

Uniform asymptotic preserving and well-balanced schemes for hyperbolic systems with source terms

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Mathematical and physical context

AP scheme for the P_1 model

Extension to the Euler model

Mathematic and physical context

Stiff hyperbolic systems

- **Stiff hyperbolic system with source terms:**

$$\partial_t \mathbf{U} + \frac{1}{\varepsilon} \partial_x F(\mathbf{U}) + \frac{1}{\varepsilon} \partial_y G(\mathbf{U}) = \frac{1}{\varepsilon} S(\mathbf{U}) - \frac{\sigma}{\varepsilon^2} R(\mathbf{U}), \quad \mathbf{U} \in \mathbb{R}^n$$

with $\varepsilon \in]0, 1]$ et $\sigma > 0$.

- Subset of solutions given by the balance between the source terms and the convective part:

- **Diffusion solutions** for $\varepsilon \rightarrow 0$ and $S(\mathbf{U}) = 0$:

$$\partial_t \mathbf{V} - \operatorname{div} (K(\nabla \mathbf{V}, \sigma)) = 0, \quad \mathbf{V} \in \operatorname{Ker} R.$$

- **Steady-state** for $\sigma = 0$ et $\varepsilon \rightarrow 0$:

$$\partial_x F(\mathbf{U}) + \partial_y G(\mathbf{U}) = S(\mathbf{U}).$$

- Applications: biology, neutron transport, fluid mechanics, plasma physics, Radiative hydrodynamic (hydrodynamic + linear transport of photon).

Notion of WB and AP schemes

- Acoustic equation with damping and gravity:

$$\begin{cases} \partial_t p + \frac{1}{\varepsilon} \partial_x u = 0, \\ \partial_t u + \frac{1}{\varepsilon} \partial_x p = -\frac{1}{\varepsilon} g - \frac{\sigma}{\varepsilon^2} u, \end{cases} \quad \longrightarrow \quad \partial_t p - \partial_x \left(\frac{1}{\sigma} (\partial_x p + g) \right) = 0.$$

- Steady-state: $u = 0$, $\partial_x p = -g$.
- **Godunov-type** schemes give an error homogeneous to $O(\Delta x)$.
- For nearly uniform flows, spurious velocities larger than physical velocity.
- **Important deviation of the steady-state.**
- **WB scheme:** discretize the steady-state **exactly of with high accuracy.**
- Ref: S. Jin, *A steady-state capturing method for hyperbolic method with geometrical source terms.*
- To construct WB and AP schemes: **incorporate the source in the fluxes** to capture the balance between source and convective terms.
- Consistency of **Godunov-type** schemes: $O\left(\frac{\Delta x}{\varepsilon} + \Delta t\right)$.
- CFL condition: $\Delta t \left(\frac{1}{\Delta x \varepsilon} + \frac{\sigma}{\varepsilon^2} \right) \leq 1$.
- Consistency of AP schemes: $O(\Delta x + \Delta t)$.
- CFL condition: degenerate on **parabolic CFL** at the limit.
- Ref: S. Jin, D. Levermore *Numerical schemes for hyperbolic conservation laws with stiff relaxation.*

Reduced bibliography

- 1D asymptotic preserving schemes
 - S. Jin, D. Levermore, *Numerical schemes for hyperbolic conservation laws with stiff relaxation terms*, (1996).
 - C. Berthon, R. Turpault, *Asymptotic preserving HLL schemes*, (2011).
 - L. Gosse, G. Toscani, *An asymptotic-preserving well-balanced scheme for the hyperbolic heat equations*, (2002).
 - C. Berthon, P. Charrier and B. Dubroca, *An HLLC scheme to solve the M_1 model of radiative transfer in two space dimensions*, (2007).
 - C. Chalons, M. Girardin, S. Kokh, *Large time step asymptotic preserving numerical schemes for the gas dynamics equations with source terms*, (2013).
- Well balanced schemes for chemotaxis and Euler equations
 - R. Natalini and M. Ribot, *An asymptotic high order mass-preserving scheme for a hyperbolic model of chemotaxis*, (2012).
 - V. Desveaux, M. Zenk, C. Berthon, C. Klingenberg, *A well-balanced scheme to capture non-explicit steady states in the Euler equations with gravity*, (2015).
 - J. Greenberg, A. Y. Leroux, *A well balanced scheme for the numerical processing of source terms in hyperbolic equations*, (1996).
 - R. Kappeli, S. Mishra, *Well-balanced schemes for the Euler equations with gravitation*, (2013).
- 2D asymptotic preserving schemes
 - A. Duran, F. Marche, R. Turpault, C. Berthon, *Asymptotic preserving scheme for the shallow water equations with source terms on unstructured meshes*, (2015).
 - C. Berthon, G. Moebs, C. Sarazin-Desbois and R. Turpault, *An AP scheme for systems of conservation laws with source terms on 2D unstructured meshes*, (2014).

Exemple of AP and WB Godunov schemes

- **Jin-Levermore (or Gosse-Toscani) scheme.** Plug the balance law $\partial_x E p = -\frac{\sigma}{\varepsilon} u + O(\varepsilon^2)$ in the fluxes. We write

$$p(x_j) = p(x_{j+\frac{1}{2}}) + (x_j - x_{j+\frac{1}{2}}) \partial_x p(x_{j+\frac{1}{2}})$$

$$p(x_j) = p(x_{j+\frac{1}{2}}) - (x_j - x_{j+\frac{1}{2}}) \frac{\sigma}{\varepsilon} u(x_{j+\frac{1}{2}})$$

Coupling the previous relation (and the same for x_{j+1}) with the fluxes

$$\begin{cases} u_j + p_j = u_{j+\frac{1}{2}} + p_{j+\frac{1}{2}} + \frac{\sigma \Delta x}{2\varepsilon} u_{j+\frac{1}{2}}, \\ u_{j+1} - p_{j+1} = u_{j+\frac{1}{2}} - p_{j+\frac{1}{2}} + \frac{\sigma \Delta x}{2\varepsilon} u_{j+\frac{1}{2}}. \end{cases}$$

- To finish, we take the following source term $\frac{1}{2}(u_{j+\frac{1}{2}} + u_{j-\frac{1}{2}})$.

