

# A-posteriori estimators for conservation laws

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## Scalar non linear conservation law:

Find  $u : \mathbb{R}^d \times \mathbb{R}^+ \rightarrow \mathbb{R}$  solution of

$$\partial_t u(x, t) + \nabla \cdot f(u(x, t)) = 0 \quad u(x, 0) = u_0(x)$$

with e.g.  $f(u) = \frac{1}{2}u^2$

**Viscosity limit:** let  $u_\varepsilon$  be a classical solution of the regularized problem:

$$\partial_t u_\varepsilon(x, t) + \nabla \cdot f(u_\varepsilon(x, t)) = \varepsilon \Delta u_\varepsilon(x, t), \quad u_\varepsilon(x, 0) = u_0(x)$$

There exists  $u = \lim_{\varepsilon \rightarrow 0} u_\varepsilon$  (a.e.) and  $u$  is weak solution.

**$u$  is physically relevant weak solution**

## Equivalent: Entropy Solution

$$-\int_{\mathbb{R}^d} \int_{\mathbb{R}^+} (S(u) \partial_t \phi + F_S(u) \cdot \nabla \phi) dt dx - \int_{\mathbb{R}^d} S(u_0) \phi(x, 0) dx \leq 0$$

for all entropy pairs  $(S, F_S)$ , i.e.,  $S$  convex and  $F'_S = S'f'$

## System of conservation law (e.g. Euler Equations):

Find  $U : \mathbb{R}^d \times \mathbb{R}^+ \rightarrow \mathbb{R}^m$  entropy solution

$$\partial_t U(x, t) + \nabla \cdot F(U(x, t)) = 0 \quad U(x, 0) = U_0(x)$$

Integrate over  $T \in \mathbf{T}$  with tessellation  $\mathbf{T}$  of  $\Omega$ :

$$\int_T \partial_t U(\cdot, t) = - \int_T \nabla \cdot F(U(\cdot, t)) = - \int_{\partial T} F(U(\cdot, t)) \cdot n$$

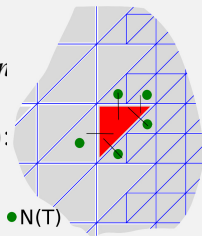
Piecewise constant approximation  $U_T(t) \approx \frac{1}{|T|} \int_T U(\cdot, t)$ :

$$\frac{d}{dt} U_T(t) = - \frac{1}{|T|} \int_{\partial T} F_h(t)$$

with numerical flux  $F_h(t) = F_{T, T'}(U_T(t), U_{T'}(t))$  on intersection between neighboring elements  $T, T'$ :

$$\frac{d}{dt} U_T(t) = - \frac{1}{|T|} \sum_{T' \in N(T)} F_{T, T'}(U_T(t), U_{T'}(t))$$

$N(T)$  is set of all neighbors of  $T$ .



**System of conservation law (e.g. Euler Equations):**Find  $U : \mathbb{R}^d \times \mathbb{R}^+ \rightarrow \mathbb{R}^m$  entropy solution

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Semi discrete scheme:

$$\frac{d}{dt} U_T(t) = - \frac{1}{|T|} \sum_{T' \in N(T)} F_{T, T'}(U_T(t), U_{T'}(t))$$

 $N(T)$  is set of all neighbors of  $T$ .Forward Euler in time for time steps  $t^n$  and  $\Delta t^n = t^{n+1} - t^n$ :

$$U_T^{n+1} = U_T^n - \frac{\Delta t^n}{|T|} \sum_{T' \in N(T)} F_{T, T'}(U_T^n, U_{T'}^n)$$

## System of conservation law (e.g. Euler Equations):

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Fully discrete scheme:

$$U_T^{n+1} = U_T^n - \frac{\Delta t^n}{|T|} \sum_{T' \in N(T)} F_{T, T'}(U_T^n(t), U_{T'}^n(t))$$

Define  $U_h(x, t) := U_T^n$  for  $x \in T, t \in [t^n, t^{n+1})$ .

