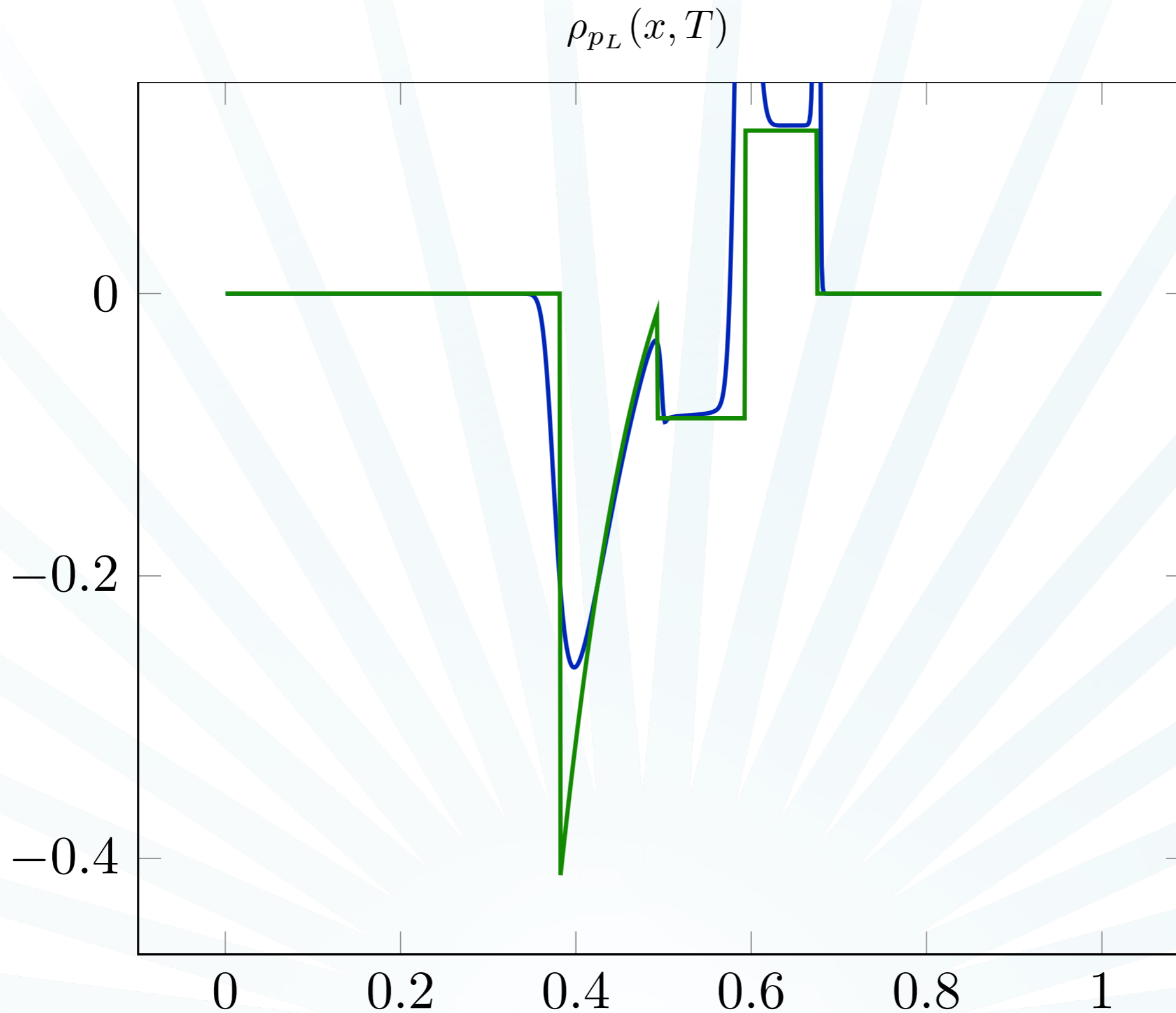
HLL consistency conditions + Rankine-Hugoniot conditions. Equivalent to:

$$\mathbf{U}_{a,L}^* = \mathbf{U}_{a,L} + \alpha_{1,a} \tilde{\mathbf{r}}_1 + \alpha_1 \tilde{\mathbf{r}}_{1,a} \quad \mathbf{U}_{a,R}^* = \mathbf{U}_{a,R} - \alpha_{3,a} \tilde{\mathbf{r}}_3 - \alpha_3 \tilde{\mathbf{r}}_{3,a}$$