Gosse-Toscani scheme:

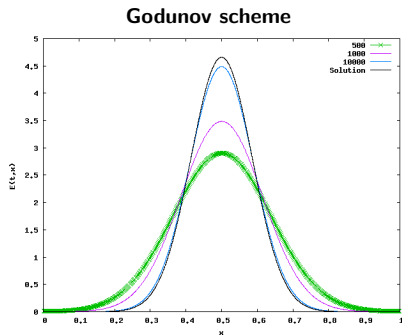
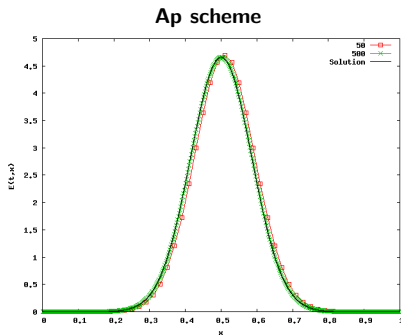
$$\begin{cases} \frac{p_j^{n+1} - p_j^n}{\Delta t} + M \frac{u_{j+1}^n - u_{j-1}^n}{2\varepsilon \Delta x} - M \frac{p_{j+1}^n - 2p_j^n + p_{j-1}^n}{2\varepsilon \Delta x} = 0, \\ \frac{u_j^{n+1} - u_j^n}{\Delta t} + M \frac{p_{j+1}^n - p_{j-1}^n}{2\varepsilon \Delta x} - M \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{2\varepsilon \Delta x} + M \frac{\sigma}{\varepsilon^2} u_j^n = 0, \end{cases}$$

with $M = \frac{2\varepsilon}{2\varepsilon + \sigma \Delta x}$.

- Consistency error of the **Gosse-Toscani** scheme: $O(\Delta x + \Delta t)$.
- Explicit CFL: $\Delta t \left(\frac{1}{\Delta x \varepsilon} \right) \leq 1$, Semi-implicit CFL: $\Delta t \left(\frac{1}{\Delta x \varepsilon + \Delta x^2} \right) \leq 1$.

Numerical example

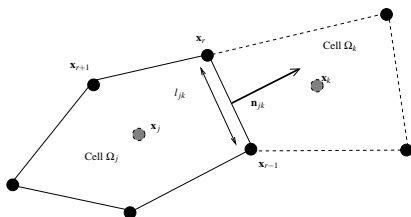
- Validation test for the AP scheme: the data are $p(0, x) = G(x)$ with $G(x)$ a Gaussian $u(0, x) = 0$ and $\sigma = 1$, $\varepsilon = 0.001$.



Scheme	L^1 error	CPU time
Godunov, 10000 cells	0.0366	1485m4.26s
Godunov, 500 cells	0.445	0m24.317s
AP, 500 cells	0.0001	0m15.22s
AP, 50 cells	0.0065	0m0.054s

Schémas "Asymptotic preserving" 2D

- **Classical extension in 2D of the Jin-Levermore scheme** : modify the upwind fluxes (1D fluxes write in the normal direction) plugging the steady-state in the fluxes.



- l_{jk} and \mathbf{n}_{jk} the normal and length associated with the edge $\partial\Omega_{jk}$.

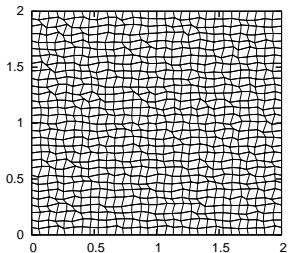
Asymptotic limit of the hyperbolic scheme:

$$|\Omega_j| \partial_t p_j(t) - \frac{1}{\sigma} \sum_k l_{jk} \frac{p_k^n - p_j^n}{d(\mathbf{x}_j, \mathbf{x}_k)} = 0.$$

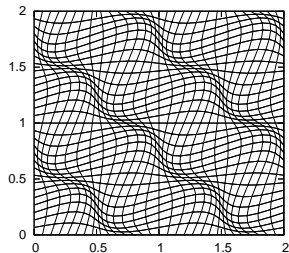
- $\|P_h^0 - P_h\| \rightarrow 0$ only on strong geometrical conditions.
- **Additional difficulty in 2D**: The basic extension of AP schemes **do not converge** on 2D general meshes $\forall \varepsilon$.

Example of unstructured meshes

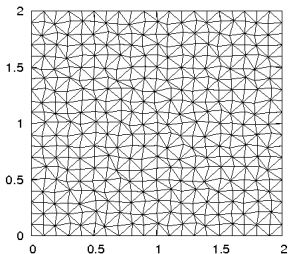
Random mesh



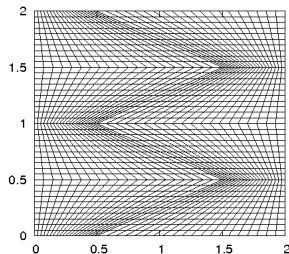
Collela mesh



Random triangular mesh



Kershaw mesh



AP scheme for the P_1 model

Nodal scheme : linear case

- Linear case: P_1 model

$$\begin{cases} \partial_t p + \frac{1}{\varepsilon} \operatorname{div}(\mathbf{u}) = 0, \\ \partial_t \mathbf{u} + \frac{1}{\varepsilon} \nabla p = -\frac{\sigma}{\varepsilon^2} \mathbf{u}. \end{cases} \quad \longrightarrow \quad \partial_t p - \operatorname{div} \left(\frac{1}{\sigma} \nabla p \right) = 0.$$

Idea:

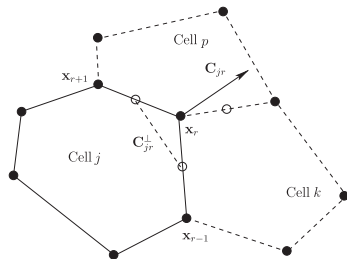
Nodal finite volume methods for P_1 model + AP and WB method.

Nodal schemes:

The fluxes are localized at the nodes of the mesh (for the classical scheme this is at the edge).

- Nodal geometrical quantities $\mathbf{C}_{jr} = \nabla_{\mathbf{x}_r} |\Omega_j|$.
- $\sum_j \mathbf{C}_{jr} = \sum_r \mathbf{C}_{jr} = \mathbf{0}$.