## A-priori error estimate for scalar case

Let  $u_h$  be a first order finite-volume approximation then under suitable conditions on the numerical flux:

$$\max_t \|u_h(\cdot, t) - u(\cdot, t)\|_{L^1(\mathbb{R}^d)} \leq C(u) h^{\frac{1}{4}}$$

Should be  $h^{\frac{1}{2}}$ , only proven for structured grids.

$$\partial_t \rho + \nabla \cdot (\rho \vec{u}) = 0 \quad (\text{conservation of mass}),$$

$$\partial_t (\rho \vec{u}) + \nabla \cdot (\rho \vec{u} \vec{u}^T + \mathbf{P}) = 0 \quad (\text{conservation of momentum}),$$

$$\partial_t (\rho e) + \nabla \cdot (\rho e \vec{u} + \mathbf{P} \vec{u}) = 0 \quad (\text{conservation of energy}),$$

$$e = \varepsilon + \frac{1}{2} |\vec{u}|^2 \quad (\text{total energy}),$$

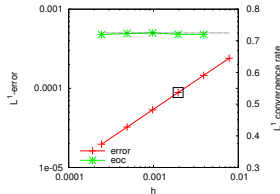
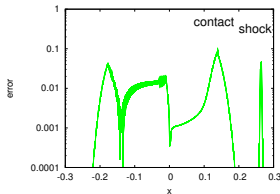
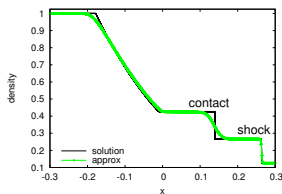
$$\mathbf{P} = p(\rho, \varepsilon) \mathbf{I} \quad (\text{equation of state for pressure}),$$

$$(\rho, \rho \vec{u}, \rho e)(\cdot, 0) = (\rho_0, \rho_0 \vec{u}_0, \rho_0 e_0) \quad (\text{initial conditions})$$

$\rho$  : density,       $\rho \vec{u}$  : momentum,       $\rho e$  : total energy density

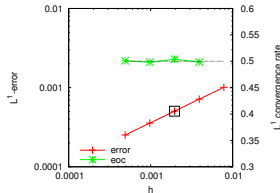
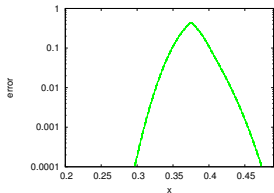
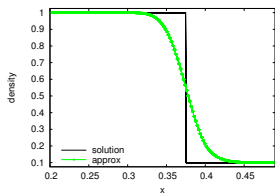
## Riemann problem (Euler)

$u_0 = u_L (x < 0), u_0 = u_R (x > 0)$ . Solution: rarefaction, contact, shock



## Single contact (Euler)

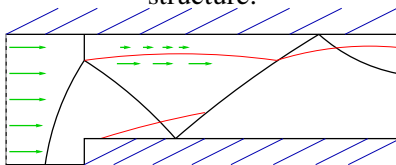
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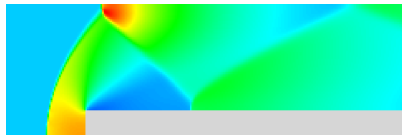
resolution in 3d requires 130.000.000 elements for results shown

Forward facing step (Euler)

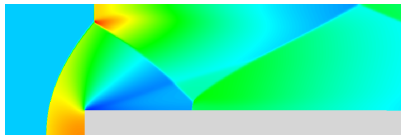
Right moving Mach 3 flow  
structure:



15.000 elements



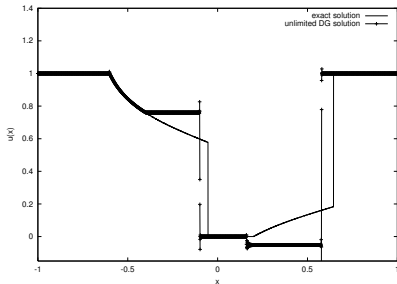
230.000 element



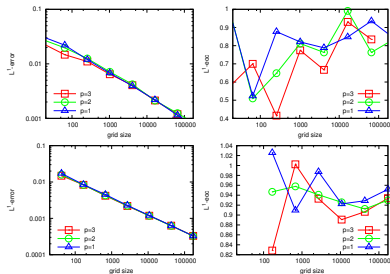


- highly efficient for smooth solution
- Loss of efficiency for non-smooth solution ...
- ... and unstable for non linear discontinuities (shocks)

## No stabilization



## With simple stabilization



- highly efficient for smooth solution
- Loss of efficiency for non-smooth solution ...
- ... and unstable for non linear discontinuities (shocks)