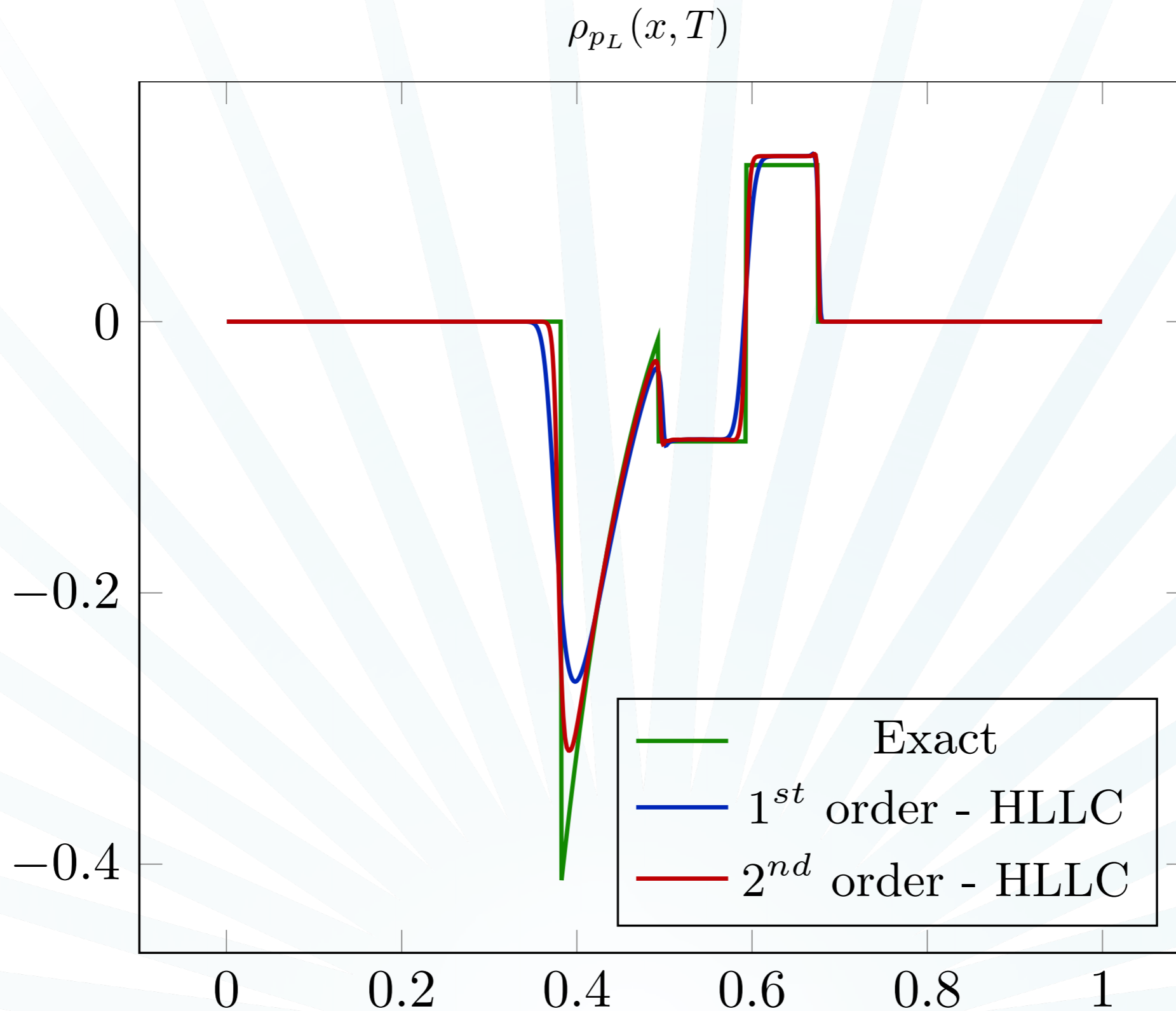
# Classical numerical schemes



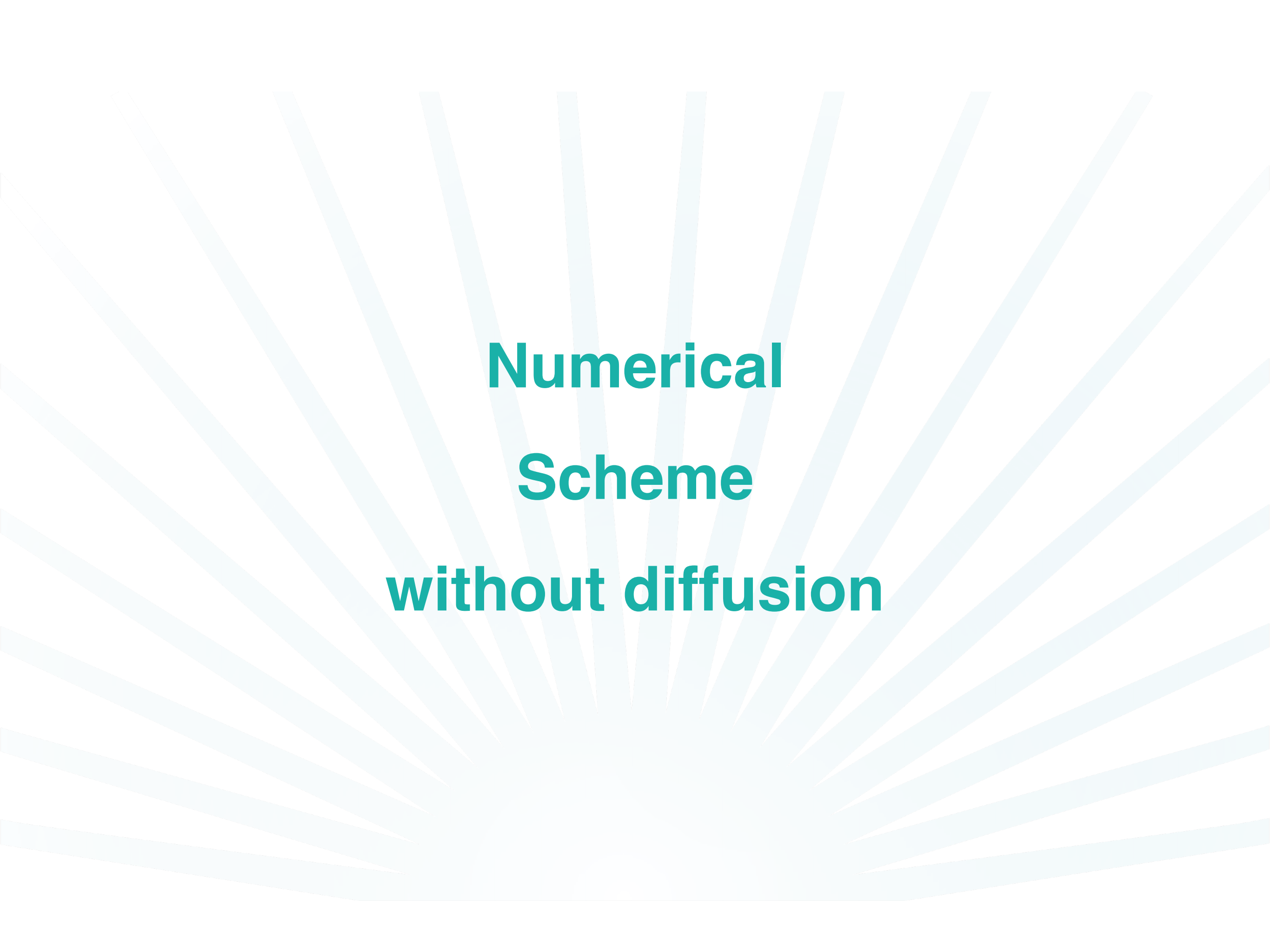
# Classical numerical schemes



# Classical numerical schemes







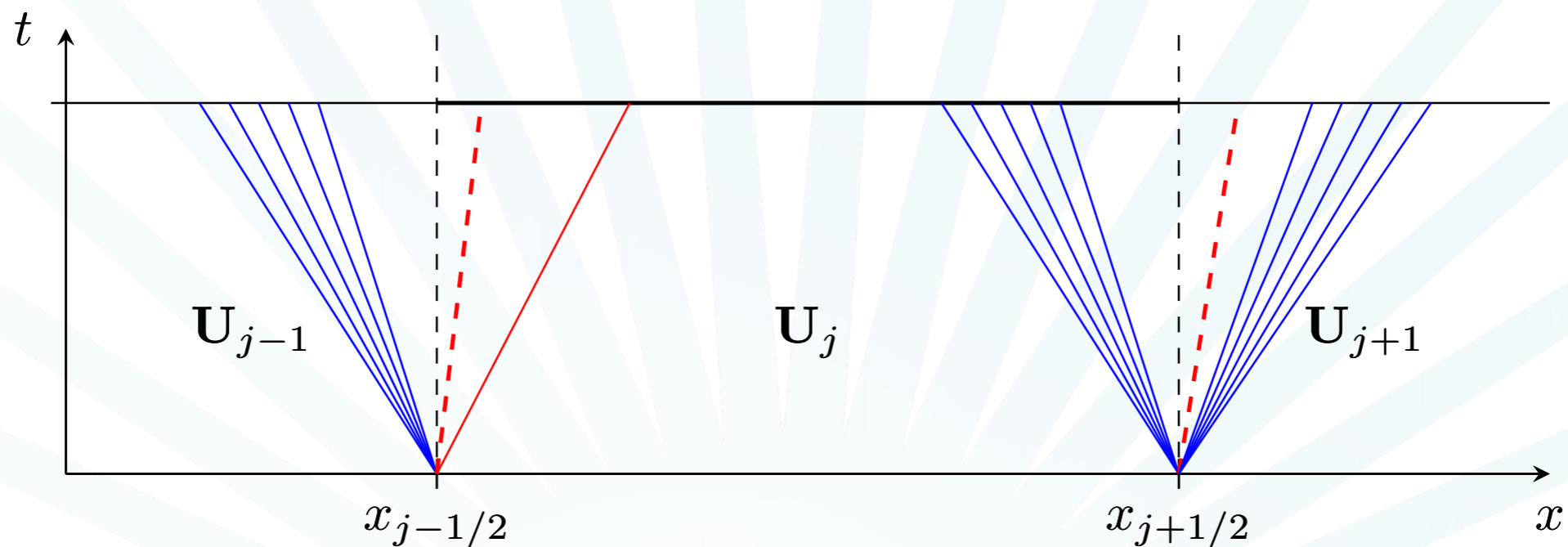
**Numerical  
Scheme  
without diffusion**