Notations



Nodal AP schemes

$$\begin{cases} |\Omega_j| \partial_t p_j(t) + \frac{1}{\varepsilon} \sum_r (\mathbf{u}_r, \mathbf{C}_{jr}) = 0, \\ |\Omega_j| \partial_t \mathbf{u}_j(t) + \frac{1}{\varepsilon} \sum_r \mathbf{p} \mathbf{C}_{jr} = \mathbf{S}_j. \end{cases}$$

- Classical nodal fluxes:

$$\begin{cases} \mathbf{p} \mathbf{C}_{jr} - p_j \mathbf{C}_{jr} = \hat{\alpha}_{jr} (\mathbf{u}_j - \mathbf{u}_r), \\ \sum_j \mathbf{p} \mathbf{C}_{jr} = \mathbf{0}, \end{cases}$$

with $\hat{\alpha}_{jr} = \frac{\mathbf{C}_{jr} \otimes \mathbf{C}_{jr}}{\|\mathbf{C}_{jr}\|}$.

- New fluxes obtained plugging steady-state $\nabla p = -\frac{\sigma}{\varepsilon} \mathbf{u}$ in the fluxes:

$$\begin{cases} \mathbf{p} \mathbf{C}_{jr} - p_j \mathbf{C}_{jr} = \hat{\alpha}_{jr} (\mathbf{u}_j - \mathbf{u}_r) - \frac{\sigma}{\varepsilon} \hat{\beta}_{jr} \mathbf{u}_r, \\ \left(\sum_j \hat{\alpha}_{jr} + \frac{\sigma}{\varepsilon} \sum_j \hat{\beta}_{jr} \right) \mathbf{u}_r = \sum_j p_j \mathbf{C}_{jr} + \sum_j \hat{\alpha}_{jr} \mathbf{u}_j. \end{cases}$$

with $\hat{\beta}_{jr} = \mathbf{C}_{jr} \otimes (\mathbf{x}_r - \mathbf{x}_j)$.

- Source term: (1) $\mathbf{S}_j = -\frac{\sigma}{\varepsilon^2} |\Omega_j| \mathbf{u}_j$ ou (2) $\mathbf{S}_j = -\frac{\sigma}{\varepsilon^2} \sum_r \hat{\beta}_{jr} \mathbf{u}_r$, $\sum_r \hat{\beta}_{jr} = \hat{l}_d |\Omega_j|$.
- Using the second source term and rewriting the scheme we obtain an **local semi implicit scheme with a CFL independent of ε** .

Assumptions for the convergence proof

Geometrical assumptions

- $(\mathbf{u}, \left(\sum_r \frac{\mathbf{c}_{jr} \otimes \mathbf{c}_{jr}}{|\mathbf{c}_{jr}|}\right) \mathbf{u}) \geq \beta h(\mathbf{u}, \mathbf{u}),$
- $(\mathbf{u}, \left(\sum_j \frac{\mathbf{c}_{jr} \otimes \mathbf{c}_{jr}}{|\mathbf{c}_{jr}|}\right) \mathbf{u}) \geq \gamma h(\mathbf{u}, \mathbf{u}),$
- $(\mathbf{u}, \left(\sum_j \mathbf{C}_{jr} \otimes (\mathbf{x}_r - \mathbf{x}_j)\right) \mathbf{u}) \geq \alpha h^2(\mathbf{u}, \mathbf{u}).$

- First and second assumptions: true on all non degenerated meshes.
- Last assumption: we have obtained sufficient but not necessary conditions on the meshes to satisfy this assumption.
- Example for triangles: all the angles must be larger than 12 degrees.

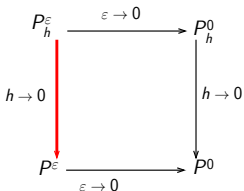
Assumption on regularity and initial data

- $\mathbf{u}(t = 0, \mathbf{x}) = -\frac{\varepsilon}{\sigma} \nabla p(t = 0, \mathbf{x})$
- Regularity for exact data: $\mathbf{V}(t, \mathbf{x}) \in H^4(\Omega)$
- Regularity for initial data of the scheme: $\mathbf{V}_h(t = 0, \mathbf{x}) \in L^2(\Omega)$

Uniform convergence in space

- Naive convergence estimate : $\|P_h^\varepsilon - P^\varepsilon\|_{naive} \leq C\varepsilon^{-b}h^c$
- **Idea:** use triangular inequalities and AP diagram (Jin-Levermore-Golse).

$$\|P_h^\varepsilon - P^\varepsilon\|_{L^2} \leq \min(\|P_h^\varepsilon - P^\varepsilon\|_{naive}, \|P_h^\varepsilon - P_h^0\| + \|P_h^0 - P^0\| + \|P^\varepsilon - P^0\|)$$



- Intermediary estimations :

- $\|P^\varepsilon - P^0\| \leq C_a \varepsilon^a,$
- $\|P_h^0 - P^0\| \leq C_d h^d,$
- $\|P_h^\varepsilon - P_h^0\| \leq C_e \varepsilon^e,$
- $d \leq c, e \geq a.$

- We obtain:

$$\|P_h^\varepsilon - P^\varepsilon\|_{L^2} \leq C \min(\varepsilon^{-b}h^c, \varepsilon^a + h^d + \varepsilon^e)$$

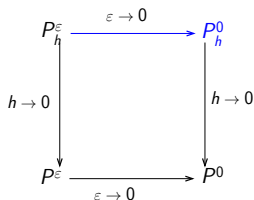
- Comparing ε and $\varepsilon_{threshold} = h^{\frac{ac}{a+b}}$ we obtain the final estimation:

$$\|P_h^\varepsilon - P^\varepsilon\|_{L^2} \leq h^{\frac{ac}{a+b}}$$

Diffusion scheme

Limit diffusion scheme (P_h^0)

$$\left\{ \begin{array}{l} |\Omega_j| \partial_t p_j(t) - \sum_r (\mathbf{u}_r, \mathbf{C}_{jr}) = 0, \\ \sum_r \hat{\alpha}_{jr} \mathbf{u}_j = \sum_r \hat{\alpha}_{jr} \mathbf{u}_r, \\ \sigma A_r \mathbf{u}_r = \sum_j p_j \mathbf{C}_{jr}, \quad A_r = - \sum_j \mathbf{C}_{jr} \otimes (\mathbf{x}_r - \mathbf{x}_j). \end{array} \right.$$



- **Problem:** estimate $\|P_h^\epsilon - P_h^0\|$.
- In practice, we have obtained $\|P_h^\epsilon - P_h^0\| \leq C \frac{\epsilon}{h}$.

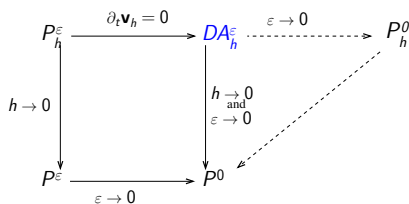
Condition H:

The discrete Hessian of P_h^0 can be bounded or the error estimate $\|P_h^\epsilon - P_h^0\|$ can be obtained independently of the discrete Hessian.