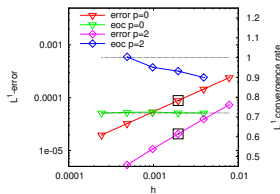
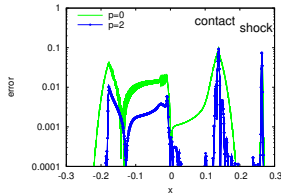
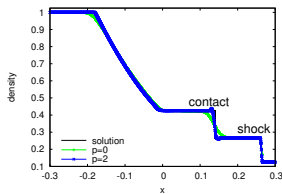
Basis Algorithm:

- 1 determine **troubled cells** where the error is high or the scheme is unstable
- 2 for each troubled cell either increase the order (if solution is smooth) or reduce the order und refine the grid

Determine troubled cells **heuristically** or by **error estimate**.

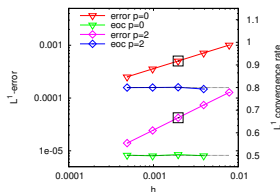
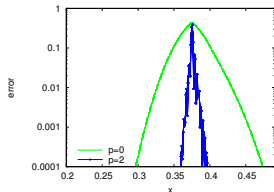
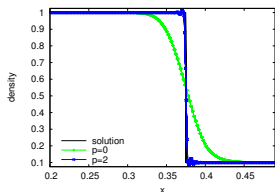
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- Find interface between cells where solution has a large jump
- Mark the two elements at that intersection
- Mark a neighborhood of marked elements
- Mark elements for coarsening where the jump is very small

Possibly look at curvature of solution (i.e. jumps between gradients)

**Idea:** Error is where the discontinuities are

**Problems?**

- Can not distinguish between contacts and shocks
- Could coarsen wrongly (kinks at ends of rarefactions)
- Indicator does not get smaller with reduction of grid size
- No mathematical proof that it works only many people using successfully...

## Estimator

Need indicator  $\eta_K$  for "smoothness" of solution

$$\eta_K = \begin{cases} O(h_K^q) & \text{smooth region} \\ O(h_K^{-1}) & \text{troubled region} \end{cases}$$

# A posteriori error estimate

## Kruzkov framework (semi implicit)

D, Makridakis, Ohlberger '06

Structure of a posteriori error estimate:

$$\|(u - u_h)(T)\|_{L^1(B_R(x_0))}^2 \leq K \sum_n \sum_{j \in J^n} \left( h_j + \|\bar{u}_j^n - \tilde{u}_j^n\|_{L^\infty} \right) (\mathbf{R}_{T,j}^n + \mathbf{R}_{S,jl}^n + \mathbf{R}_{L,j}^n)$$

- $\mathbf{R}_{T,j}^n$ : Element residual
- $\mathbf{R}_{S,jl}^n$ : Jump residual (numerical viscosity)
- $\mathbf{R}_{L,j}^n$ : Coarsening error
- $\|\bar{u}_j^n - \tilde{u}_j^n\|_{L^\infty}$ : difference between average and higher order polynomial

Numerical test show:

$$\bar{R}_j^n := \frac{h_j}{|T_j| \Delta t^n} (\mathbf{R}_{T,j}^n + \mathbf{R}_{S,jl}^n + \mathbf{R}_{L,j}^n) = \begin{cases} O(h_j^{p-1}) & \text{solution is smooth} \\ O(1) & \dots \text{ discontinuous} \end{cases}$$

## First step:

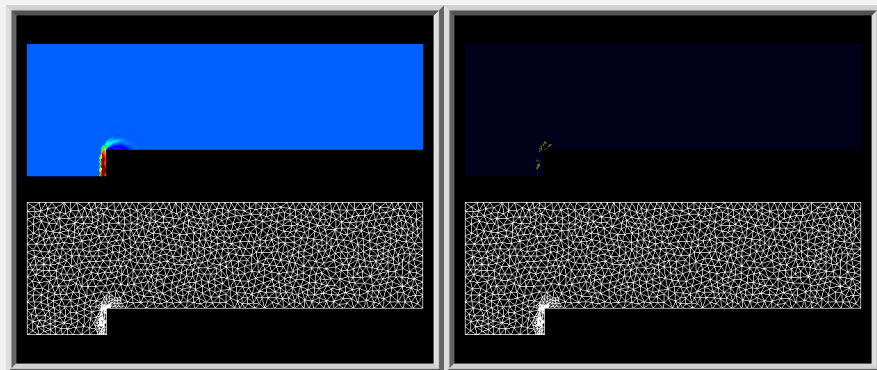
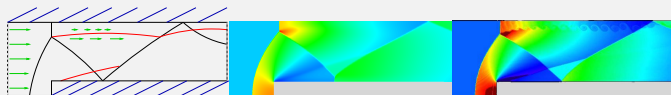
Define the set of grid cells  $\mathbf{I}_s$  with a significant contribution to the overall error indicator  $\eta_h$ .