# Scheme without numerical diffusion

Step 0 : initial data discretisation

Step 1 : solution of the Riemann problems, one for each interface

~~Step 2 : average~~

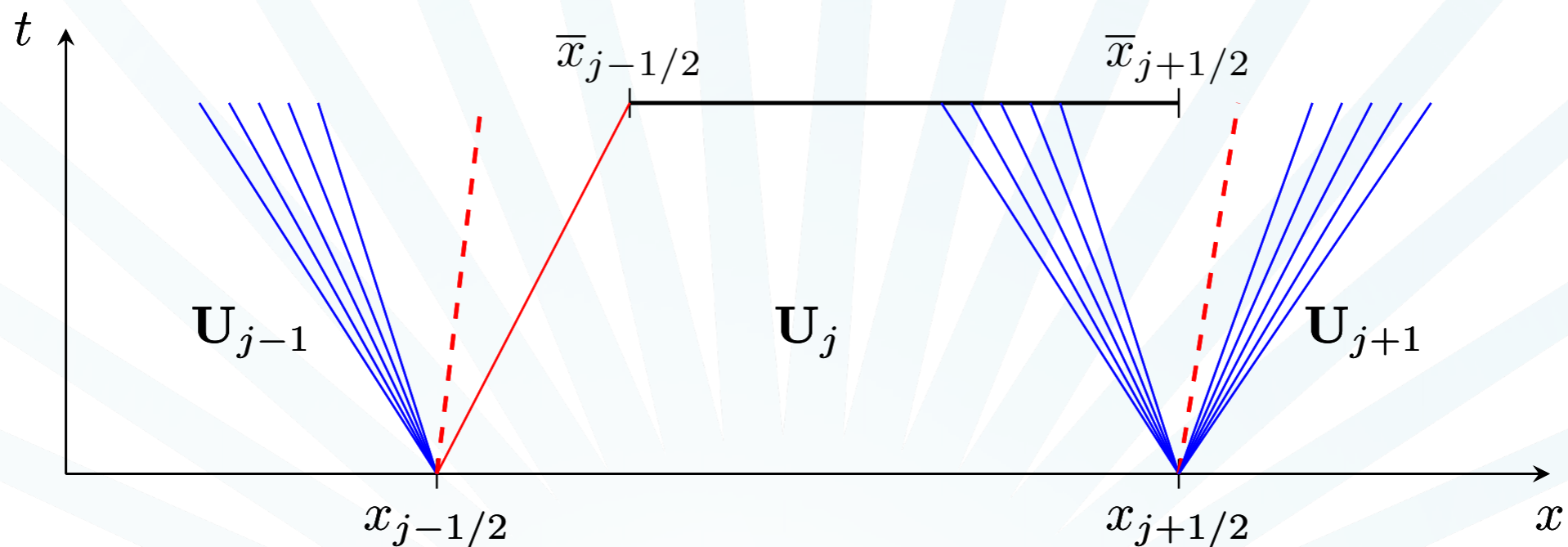


# Scheme without numerical diffusion

Step 0 : initial data discretisation

Step 1 : solution of the Riemann problems, one for each interface

Step 2 : definition of a staggered mesh on which the average is performed



$$\bar{x}_{j-1/2} = x_{j-1/2} + \sigma_{j-1/2} \Delta t$$

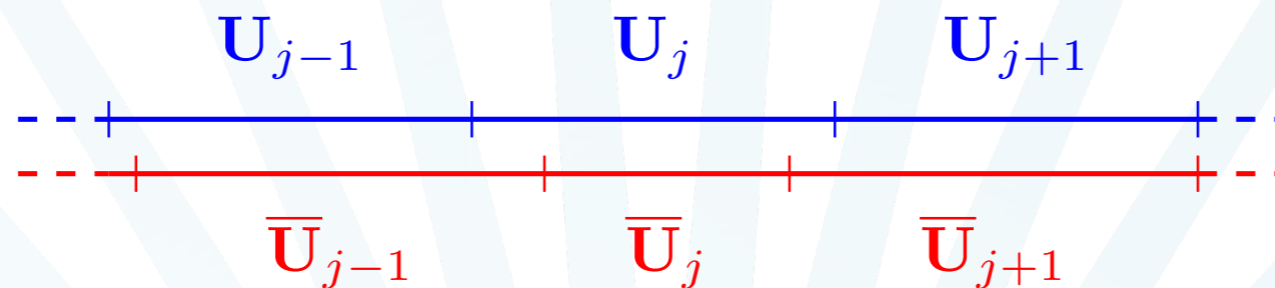
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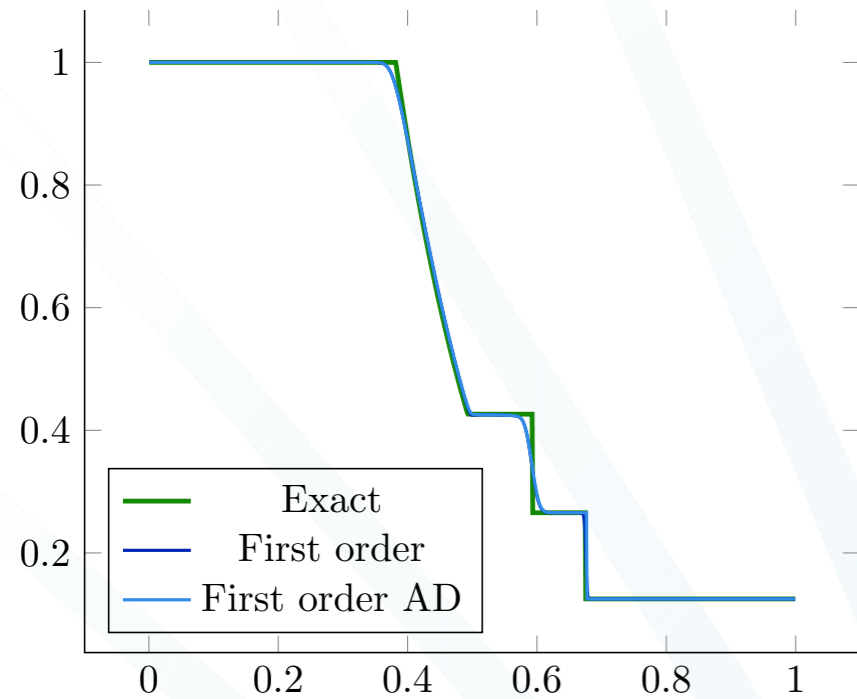
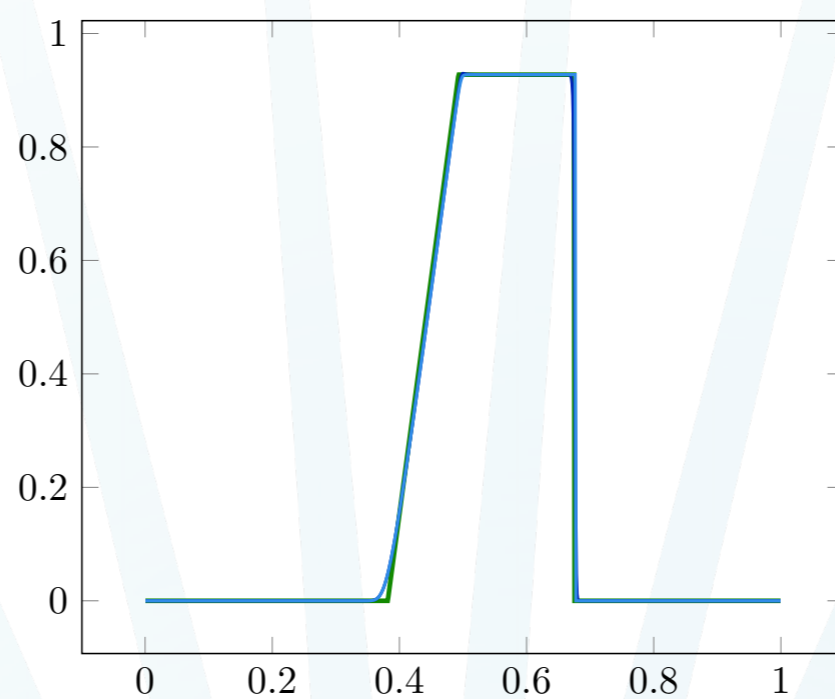
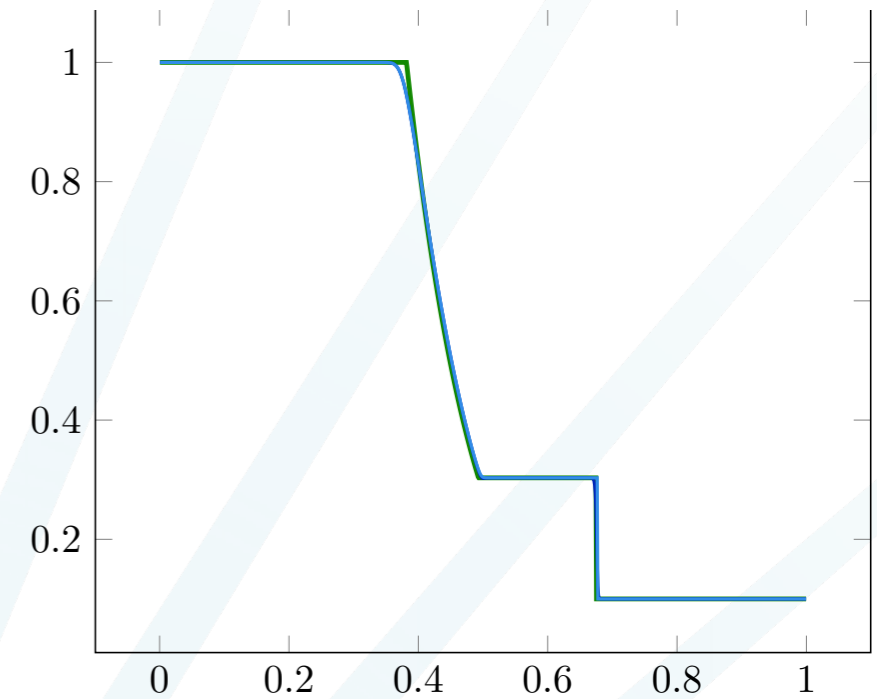
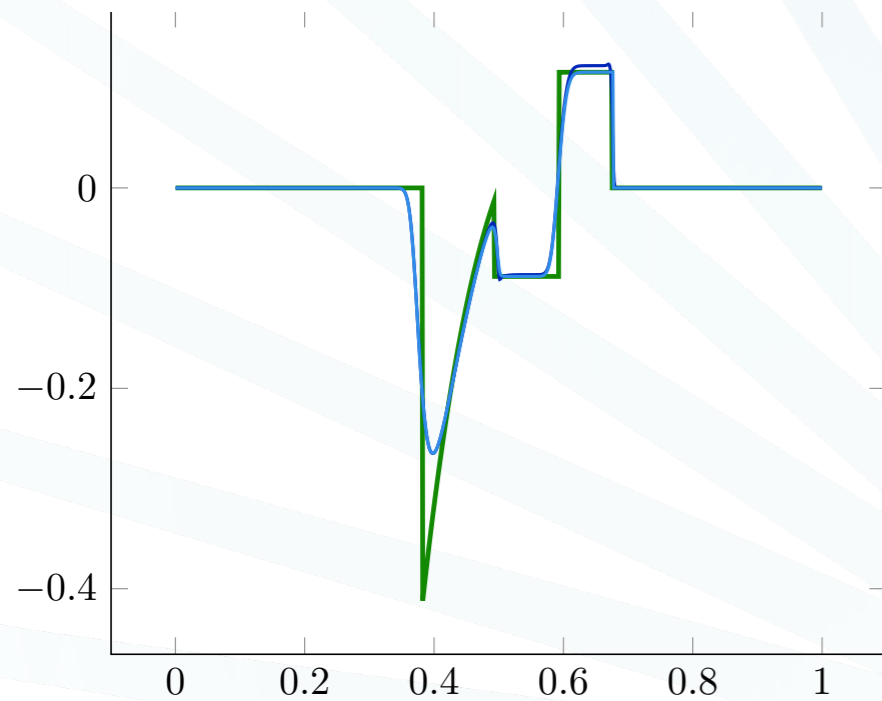
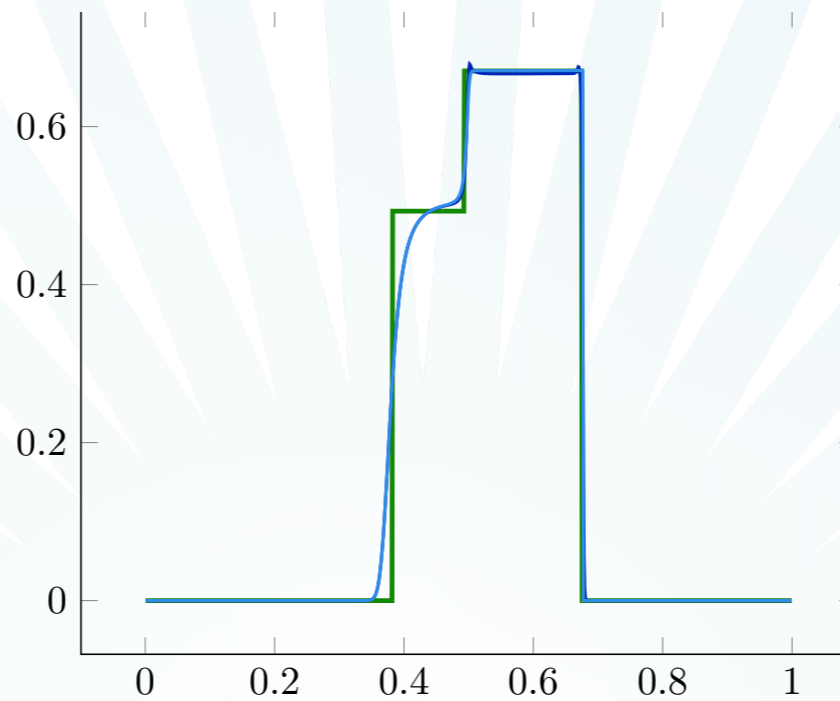
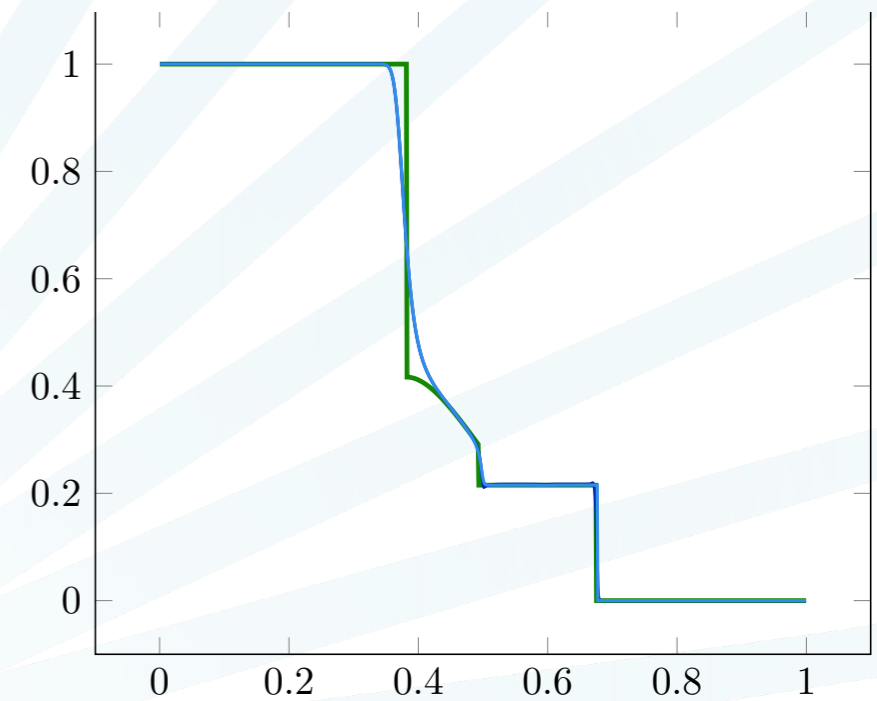
Step 3 : projection on the initial mesh