Diffusion scheme

Limit diffusion scheme (P_h^0)

$$\left\{ \begin{array}{l} |\Omega_j| \partial_t p_j(t) - \sum_r (\mathbf{u}_r, \mathbf{C}_{jr}) = 0, \\ \sum_r \hat{\alpha}_{jr} \mathbf{u}_j = \sum_r \hat{\alpha}_{jr} \mathbf{u}_r, \\ \sigma A_r \mathbf{u}_r = \sum_j p_j \mathbf{C}_{jr}, \quad A_r = - \sum_j \mathbf{C}_{jr} \otimes (\mathbf{x}_r - \mathbf{x}_j). \end{array} \right.$$



- **Problem:** estimate $\|P_h^\epsilon - P_h^0\|$.
- In practice, we have obtained $\|P_h^\epsilon - P_h^0\| \leq C \frac{\epsilon}{h}$.
- Introduction of an **intermediary diffusion scheme** DA_h^ϵ .
- DA_h^ϵ : P_h^ϵ scheme with $\partial_t \mathbf{F}_j = \mathbf{0}$.
- In the previous estimation we replace P_h^0 by DA_h^ϵ .

Condition H:

The discrete Hessian of P_h^0 can be bounded or the error estimate $\|P_h^\epsilon - P_h^0\|$ can be obtained independently of the discrete Hessian.

Final results

Space result:

We assume that the assumptions are verified. There exist $C(T) > 0$ such that:

$$\|\mathbf{V}^\varepsilon - \mathbf{V}_h^\varepsilon\|_{L^2([0,T] \times \Omega)} \leq C f(h, \varepsilon) \|p_0\|_{H^4(\Omega)} \leq Ch^{\frac{1}{4}} \|p_0\|_{H^4(\Omega)}$$

with

$$f(h, \varepsilon) = \min \left(\sqrt{\frac{h}{\varepsilon}}, \varepsilon \max \left(1, \sqrt{\frac{\varepsilon}{h}} \right) + h + (h + \varepsilon) + \varepsilon \right)$$

- Case $\varepsilon \leq h$: $\|\mathbf{V}^\varepsilon - \mathbf{V}_h^\varepsilon\| \leq C_1 \min(\sqrt{\frac{\varepsilon}{h}}, 1) \leq C_1 h$
- Case $\varepsilon \geq h$: $\|\mathbf{V}^\varepsilon - \mathbf{V}_h^\varepsilon\| \leq C_1 \min(\sqrt{\frac{h}{\varepsilon}}, \sqrt{\frac{\varepsilon^3}{h}})$
- Introducing $\varepsilon_{\text{thresh}} = h^{\frac{1}{2}}$ we prove that **the worst case is $\|\mathbf{V}^\varepsilon - \mathbf{V}_h^\varepsilon\| \leq C_2 h^{\frac{1}{4}}$** .

Space-time result:

We assume that the assumptions are verified. There exist $C > 0$ such that:

$$\|\mathbf{V}^\varepsilon(t_n) - \mathbf{V}_h^\varepsilon(t_n)\|_{L^2(\Omega)} \leq C (f(h, \varepsilon) + \Delta t^2) \|p_0\|_{H^4(\Omega)}$$

Remark: The condition H is not satisfied. The diffusion scheme used is DA_ε .

Intermediary results I

Estimation of $\|\mathbf{V}^\varepsilon - \mathbf{V}_h^\varepsilon\|$:

We assume that assumptions are verified. There exist $C > 0$ such that:

$$\|\mathbf{V}_h^\varepsilon - \mathbf{V}^\varepsilon\|_{L^\infty((0,T);L^2(\Omega))} \leq C \sqrt{\frac{h}{\varepsilon}}.$$

■ Principle of proof:

- Control the stability of the discrete quantities \mathbf{u}_r and \mathbf{u}_j by ε
- We define the error $E(t) = \|\mathbf{V}^\varepsilon - \mathbf{V}_h^\varepsilon\|_{L^2}$ and we estimate $E'(t)$ using Young and Cauchy-Schwartz inequalities, stability estimates and integration in time.

Estimation of $\|DA_h^\varepsilon - P^0\|$:

We assume that the assumptions are verified. There exist $C_1 > 0$ such that:

$$\|\mathbf{V}_h^0 - \mathbf{V}^0\|_{L^2(\Omega)} \leq C_1(T)(h + \varepsilon), \quad 0 < t \leq T.$$

■ Principle of proof:

- Control the stability of the discrete quantities $\nabla_r E$ and E_j .
- Consistency study of Div and Grad discrete operators.
- L^2 estimate using consistency error and Gronwall lemma.

Estimate $\|P_h^\varepsilon - DA_h^\varepsilon\|$:

We assume that the assumptions are verified. There exist $C_2(T) > 0$ such that:

$$\|\mathbf{V}_h^\varepsilon - \mathbf{V}_h\|_{L^2(\Omega)} \leq C_2(T)\varepsilon \max\left(1, \sqrt{\varepsilon h^{-1}}\right) + Ch, \quad 0 < t \leq T.$$

Estimate $\|P^\varepsilon - P^0\|$:

We assume that the assumptions are verified. There exist $C_3(T) > 0$ such that:

$$\|\mathbf{V}^\varepsilon - \mathbf{V}^0\|_{L^2(\Omega)} \leq C_3(T)\varepsilon, \quad 0 < t \leq T.$$

■ Principle of proof:

- Write $P^0 = P^\varepsilon + R$ (resp $DA_h^\varepsilon = P_h^\varepsilon + R$) with R a residue.
- Find a bound with ε of the residue.
- L^2 estimate of the difference between the two models and between the two schemes.