## Second step:

Use an **equal distribution strategy** to refine or coarsen the elements in  $\mathbf{I}_s$  according to the **error estimate**.

## Third step:

For elements that are marked for coarsening, check if the **projection error** is small enough.



Approximately 14.000 elements at  $t = T$



We now come to a short commercial break... 

at University of Warwick 4th to 8th of July, 2016

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**Issues with Kruzkov:** only works with scalars

**Issue with proof:** only done for semi discrete scheme

Giesselmann, Makridakis, Pryer: use of relative entropy (RE) framework

Given one convex entropy  $\eta$  then

$$\eta(U | V) := \eta(U) - \eta(V) - \eta'(V)(U - V) \approx \|U - V\|_{L^2}$$

and if  $V$  is a Lipschitz solution and  $U$  a weak solution then

$$\frac{d}{dt} \int \eta(U | V) \leq C(V) \int \eta(U | V)$$

This can also be used to study perturbed solutions  $U$ :

**Advantage:** works with system of equations (need only one entropy)

**Issue with RE:** only works in the case that a Lipschitz solution exists

**Issue with proof:** semi discrete scheme in 1d, very restrictive on flux function

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# A-posteriori estimator based on RE (semi discrete version in 1d)

- Construct semi discrete solution  $U(t)$  in the DG space of order  $p$
- Use a spacial reconstruction  $U^s(t)$  in a Lagrange space of order  $p + 1$ :  
construct solution on each element
  - local  $L^2$  projection on polynomial space of order  $p - 1$
  - use  $U^* = U^*(U^+, U^-)$  for continuity at the vertices of the grid
- Compute pointwise residual using  $U^s(t)$ :

$$R^s = \partial_t U^s + \nabla \cdot F(U^s)$$

Assumption:

- Have good choice for  $U^*$  (get to that later)
- existence of smooth solutions (we are stuck with that)

**Problem:** General ODE

$$\frac{d}{dt}U(t) = F(U(t))$$

**Assume:** some time stepping method of order  $q$  giving approximations

$$(t^0, U^0), (t^1, U^1), \dots, (t^n, U^n), (t^{n+1}, U^{n+1})$$

**Reconstruction:** Choose  $s$  and  $d$  with  $(d+1)s - 1 := r \geq q$  and assume given approximations

$$U_{ij} \approx \frac{d^i}{dt^i} U(t^{n-j})$$

for  $i = 0, \dots, d$  and  $j = -1, \dots, s$  (could be more general).

Using these values construct polynomial  $U^t := H^{s,d}$  of order  $r$  on  $(t^n, t^{n+1})$  through Hermite interpolation.

Easiest example is  $d = 1, s \geq 0$  since we can take  $U_{1j} = F(U(t^{n-j}))$

Otherwise can use finite difference approximation to approximate higher order derivatives

**Lemma: Stability of Hermite interpolation:**

we can replace exact derivatives with approximations (of the right order)

**Estimate**

Define residual  $R := \frac{d}{dt}U^t(t) - F(U^t(t))$  then

$$\|u - u^t\|_\infty \leq \|R\|_{L^1} e^{LT}$$

**Note:**  $U^t \in C^1(0, T)$  and locally computable

**Optimality**

Assume  $L^k$  is the Lipschitz constant of  $k$ th derivative of rhs  $F$  then

$$\|R\|_\infty \leq \sum_{k=0}^{q+1} L^k \mathcal{O}(\tau^{q+k}).$$

So we can have  $L = \mathcal{O}(\tau^{-1})$  for residual to be order  $q$

- 1 Compute DG solution  $U^{n+1}$
- 2 Use  $U^{n-j}$  to compute Hermite reconstruction  $U^t$  (in DG space)
- 3 Use spatial reconstruction (given  $U^*$ ) to construct  $U^{ts}$  in Lagrange space
- 4 Compute pointwise Residual  $R$