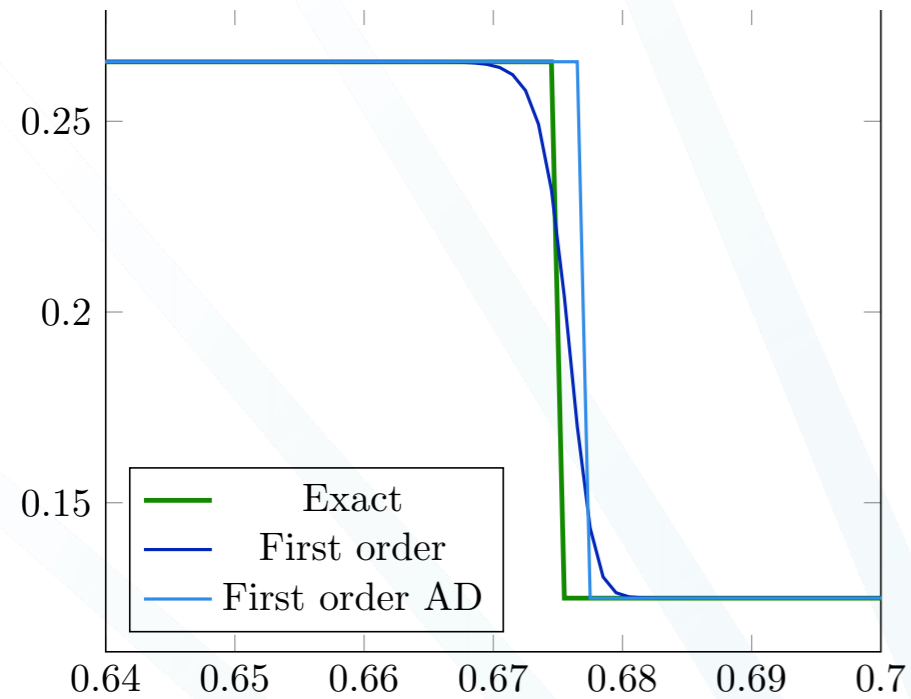
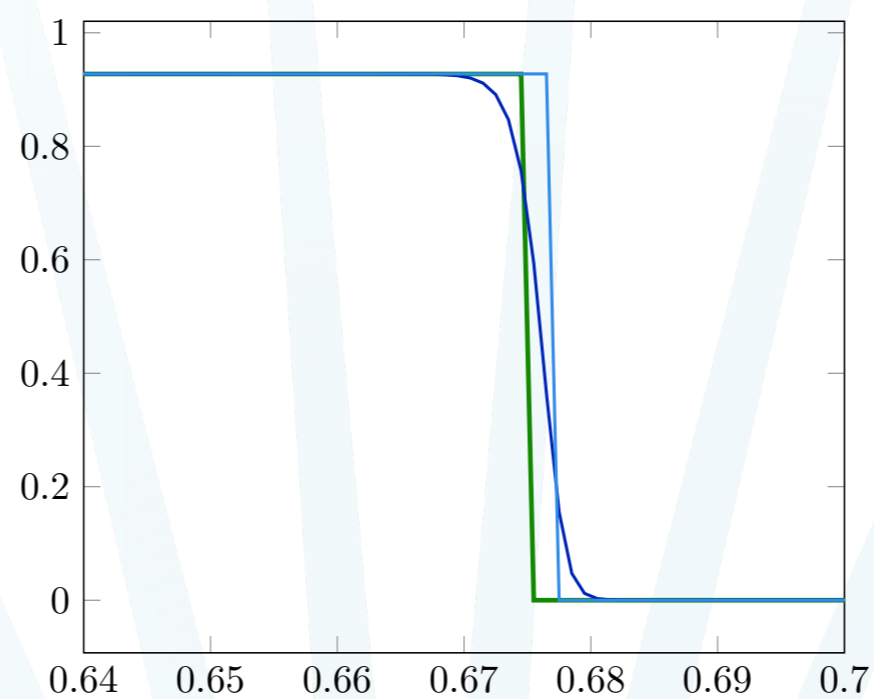
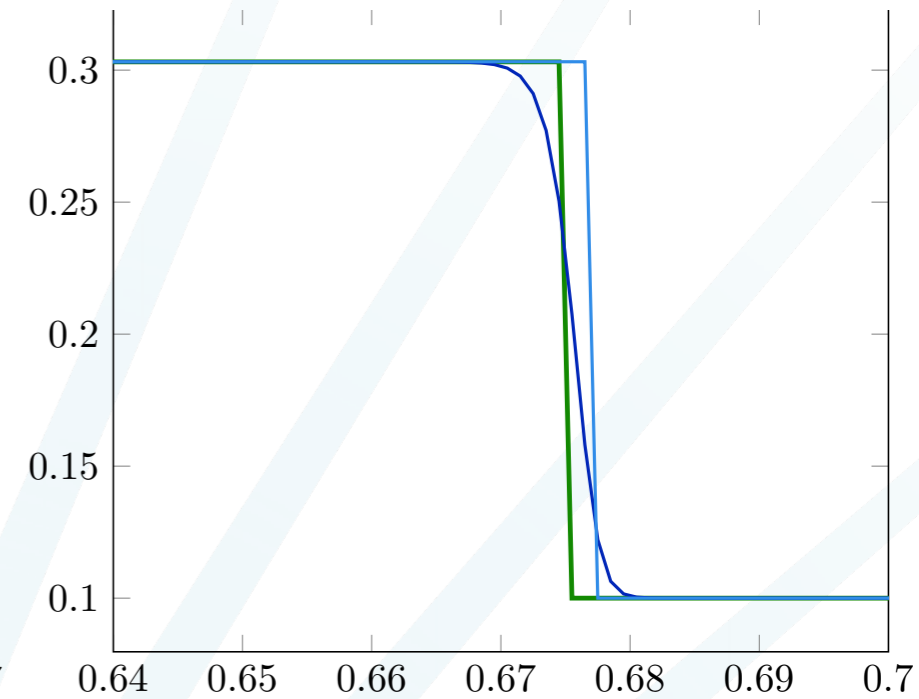
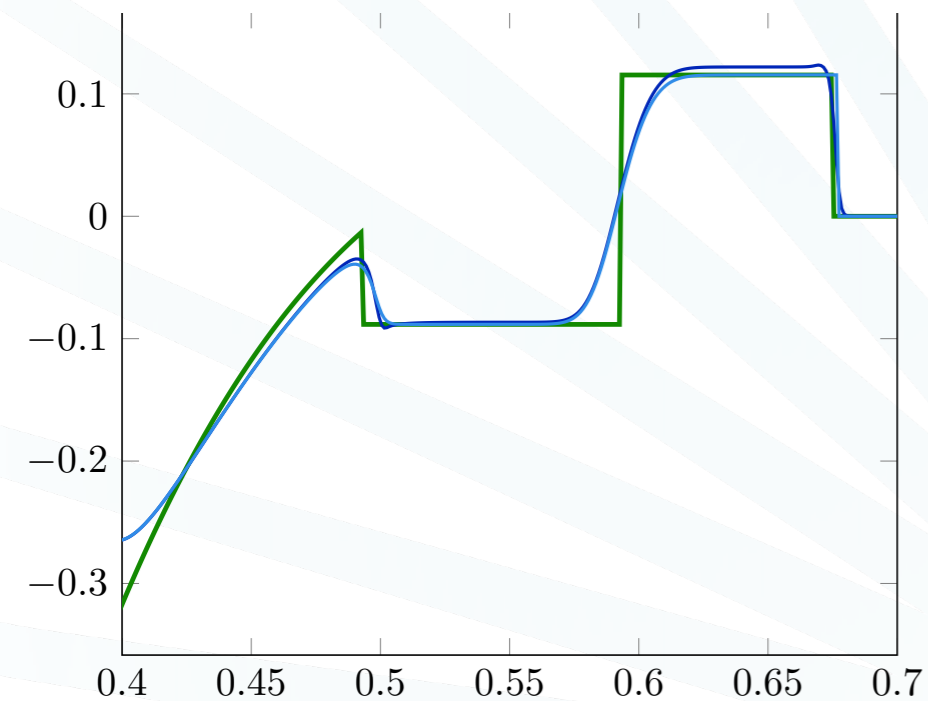
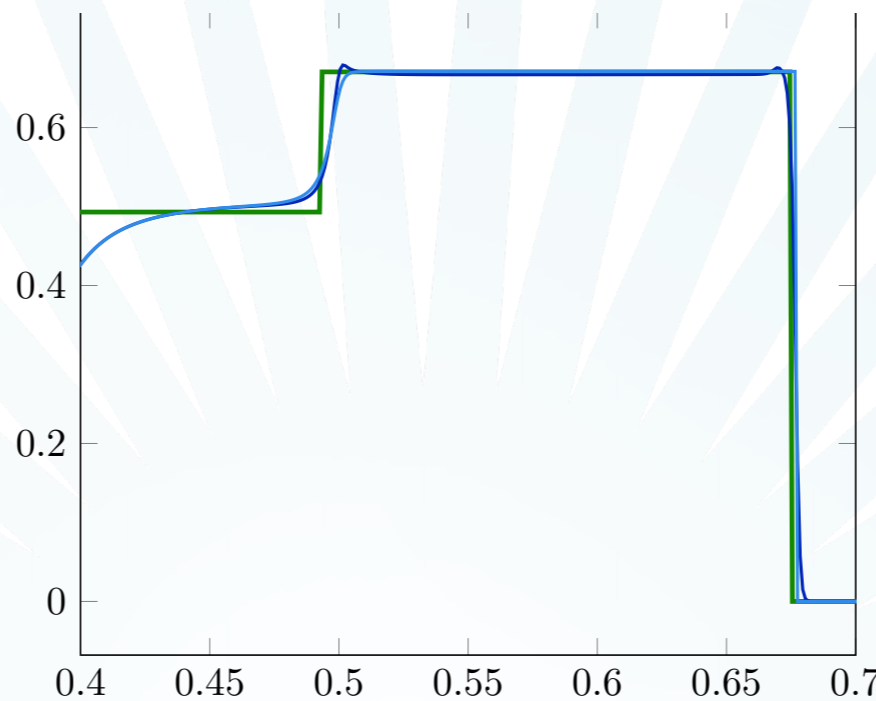
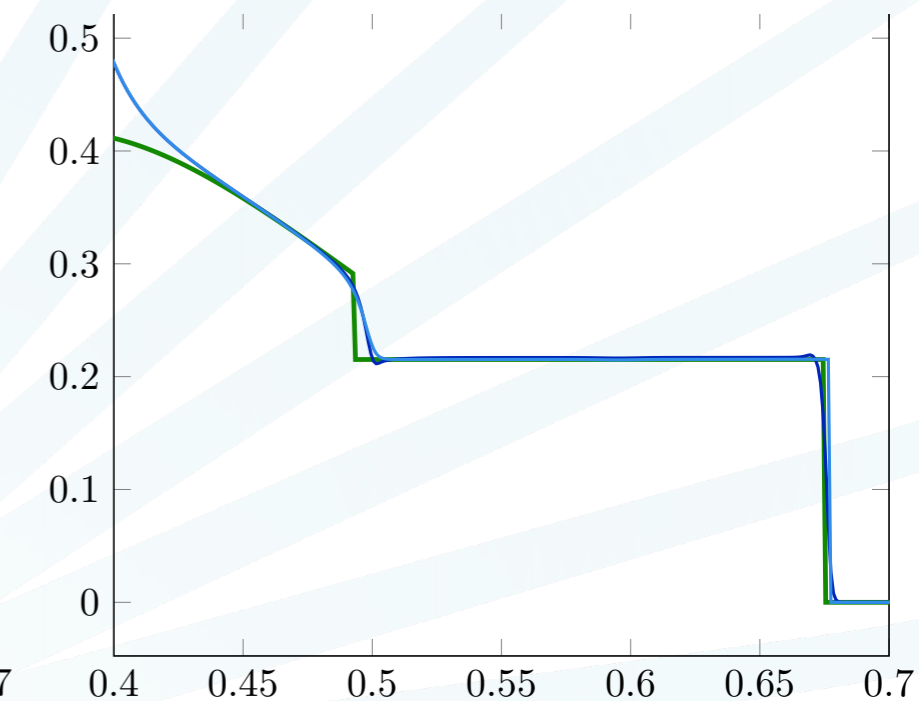
$$\mathbf{U}_j = \begin{cases} \bar{\mathbf{U}}_{j-1} & \text{if } \alpha \in (0, \frac{\Delta t}{\Delta x} \max(\sigma_{j-1/2}, 0)) , \\ \bar{\mathbf{U}}_j & \text{if } \alpha \in [\frac{\Delta t}{\Delta x} \max(\sigma_{j-1/2}, 0), 1 + \frac{\Delta t}{\Delta x} \min(\sigma_{j+1/2}, 0)) , \\ \bar{\mathbf{U}}_{j+1} & \text{if } \alpha \in [1 + \frac{\Delta t}{\Delta x} \min(\sigma_{j+1/2}, 0), 1) . \end{cases}$$