Analysis of AP schemes: modified equations

- To understand the behavior of the scheme, we use the modified equations method.
- The modified equation associated with the Upwind scheme is
- The modified equation associated to the Gosse-Toscani scheme is

$$\begin{cases} \partial_t p + \frac{1}{\varepsilon} \partial_x u - \frac{\Delta x}{2\varepsilon} \partial_{xx} p = 0, \\ \partial_t u + \frac{1}{\varepsilon} \partial_x p - \frac{\Delta x}{2\varepsilon} \partial_{xx} u = -\frac{\sigma}{\varepsilon^2} u. \end{cases}$$

$$\begin{cases} \partial_t p + M \frac{1}{\varepsilon} \partial_x u - M \frac{\Delta x}{2\varepsilon} \partial_{xx} p = 0, \\ \partial_t u + M \frac{1}{\varepsilon} \partial_x p - M \frac{\Delta x}{2\varepsilon} \partial_{xx} u = -M \frac{\sigma}{\varepsilon^2} u. \end{cases}$$

- Plugging $\varepsilon \partial_x p + O(\varepsilon^2) = -\sigma u$ in the first equation, we obtain the diffusion limit

$$\partial_t p - \frac{1}{\sigma} \partial_{xx} p - \frac{\Delta x}{2\varepsilon} \partial_{xx} p = 0.$$

- Plugging $M\varepsilon \partial_x p + O(\varepsilon^2) = -M\sigma u$ in the first equation, we obtain the diffusion limit

$$\partial_t p - \frac{M}{\sigma} \partial_{xx} p - \frac{1-M}{\sigma} \partial_{xx} p = 0$$

- **Conclusion:** the regime is captured only on fine grids.

- **Conclusion:** the regime is captured only on all grids.

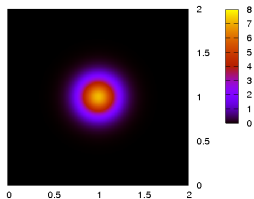
Construction of the AP scheme in 2D

- We must modify the viscosity to a consistent diffusion scheme with the good coefficient on coarse grids.
- We must **also discretize correctly** the source term and the gradient of pressure to obtain a consistent diffusion scheme on fine grids (WB schemes).

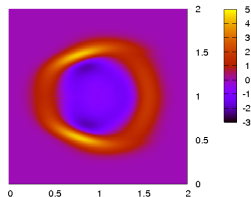
AP scheme vs classical scheme

- Test case: heat fundamental solution. Results for different hyperbolic scheme with $\varepsilon = 0.001$ on Kershaw mesh.

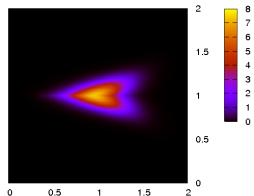
Diffusion solution



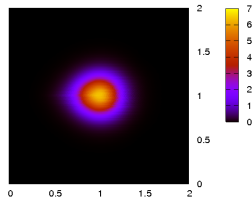
Non AP scheme



Standard AP scheme

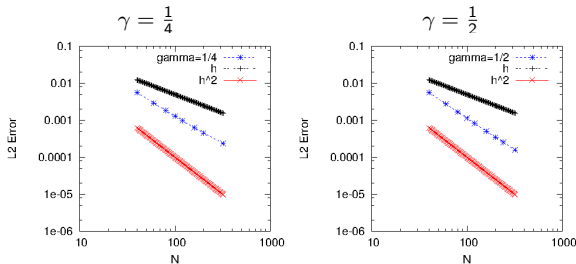


Nodal AP scheme



Uniform convergence

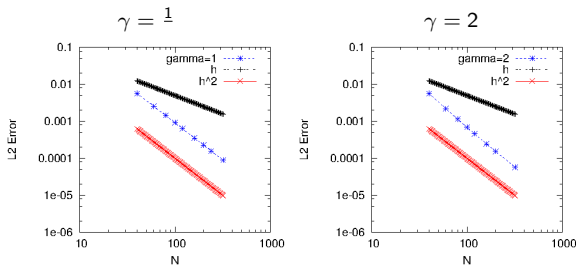
- ε dependent periodic solution for the P_1 model.
- $p(t, \mathbf{x}) = (\alpha(t) + \frac{\varepsilon^2}{\sigma} \alpha'(t)) \cos(\pi x) \cos(\pi y)$
- $\mathbf{u}(t, \mathbf{x}) = (-\frac{\varepsilon}{\sigma} \alpha(t) \sin(\pi x) \cos(\pi y), -\frac{\varepsilon}{\sigma} \alpha(t) \sin(\pi y) \cos(\pi x))$
- Convergence study for $\varepsilon = h^\gamma$ on random mesh.



- Numerical results show that the error is homogenous to $O(h\varepsilon + h^2)$.
- Theoretical estimate that we can hope: $O((h\varepsilon)^{\frac{1}{2}} + h)$.
- Non optimal estimation in the intermediary regime.

Uniform convergence

- ε dependent periodic solution for the P_1 model.
- $p(t, \mathbf{x}) = (\alpha(t) + \frac{\varepsilon^2}{\sigma} \alpha'(t)) \cos(\pi x) \cos(\pi y)$
- $\mathbf{u}(t, \mathbf{x}) = (-\frac{\varepsilon}{\sigma} \alpha(t) \sin(\pi x) \cos(\pi y), -\frac{\varepsilon}{\sigma} \alpha(t) \sin(\pi y) \cos(\pi x))$
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Extension to the Euler model

Euler equation with external forces

- Euler equation with gravity and friction:

$$\left\{ \begin{array}{l} \partial_t \rho + \frac{1}{\varepsilon^\alpha} \operatorname{div}(\rho \mathbf{u}) = 0, \\ \partial_t \rho \mathbf{u} + \frac{1}{\varepsilon^\alpha} \operatorname{div}(\rho \mathbf{u} \otimes \mathbf{u}) + \frac{1}{\varepsilon^\alpha} \nabla p = -\frac{1}{\varepsilon^\alpha} (\rho \nabla \phi + \frac{\sigma}{\varepsilon^\beta} \rho \mathbf{u}), \\ \partial_t \rho e + \frac{1}{\varepsilon^\alpha} \operatorname{div}(\rho e \mathbf{u}) + \operatorname{div}(\rho \mathbf{u}) = -\frac{1}{\varepsilon^\alpha} (\rho (\nabla \phi, \mathbf{u}) + \frac{\sigma}{\varepsilon^\beta} \rho (\mathbf{u}, \mathbf{u})). \end{array} \right.$$

- with ϕ the gravity potential, σ the friction coefficient.