Estimate:

$$\|U(t^n, \cdot) - U^n(\cdot)\|_{L^2}^2 \leq \|u^{st}(t^n, \cdot) - U^n(\cdot)\|_{L^2}^2 + \|R\|_{L^2((0,t^n) \times \Omega)}^2 \exp(\dots)$$



## Condition on the numerical flux

There exists  $\mathbf{U}^* : \mathcal{U} \times \mathcal{U} \rightarrow \mathcal{U}$  so that

$$\mathbf{F}_{\mathbf{T},\mathbf{T}'}(\mathbf{a}, \mathbf{b}) = \mathbf{F}(\mathbf{U}^*(\mathbf{a}, \mathbf{b})) - \mu(\mathbf{a}, \mathbf{b}; h)h^\nu(\mathbf{b} - \mathbf{a}) \quad \forall \mathbf{a}, \mathbf{b} \in \mathcal{U}$$

with  $\nu \in \mathbb{N}_0$ , matrix-valued  $\mu$  with  $|\mu(\mathbf{a}, \mathbf{b}; h)| \leq \mu_K \left(1 + \frac{\|\mathbf{b} - \mathbf{a}\|}{h}\right)$ .

- Richtmyer flux ( $\mu = 0$ ):

$$U^*(V, W) = \frac{1}{2}(V + W) - \frac{\Delta t}{2h}(F(W) - F(V))$$

- Richtmyer flux with artificial viscosity of the form  $h^2 \partial_x^h (|\partial_x^h u| \partial_x^h u)$  ( $\nu = 1$ )
- The Lax Friedrichs flux

$$\mathbf{F}_{\mathbf{T},\mathbf{T}'}(\mathbf{a}, \mathbf{b}) = \frac{1}{2}(\mathbf{F}(\mathbf{a}) + \mathbf{F}(\mathbf{b})) - \lambda(\mathbf{b} - \mathbf{a})$$

with  $\mathbf{U}^*(\mathbf{a}, \mathbf{b}) = \frac{1}{2}(\mathbf{a} + \mathbf{b})$ ,  $\nu = 0$ , and

$$\mu(\mathbf{a}, \mathbf{b}, h) = \frac{\mathbf{F}(\mathbf{a}) - 2\mathbf{F}(\mathbf{U}^*(\mathbf{a}, \mathbf{b})) + \mathbf{F}(\mathbf{b})}{2\|\mathbf{b} - \mathbf{a}\|^2} \otimes (\mathbf{b} - \mathbf{a}) - \lambda.$$

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## Lemma (Conditional optimality of residuals)

*Consider fully discrete DG scheme of order  $q + 1$  in time,  $q$ -th order polynomials in space, and a numerical flux satisfying previous Assumption.*

*Let the exact error be of order  $\mathcal{O}(h^{q+\gamma})$  with  $\gamma \in \{\frac{1}{2}, 1\}$ .*

*Then, the residual  $\mathbf{R}$  is of order  $\mathcal{O}(h^{q+\gamma} + h^{q+\gamma+\nu-1})$  under CFL conditions.*

So for optimality we need

- CFL condition  
(gives  $\mathcal{O}(\tau^{-1})$  for Lipschitz constant in time reconstruction error)
- numerical flux with  $\nu = 1$  (or  $\mu = 0$ )

## Lemma (Suboptimality)

*Consider a numerical flux satisfying our Assumption with  $\nu = 0$  and  $\mu(a, b; h) = \mu_0 > 0$  and  $\tau = \mathcal{O}(h)$ .*

*Then, for a linear DG scheme the norm of the residual  $\mathbf{R}$  is bounded from below by terms of order  $h^\gamma$  even if the error of the method is  $\mathcal{O}(h^{1+\gamma})$ .*

*D, Giesselmann: A posteriori analysis of fully discrete method of lines DG schemes for systems of conservation laws, submitted*