$$\alpha \sim \mathcal{U}([0, 1])$$

# Numerical results

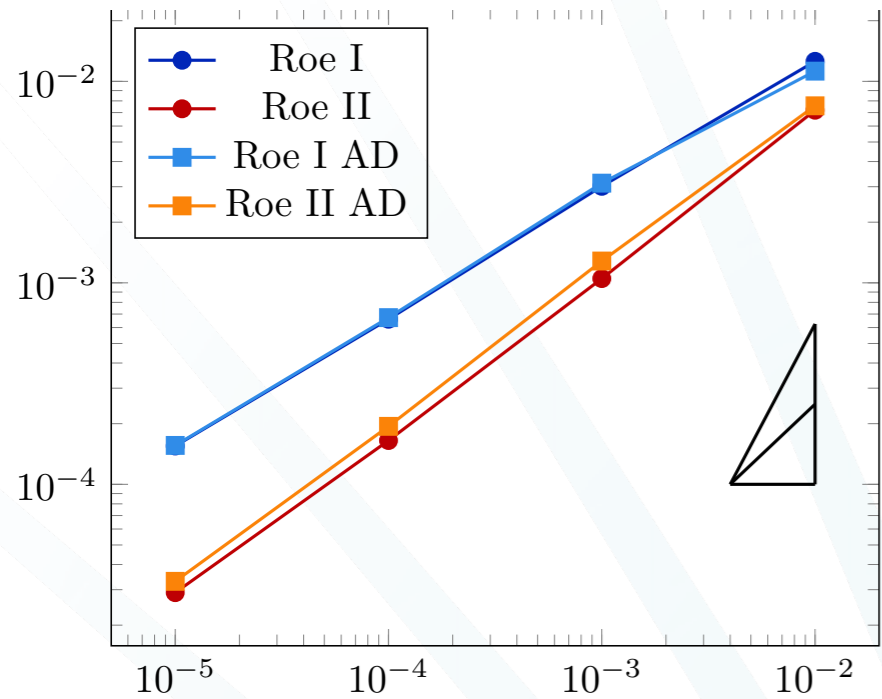
 $\rho(x, T)$  $u(x, T)$  $p(x, T)$  $\rho_{p_L}(x, T)$  $u_{p_L}(x, T)$  $p_{p_L}(x, T)$ 