Subset of solutions :

- Hydrostatic Steady-state ($\alpha = 1, \beta = 0$):

$$\left\{ \begin{array}{l} \mathbf{u} = \mathbf{0}, \\ \nabla p = -\rho \nabla \phi. \end{array} \right.$$

- High friction limit ($\alpha = 0, \beta = 1$), no gravity: $\mathbf{u} = \mathbf{0}$
- Diffusion limit ($\alpha = 1, \beta = 1$):

$$\left\{ \begin{array}{l} \partial_t \rho + \operatorname{div}(\rho \mathbf{u}) = 0, \\ \partial_t \rho e + \operatorname{div}(\rho e \mathbf{u}) + p \operatorname{div} \mathbf{u} = 0, \\ \mathbf{u} = -\frac{1}{\sigma} \left(\nabla \phi + \frac{1}{\rho} \nabla p \right). \end{array} \right.$$

Design of AP nodal scheme I

Idea :

Modify the Lagrange+remap classical scheme with the Jin-Levermore method

- Classical Lagrange+remap scheme (LP scheme):

$$\left\{ \begin{array}{l} |\Omega_j| \partial_t \rho_j + \frac{1}{\varepsilon^\alpha} \left(\sum_{R_+} \mathbf{u}_{jr} \rho_j + \sum_{R_-} \mathbf{u}_{jr} \rho_{k(r)} \right) = 0 \\ |\Omega_j| \partial_t \rho_j \mathbf{u}_j + \frac{1}{\varepsilon^\alpha} \left(\sum_{R_+} \mathbf{u}_{jr} (\rho \mathbf{U})_j + \sum_{R_-} \mathbf{u}_{jr} (\rho \mathbf{U})_{k(r)} + \sum_r \mathbf{p} \mathbf{C}_{jr} \right) = 0 \\ |\Omega_j| \partial_t \rho_j \mathbf{e}_j + \frac{1}{\varepsilon^\alpha} \left(\sum_{R_+} \mathbf{u}_{jr} (\rho \mathbf{e})_j + \sum_{R_-} \mathbf{u}_{jr} (\rho \mathbf{e})_{k(r)} + \sum_r (\mathbf{p} \mathbf{C}_{jr}, \mathbf{u}_r) \right) = 0 \end{array} \right.$$

with Lagrangian fluxes

$$\left\{ \begin{array}{l} \mathbf{G}_{jr} = \rho_j \mathbf{C}_{jr} + \rho_j c_j \hat{\alpha}_{jr} (\mathbf{u}_j - \mathbf{u}_r) \\ \sum_j \rho_j c_j \hat{\alpha}_{jr} \mathbf{u}_r = \sum_j \rho_j \mathbf{C}_{jr} + \sum_j \rho_j c_j \hat{\alpha}_{jr} \mathbf{u}_j \end{array} \right.$$

- Advection fluxes: $\mathbf{u}_{jr} = (\mathbf{C}_{jr}, \mathbf{u}_r)$, $R_+ = (r/\mathbf{u}_{jr} > 0)$, $R_- = (r/\mathbf{u}_{jr} < 0)$ et

$$\rho_{k(r)} = \frac{\sum_{j/\mathbf{u}_{jr} > 0} \mathbf{u}_{jr} \rho_j}{\sum_{j/\mathbf{u}_{jr} > 0} \mathbf{u}_{jr}}.$$

Design of AP nodal scheme II

Jin Levermore method:

Plug the relation $\nabla p + O(\varepsilon^2) = -\rho \nabla \phi - \frac{\sigma}{\varepsilon} \rho \mathbf{u}$ in the Lagrangian fluxes

- The modified scheme is given by

$$\left\{ \begin{array}{l} |\Omega_j| \partial_t \rho_j + \frac{1}{\varepsilon \alpha} \left(\sum_{R_+} \mathbf{u}_{jr} \rho_j + \sum_{R_-} \mathbf{u}_{jr} \rho_{k(r)} \right) = 0 \\ |\Omega_j| \partial_t \rho_j \mathbf{u}_j + \frac{1}{\varepsilon \alpha} \left(\sum_{R_+} \mathbf{u}_{jr} (\rho \mathbf{u})_j + \sum_{R_-} \mathbf{u}_{jr} (\rho \mathbf{u})_{k(r)} + \sum_r \mathbf{p} \mathbf{C}_{jr} \right) \\ = -\frac{1}{\varepsilon \alpha} \left(\sum_r \hat{\beta}_{jr} (\rho \nabla \phi)_r + \frac{\sigma}{\varepsilon \beta} \sum_r \rho_r \hat{\beta}_{jr} \mathbf{u}_r \right) \\ |\Omega_j| \partial_t \rho_j + \frac{1}{\varepsilon \alpha} \left(\sum_{R_+} \mathbf{u}_{jr} (\rho \varepsilon)_j + \sum_{R_-} \mathbf{u}_{jr} (\rho \varepsilon)_{k(r)} + \sum_r (\mathbf{p} \mathbf{C}_{jr}, \mathbf{u}_r) \right) \\ = -\frac{1}{\varepsilon \alpha} \left(\sum_r (\hat{\beta}_{jr} (\rho \nabla \phi)_r, \mathbf{u}_r) + \frac{\sigma}{\varepsilon \beta} \sum_r \rho_r (\mathbf{u}_r, \hat{\beta}_{jr} \mathbf{u}_r) \right) \end{array} \right.$$

with the new Lagrangian fluxes

$$\left\{ \begin{array}{l} \mathbf{p} \mathbf{C}_{jr} = \rho_j \mathbf{C}_{jr} + \rho_j c_j \hat{\alpha}_{jr} (\mathbf{u}_j - \mathbf{u}_r) - \hat{\beta}_{jr} (\rho \nabla \phi)_r - \frac{\sigma}{\varepsilon \beta} \rho_r \hat{\beta}_{jr} \mathbf{u}_r \\ \left(\sum_j \rho_j c_j \hat{\alpha}_{jr} + \frac{\sigma}{\varepsilon \beta} \rho_r \sum_j \hat{\beta}_{jr} \right) \mathbf{u}_r = \sum_j \rho_j \mathbf{C}_{jr} + \sum_j \rho_j c_j \hat{\alpha}_{jr} \mathbf{u}_j - \left(\sum_j \hat{\beta}_{jr} \right) (\rho \nabla \phi)_r \end{array} \right.$$

- and $(\rho \nabla \phi)_r$ a discretization of $\rho \nabla \phi$ at the interface.