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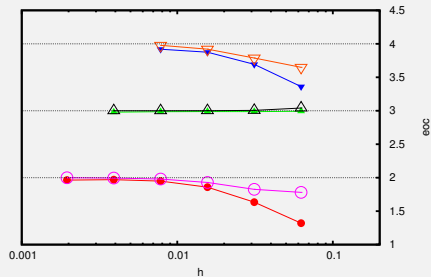
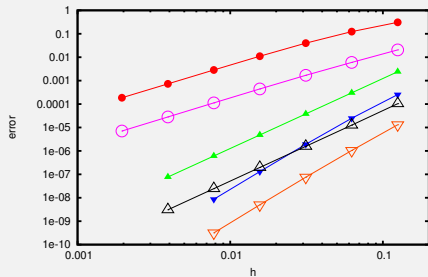
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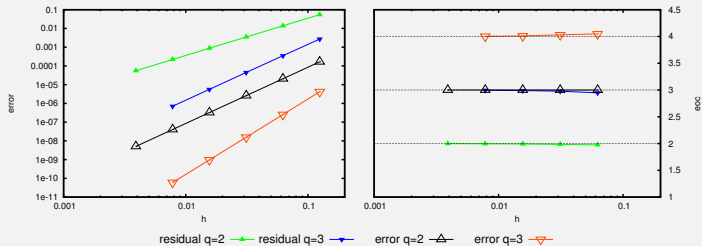
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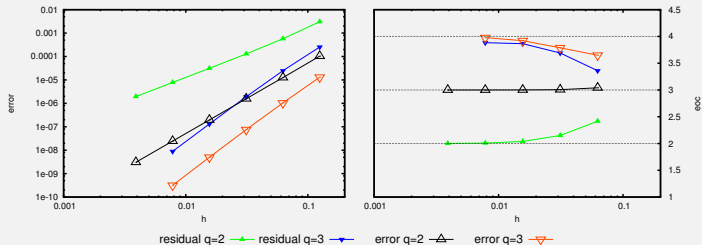


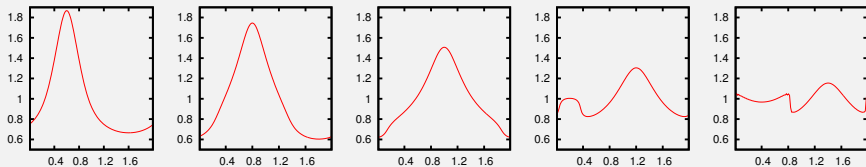
residual q=1 —●— residual q=2 —▲— residual q=3 —▼— error q=1 —○— error q=2 —△— error q=3 —▽—

## ... sub optimal flux

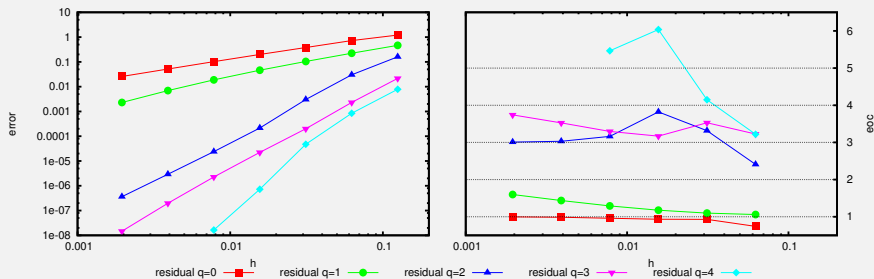


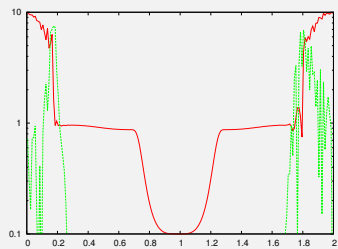
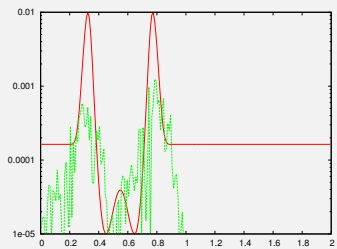
## ... optimal flux but lower temporal reconstruction





Density evolution:  $t = 0.6, 0.8, 1.0, 1.2, 1.4$ .







# A-posteriori estimator based on RE:

## Some remarks

- works now for fully discrete scheme
- we are working on extending it to higher dimensions
- only works with smooth solutions:  
For grid adaptivity and stabilization this is not really an issue?
- needs very specific flux function:  
We can show that this is required  
With general flux the Residual is not optimal
- expensive to compute:  
using a simpler spatial reconstruction again lowers the order by one  
For grid adaptivity and stabilization this is not really an issue?