# Numerical results

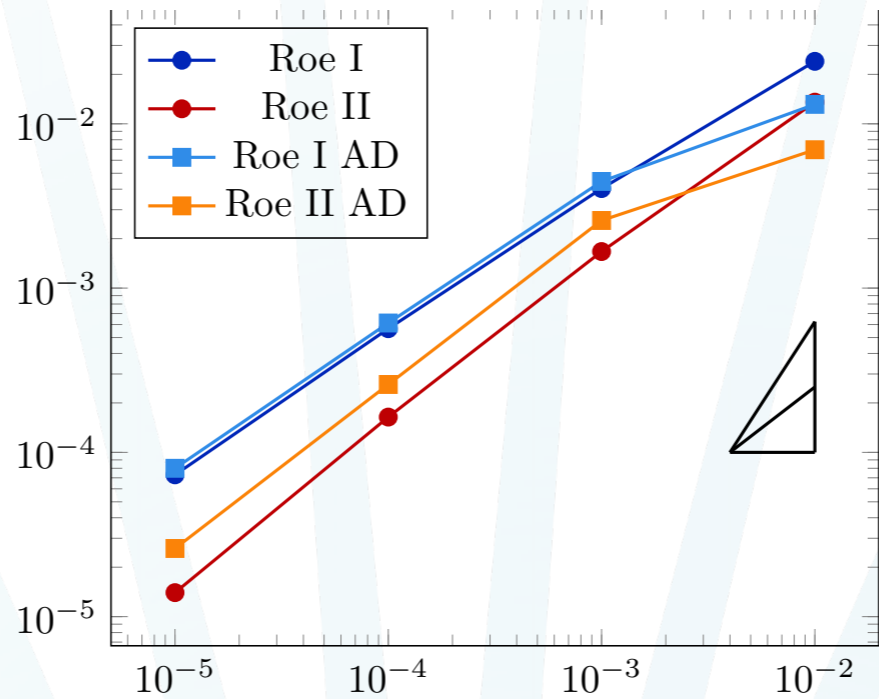
 $\rho(x, T)$  $u(x, T)$  $p(x, T)$  $\rho_{p_L}(x, T)$  $u_{p_L}(x, T)$  $p_{p_L}(x, T)$ 

# Convergence

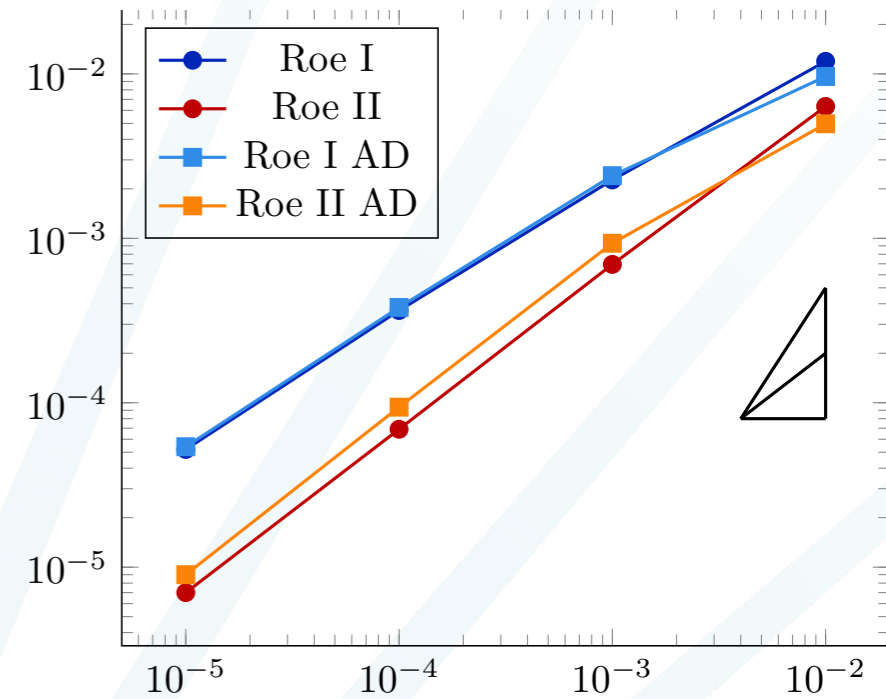
$$\|\rho^{ex}(x, T) - \rho(x, T)\|_{L^1(0,1)}$$



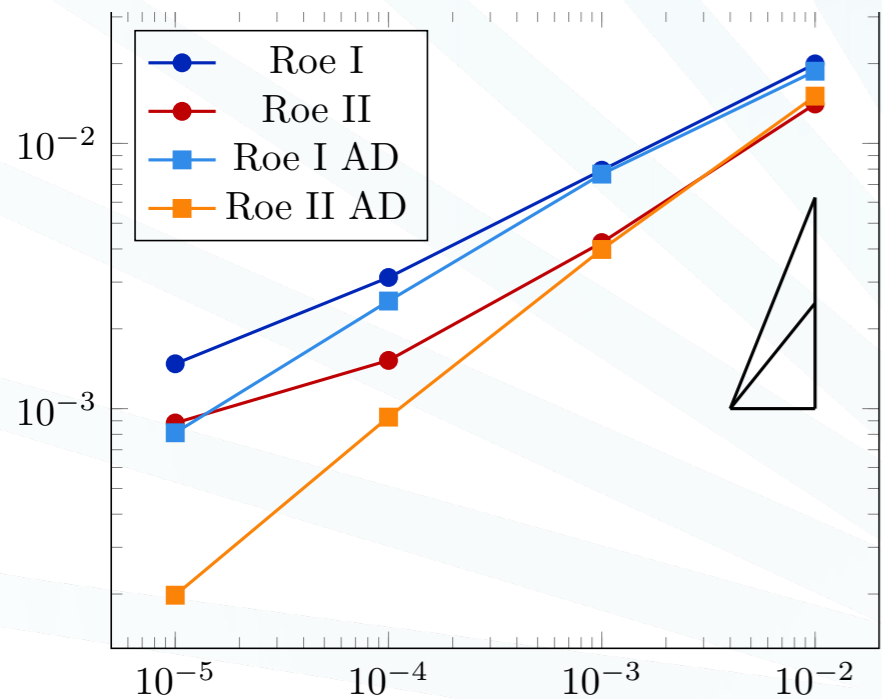
$$\|u^{ex}(x, T) - u(x, T)\|_{L^1(0,1)}$$



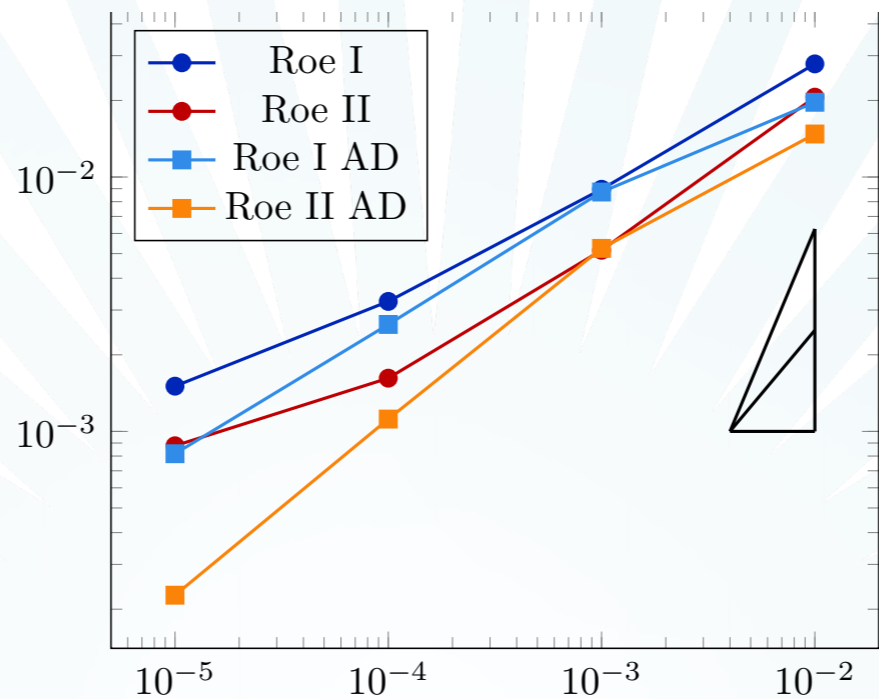
$$\|p^{ex}(x, T) - p(x, T)\|_{L^1(0,1)}$$