Limit diffusion scheme:

If the local matrices are invertible then the LR-AP scheme tends to the following scheme

$$\begin{cases} |\Omega_j| \partial_t \rho_j + \left(\sum_{R_+} (\mathbf{C}_{jr}, \mathbf{u}_r) \rho_j + \sum_{R_-} (\mathbf{C}_{jr}, \mathbf{u}_r) \rho_{k(r)} \right) = 0 \\ |\Omega_j| \partial_t \rho_j + \left(\sum_{R_+} (\mathbf{C}_{jr}, \mathbf{u}_r) (\rho \mathbf{e})_j + \sum_{R_-} (\mathbf{C}_{jr}, \mathbf{u}_r) (\rho \mathbf{e})_{k(r)} + p_j \sum_r (\mathbf{C}_{jr}, \mathbf{u}_r) \right) = 0 \\ \sigma \rho_r \left(\sum_j \hat{\beta}_{jr} \right) \mathbf{u}_r = \sum_j p_j \mathbf{C}_{jr} - \left(\sum_j \hat{\beta}_{jr} \right) (\rho \nabla \phi)_r \end{cases}$$

- The nodal gradient formula $\nabla_r p = \left(\sum_j \hat{\beta}_{jr} \right)^{-1} \left(\sum_j p_j \mathbf{C}_{jr} \right)$ is **a consistent and convergent approximation of the gradient** on unstructured meshes (Consistency study+Gronwall's lemma).
- For $p = K\rho$, numerically the schemes converge at the first scheme.
- If we use a second order advection scheme for the remap part. The full scheme converges with the second order.
- **Open question:** Verify this for a non isothermal pressure law as perfect gas law.

Well balanced property

Well balanced property

- We define the discrete gradient $\nabla_r \rho = -(\sum_j \hat{\beta}_{jr})^{-1} \sum_j \rho_j \mathbf{C}_{jr}$ and ρ_r an average of ρ_j around \mathbf{x}_r .
- If the initial data are given by the discrete steady-state $\nabla_r \rho = -(\rho \nabla \phi)_r$, $\rho_j^{n+1} = \rho_j^n$, $\mathbf{u}_j^{n+1} = \mathbf{u}_j^n$ and $e_j^{n+1} = e_j^n$,
- **Remark:** The spatial error for a steady-state is only governed by the error between discrete steady-state and the continuous steady-state

High order reconstruction of steady-state

- **Aim:** Conserve the stability property of the first order scheme, but discretize the steady-state with a high order accuracy or exactly.
- **Method:** design high order discrete steady-state
- The discrete steady-state is given $(\sum_j \hat{\beta}_{jr})^{-1} \sum_j \rho_j \mathbf{C}_{jr} = -\rho_r (\sum_j \hat{\beta}_{jr})^{-1} \sum_j \phi_j \mathbf{C}_{jr}$.
- If ρ_r is an arithmetic average around a node r , this discrete steady-state is a second order approximation of the continuous one.

High order discretization of the steady-state

- To begin we consider the steady-state $\nabla p = -\rho \nabla \phi$
- we integrate on the dual cell Ω_r^* (volume V_r) to obtain

$$V_r \left(\frac{1}{V_r} \int_{\Omega_r^*} \nabla p(\mathbf{x}) \right) = -V_r \left(\frac{1}{V_r} \int_{\Omega_r^*} \rho(\mathbf{x}) \nabla \phi(\mathbf{x}) \right).$$

- We introduce 3 polynomials $\bar{\rho}_r(\mathbf{x})$ (order q), $\bar{p}_r(\mathbf{x})$ and $\bar{\phi}_r(\mathbf{x})$ ($q+1$ order) with

$$\int_{\Omega_r^*} \bar{\rho}_r(\mathbf{x}) = |\Omega_l| \rho_l, \quad \int_{\Omega_r^*} \bar{p}_r(\mathbf{x}) = |\Omega_l| p_l, \quad \int_{\Omega_r^*} \bar{\phi}_r(\mathbf{x}) = |\Omega_l| \phi_l$$

and $l \in S(r)$ ($S(r)$ a subset of cell around the node r).

- Now we incorporate this high-order reconstruction in the scheme. For this we need to have a pressure gradient which corresponds to the viscosity of the scheme.
- We obtain a **q-order steady-state**:

$$-\underbrace{\left(\sum_j \hat{\beta}_{jr} \right)^{-1} \sum_j p_j \mathbf{C}_{jr}}_{\nabla p_r} = -(\rho \nabla \phi)_r^{HO}$$

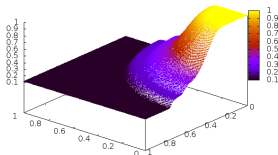
with

$$(\rho \nabla \phi)_r^{HO} = \frac{1}{V_r} \left(\left(\int_{\Omega_r^*} \nabla p(\mathbf{x}) \right) + \left(\int_{\Omega_r^*} \rho(\mathbf{x}) \nabla \phi(\mathbf{x}) \right) \right) + \left(\sum_j \hat{\beta}_{jr} \right)^{-1} \sum_j p_j \mathbf{C}_{jr}$$

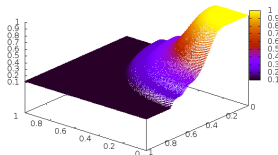
Numerical result : large opacity

- Test case: sod problem with $\sigma > 0$, $\varepsilon = 1$ and $\nabla\phi = \mathbf{0}$.
- $\sigma = 1$

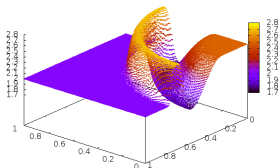
AP scheme, ρ



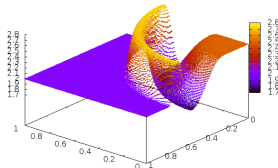
non-AP scheme, ρ



AP scheme, ϵ



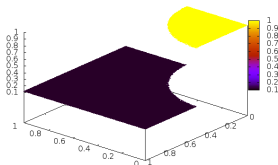
non-AP scheme, ϵ



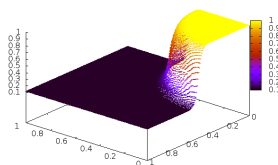
Numerical result : large opacity

- Test case: sod problem with $\sigma > 0$, $\varepsilon = 1$ and $\nabla\phi = \mathbf{0}$.
- $\sigma = 10^6$

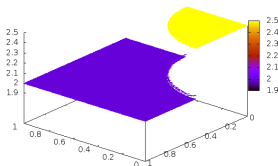
AP scheme, ρ



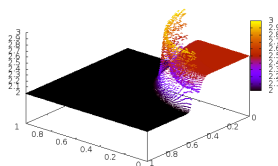
non-AP scheme, ρ



AP scheme, ϵ



non-AP scheme, ϵ



Result for steady-state

- **1D Steady-state:** $\rho(t, x) = 3 + 2 \sin(2\pi x)$, $u(t, x) = 0$
- $\rho(t, x) = 3 + 3 \sin(2\pi x) - \frac{1}{2} \cos(4\pi x)$ and $\phi(x) = -\sin(2\pi x)$. **Random 1D Grid.**