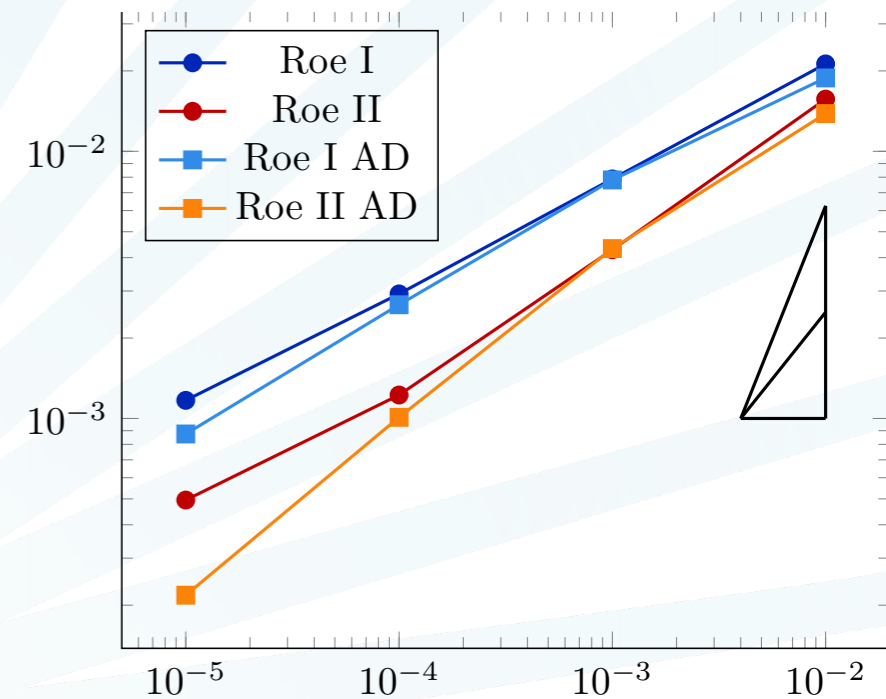
$$\|\rho_{p_L}^{ex}(x, T) - \rho_{p_L}(x, T)\|_{L^1(0,1)}$$



$$\|u_{p_L}^{ex}(x, T) - u_{p_L}(x, T)\|_{L^1(0,1)}$$



$$\|p_{p_L}^{ex}(x, T) - p_{p_L}(x, T)\|_{L^1(0,1)}$$





# Uncertainty quantification



# Uncertainty Quantification

Riemann problem with uncertain parameters:  $\mathbf{a} = (\rho_L, \rho_R, u_L, u_R, p_L, p_R)^t$

Aim: determine a **confidence interval**  $CI_X = [\mu_X - \kappa\sigma_X, \mu_X + \kappa\sigma_X]$

Monte Carlo approach:  $N$  samples  $X_k$

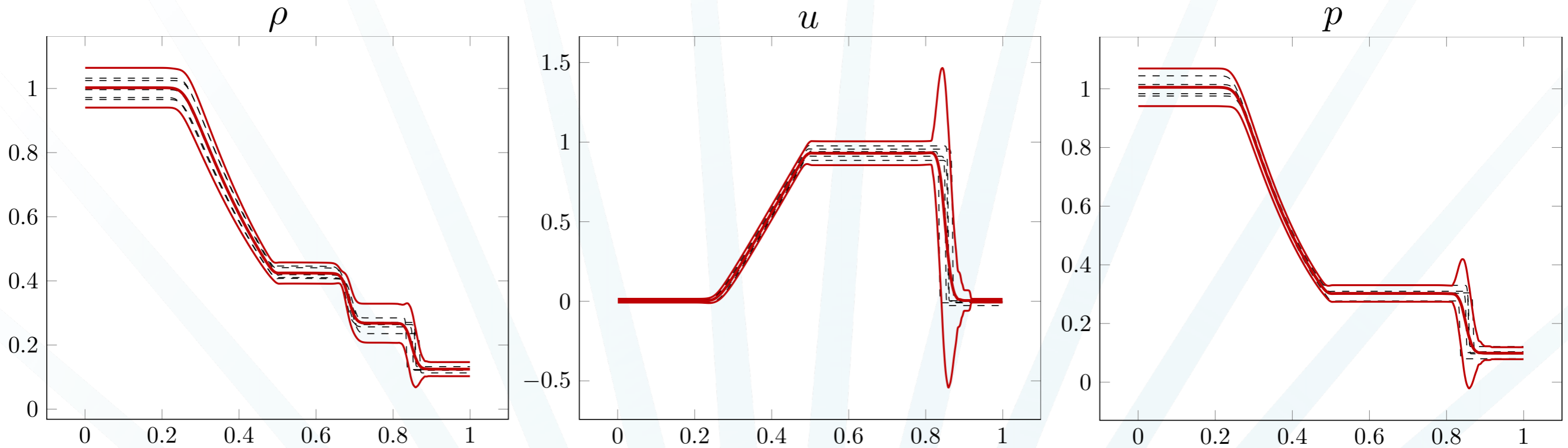
$$\mu_X = \frac{1}{N} \sum_{k=1}^N X_k \quad \sigma_X^2 = \frac{1}{N-1} \sum_{k=1}^N |\mu_X - X_k|^2$$

Sensitivity approach: state  $X$ , sensitivities  $X_{a_i}$

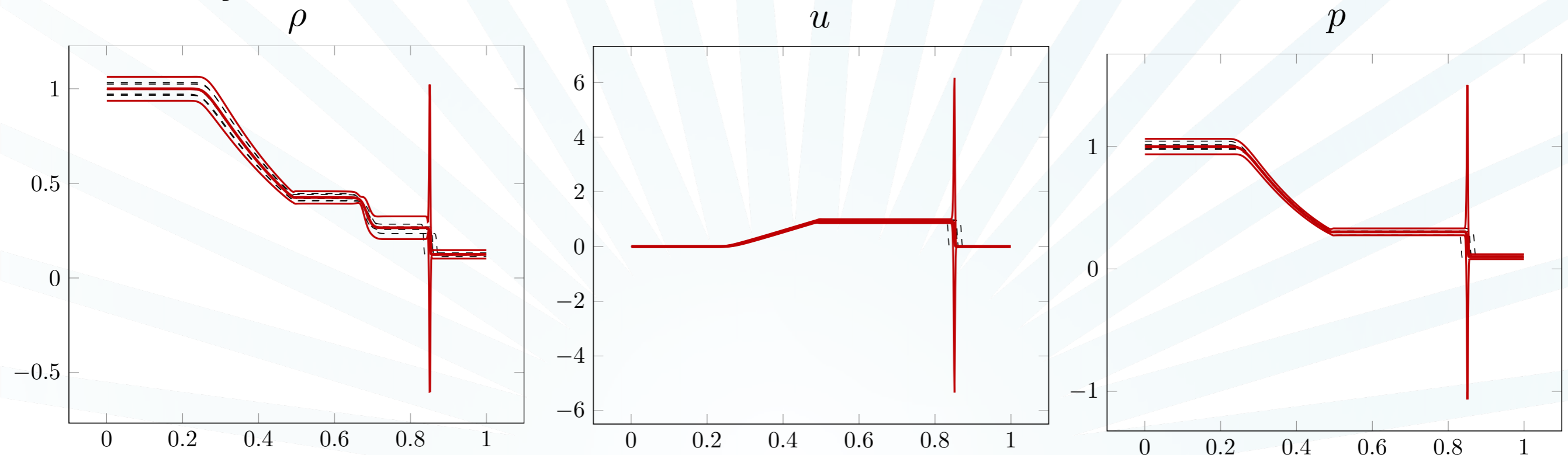
$$\mu_X = X \quad \sigma_X^2 = \sum_{i=1}^6 X_{a_i}^2 \sigma_{a_i}^2$$

# Uncertainty Quantification

## Monte Carlo method:

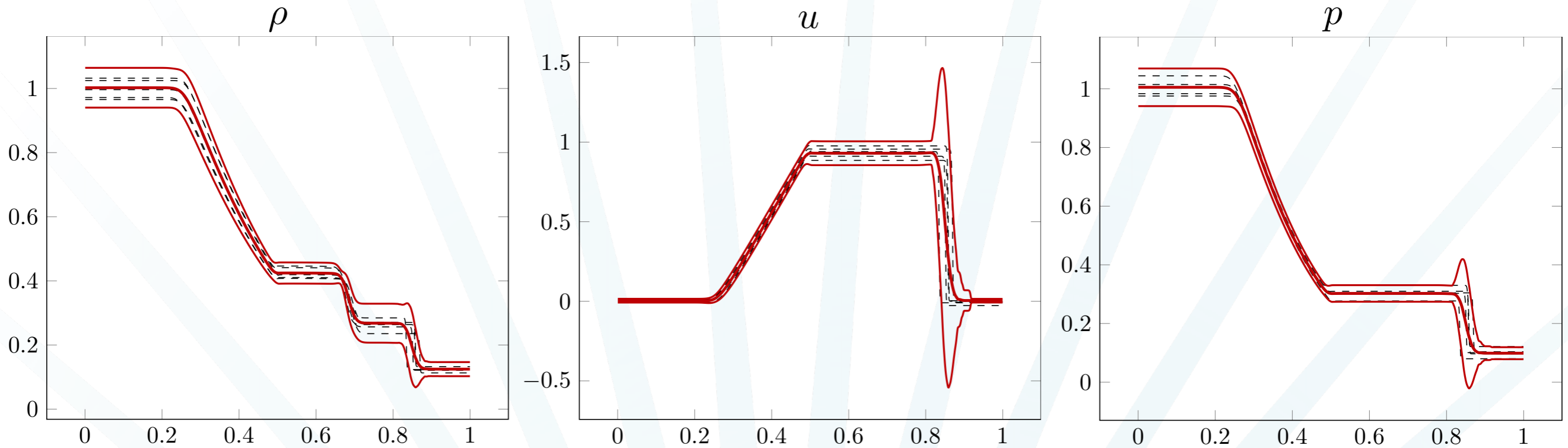


## Sensitivity method without correction:

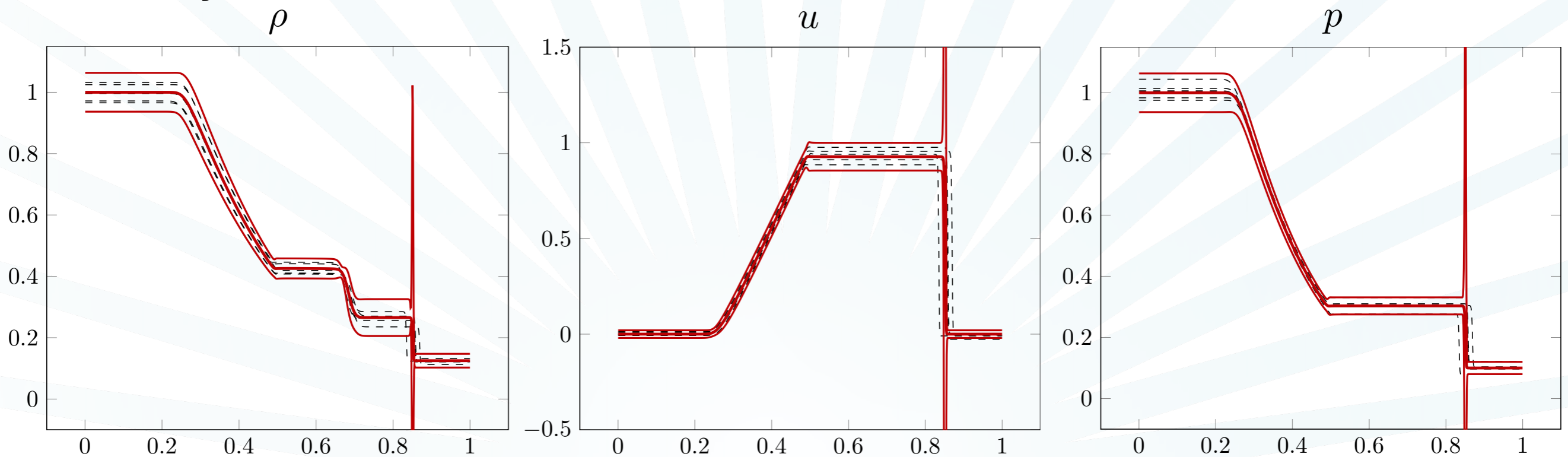


# Uncertainty Quantification

Monte Carlo method:

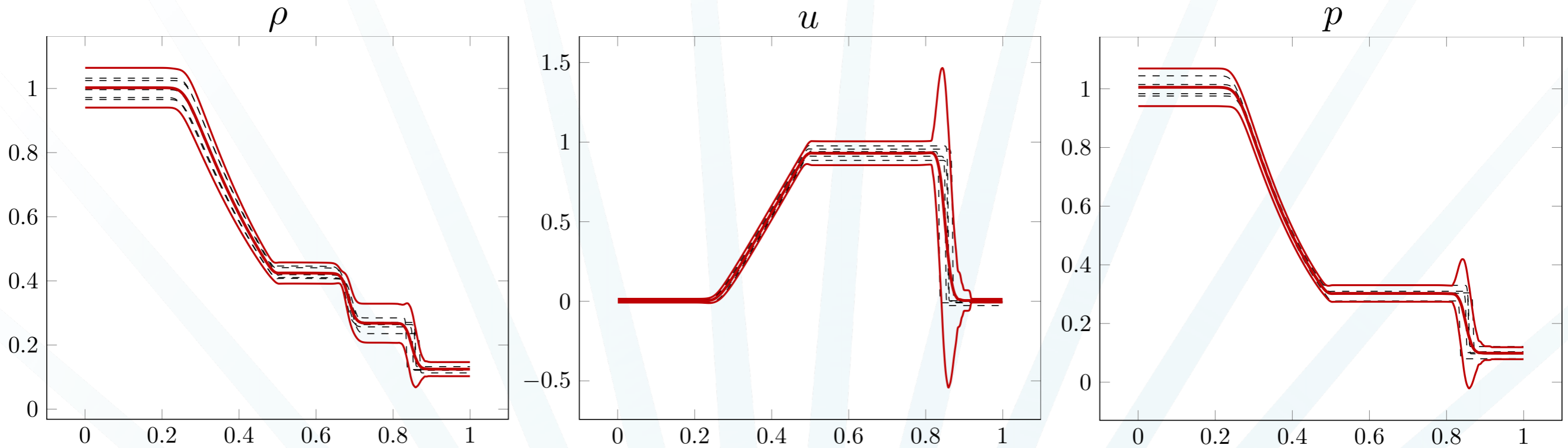


Sensitivity method without correction:

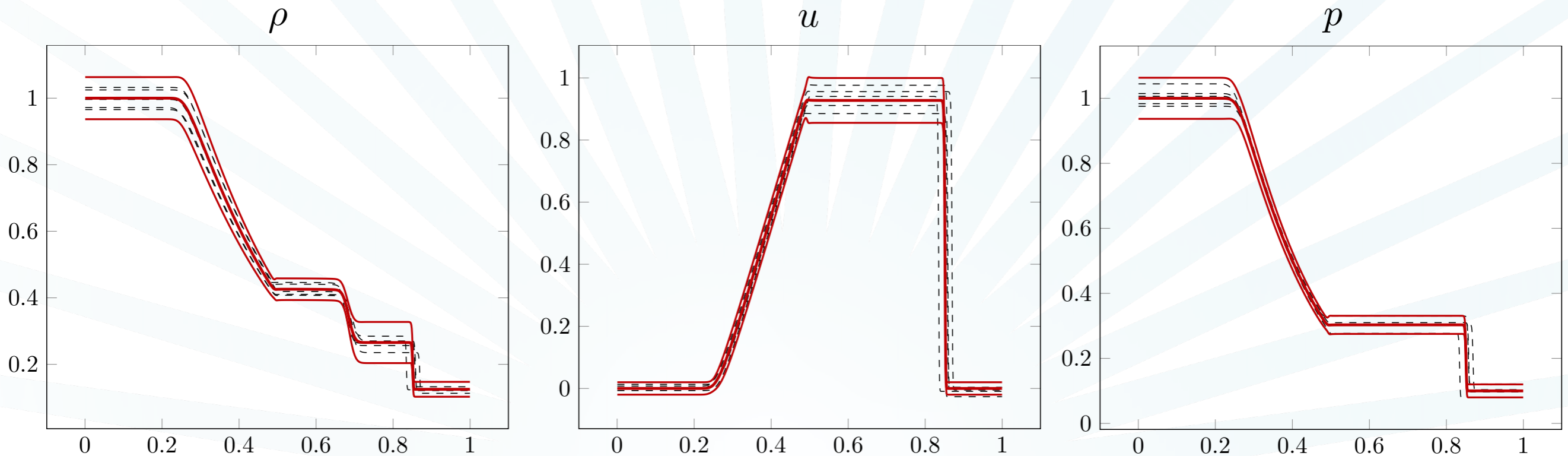


# Uncertainty Quantification

Monte Carlo method:



Sensitivity method with correction (diffusive method):



# Conclusion and future development

## Conclusion:

- ▶ We defined a sensitivity system valid in case of discontinuous state
- ▶ The correction term is well defined
- ▶ The correction term is important in applications

## Future development:

- ▶ Effects of the numerical diffusion for the applications
- ▶ Extension to 2D



**Thank you  
for your attention!**