Cells	LR		LR-AP(2)		LR-AP O(3)		LR-AP O(4)	
	Error	q	Error	q	Error	q	Error	q
20	0.8335	-	0.0102	-	0.0079	-	0.0067	-
40	0.4010	1.05	0.0027	1.91	8.4E-4	3.23	1.5E-4	5.48
80	0.2065	0.96	7.0E-4	1.95	7.7E-5	3.45	4.1E-6	5.19
160	0.1014	1.02	1.7E-4	2.04	7.0E-6	3.46	1.0E-7	5.36

- **2D Steady-state:** $\rho(t, x) = e^{-x \cdot g}$, $u(t, x) = 0$, $p(t, x) = e^{-x \cdot g}$ and $\phi = (x, g)$.

	Cells	LR		LR-AP O(2)		LR-AP O(3)	
		Error	q	Error	q	Error	q
Cartesian Mesh	16 × 16	0.04132	1.07	0.00147	2.34	5.47E-6	3.8
	32 × 32	0.02013	1.04	3.28E-4	2.16	3.67E-7	3.9
	64 × 64	0.00993	1.02	7.65E-5	2.1	2.38E-8	3.95
	128 × 128	0.00493	1.01	1.90E-5	2.1	1.52E-9	3.96
Random Cartesian Mesh	16 × 16	0.05465	0.86	0.00155	2.7	8.25E-6	3.47
	32 × 32	0.02940	0.89	3.4E-4	2.18	7.55E-7	3.45
	64 × 64	0.01488	0.98	7.98E-5	2.09	8.5E-8	3.15
	128 × 128	0.00742	1.00	2.06E-5	1.95	2.37E-8	1.84

Result for steady-state

- **1D Steady-state:** $\rho(t, x) = 3 + 2 \sin(2\pi x)$, $u(t, x) = 0$
- $\rho(t, x) = 3 + 3 \sin(2\pi x) - \frac{1}{2} \cos(4\pi x)$ and $\phi(x) = -\sin(2\pi x)$. **Random 1D Grid.**

Cells	LR		LR-AP(2)		LR-AP O(3)		LR-AP O(4)	
	Error	q	Error	q	Error	q	Error	q
20	0.8335	-	0.0102	-	0.0079	-	0.0067	-
40	0.4010	1.05	0.0027	1.91	8.4E-4	3.23	1.5E-4	5.48
80	0.2065	0.96	7.0E-4	1.95	7.7E-5	3.45	4.1E-6	5.19
160	0.1014	1.02	1.7E-4	2.04	7.0E-6	3.46	1.0E-7	5.36

- **2D Steady-state:** $\rho(t, x) = e^{-x \cdot g}$, $u(t, x) = 0$, $p(t, x) = e^{-x \cdot g}$ and $\phi = (x, g)$.

	Cells	LR		LR-AP O(2)		LR-AP O(3)	
		Error	q	Error	q	Error	q
Collela Mesh	16 × 16	0.08902	0.45	0.00197	2.44	2.97E-5	1.9
	32 × 32	0.05725	0.63	5.9E-4	1.74	5.43E-6	2.45
	64 × 64	0.03232	0.82	1.6E-4	1.88	5.93E-7	3.19
	128 × 128	0.01711	0.92	4.5E-5	1.86	4.68E-8	3.66
Kershaw Mesh	16 × 16	0.08376	0.83	3.38E-4	2.36	6.13E-6	3.84
	32 × 32	0.04253	0.98	7.29E-5	2.24	3.97E-7	3.95
	64 × 64	0.02060	1.05	7.87E-5	2.13	2.03E-8	4.3
	128 × 128	0.00988	1.06	4.34E-6	1.9	1.77E-9	3.52

Conclusion and perspectives

■ Conclusion

- P_1 **model**: First AP scheme on unstructured meshes (now other schemes have been developed).
- P_1 **model**: Uniform proof of convergence on unstructured meshes in 1D and 2D for the implicit scheme.
- An extension for general Friedrich's systems have been also studied (algebraic micro-macro decomposition)
- **Euler model with external force**: AP schemes for the high friction regime.
- **Euler model with external force**: new high-order reconstruction of the hydrostatic steady-state.
- **Problem for all the schemes** : spurious mods in few cases (example: Cartesian mesh + Dirac Initial data).

■ Possible perspectives

- P_1 **model**: Theoretical study of the explicit and semi-implicit scheme (CFL independent of ε).
- **Euler model**: Entropy study for the AP-WB scheme.
- **Euler model**: Validate on analytic case the convergence of the diffusion scheme for nonlinear pressure law.
- Find a **generic procedure to stabilize the nodal schemes** (B. Després and E. Labourasse for the Lagrangian Euler equations).

- **Project:** "implicit scheme and preconditioning for radiative transfer" models" with Xavier Blanc, Emmanuel Labourasse + Master student ?

Transport equation (photonics neutronic):

- The distribution function $f(t, \mathbf{x}, \Omega)$ with Ω the direction, c the light speed satisfy

$$\partial_t f + c\Omega \cdot \nabla f = c\sigma \left(\int_{S^2} f d\Omega - f \right)$$

- The kinetic equations are approximated by linear hyperbolic P_n systems:

$$\partial_t \mathbf{U} + cA_x \partial_x \mathbf{U} + cA_y \partial_y \mathbf{U} + cA_z \partial_z \mathbf{U} = -c\sigma R \mathbf{U}$$

- Important regimes: free transport regime ($\sigma \rightarrow 0$) : exact transport of the solution and diffusion regime ($\sigma \rightarrow \infty$).
- **Problems for explicit scheme:** **Very large and stiff hyperbolic systems.** Stiff hyperbolic CFL for explicit schemes, Stiff parabolic CFL condition for the AP schemes.
- **Problems for implicit scheme:** the large hyperbolic system (bad structure) and the large ratio between wave velocities ($\{\lambda_{min}c, \dots, \lambda_{max}c\}$ with $\lambda_{min} \approx -1$, $\lambda_{max} \approx 1$).
- **Aim:** Test a physic-based preconditioning + GMRES for the P_1 model. Extend this preconditioning to the P_n models and the transport regime.

Thank you