

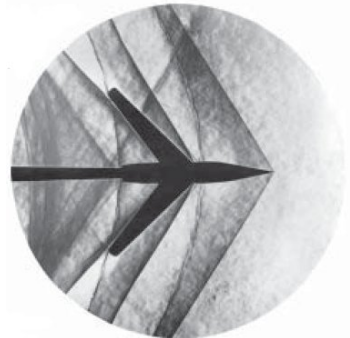
Specialized Numerical Methods for Transport Phenomena

The finite element method:
Hyperbolic and transient problems

Bruno Blais

Associate Professor
Department of Chemical Engineering
Polytechnique Montréal

February 12, 2024





Transient problems

- Motivation

- Euler's methods

- Alternatives

Advection-Diffusion problem

- Motivation

- Two toy problems

- Regular formulation

- On the need for stabilization

- Artificial diffusion

- Streamline artificial diffusion

- SUPG: Regaining consistency

Conclusions



Transient problems

- Motivation

- Euler's methods

- Alternatives

Advection-Diffusion problem

- Motivation

- Two toy problems

- Regular formulation

- On the need for stabilization

- Artificial diffusion

- Streamline artificial diffusion

- SUPG: Regaining consistency

Conclusions



So far, we have only investigated steady-state Poisson problems such as the heat transfer equation. In many situations, we are interested in solving transient problems. For example, the transient heat equation:

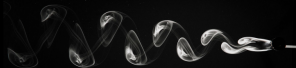
$$\rho C_p \frac{\partial T}{\partial t} = k \nabla^2 T \quad (1)$$

with an initial condition describing $T(\mathbf{x})$ and boundary conditions. We have an additional term that we need to consider, which is the time derivative.

There are two approaches we can consider:

- Interpolate in time using a Lagrange polynomial (space-time FEM)
- Use a finite difference approach. **This is generally the approach we take.**

Euler's method



You most likely have already seen at some place during your curriculum that one way to approximate time derivative is to use a finite difference scheme. For example:

$$\frac{\partial c}{\partial t} \approx \frac{c^{t+\Delta t} - c^t}{\Delta t} \quad (2)$$

is called Euler's method.

We generally distinguish between two families of Euler's method:

- Explicit: All terms at the right-hand-side are taken at time t
- Implicit: All terms at the right-hand side are taken at time $t + \Delta t$

An example



Let's consider the following system of ODEs:

$$\frac{dx}{dt} = \alpha x - \beta xy \quad (3)$$

$$\frac{dy}{dt} = \delta xy - \gamma y \quad (4)$$

Lets solve it using both implicit and explicit Euler

In the context of FEM



Solving transient problems in FEM is done in the same exact fashion. Let's take our PDE:

$$\rho C_p \frac{\partial T}{\partial t} = k \nabla^2 T \quad (5)$$

We discretize the equation in time:

$$\frac{T^{t+\Delta t} - T^t}{\Delta t} = k \nabla^2 T \quad (6)$$

Now let's see what happens when we solve it using explicit or implicit Euler.

Weak form: Implicit Euler



The weak form we obtain is:

$$\iiint_{\Omega} \left(\frac{\rho C_p}{\Delta t} u T^{t+\Delta t} + k \nabla u \nabla T^{t+\Delta t} \right) = \iiint_{\Omega} \left(\frac{\rho C_p}{\Delta t} u T^t \right) \quad (7)$$

And fully discretized it is:

$$\sum_j T_j^{t+\Delta t} \iiint_{\Omega} \left(\frac{\rho C_p}{\Delta t} \phi_i \phi_j + k \nabla \phi_i \nabla \phi_j \right) = \iiint_{\Omega} \left(\frac{\rho C_p}{\Delta t} \phi_i T^t \right) \quad (8)$$

We recognize two key blocks:

- $\phi_i \phi_j$: It is called the mass matrix. It models the inertia of a system.
- $\nabla \phi_i \nabla \phi_j$: It is called the stiffness matrix. This comes from solid mechanics.

Weak form: Explicit Euler



The weak form we obtain is:

$$\iiint_{\Omega} \left(\frac{\rho C_p}{\Delta t} u T^{t+\Delta t} \right) = \iiint_{\Omega} \left(\frac{\rho C_p}{\Delta t} u T^t - k \nabla u \nabla T^t \right) \quad (9)$$

And fully discretized it is:

$$\sum_j T_j^{t+\Delta t} \iiint_{\Omega} \left(\frac{\rho C_p}{\Delta t} \phi_i \phi_j \right) = \iiint_{\Omega} \left(\frac{\rho C_p}{\Delta t} \phi_i T^t - k \nabla \phi_i \nabla T^t \right) \quad (10)$$

Some conclusions:

- We still need to solve a linear system, even if the scheme is explicit?
- There is a stability criterion.



We have seen one approach, which is to use Euler's method to treat the transient terms. We can do the same thing with all sort of time-stepping schemes:

- Backward finite difference (BDF) from order 1 to order n
- Runge-Kutta methods (RK, IRK, DIRK, SDIRK, etc.)
- And so on and so forth...

The same idea always applies, but the devil is in the details (as always)



Transient problems

- Motivation

- Euler's methods

- Alternatives

Advection-Diffusion problem

- Motivation

- Two toy problems

- Regular formulation

- On the need for stabilization

- Artificial diffusion

- Streamline artificial diffusion

- SUPG: Regaining consistency

Conclusions

Motivation



Transport is a key transport phenomena that significantly alters the solutions we obtain. For example, the transient advection-diffusion equation:

$$\mathbf{v} \cdot \nabla c = D \nabla^2 c \quad (11)$$

where \mathbf{v} is a velocity vector. In 1D this equation becomes:

$$v_x \frac{dc}{dx} = D \frac{d^2 c}{dx^2} \quad (12)$$

Let's do some dimensional analysis on this equation.

Consequences



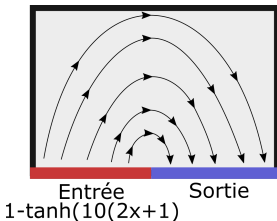
$$\underbrace{\bar{\mathbf{v}} \cdot \nabla c}_{\text{Advection}} - \frac{1}{Pe} \underbrace{\nabla^2 c}_{\text{Diffusion}} = 0$$

Two limits occur:

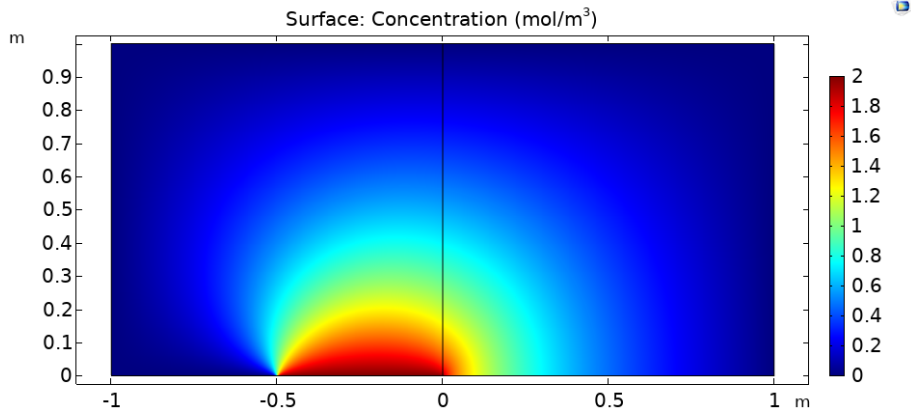
- Diffusion dominates
- Advection dominates

The behavior is controlled by the ratio between the two mechanisms:

$$Pe = \frac{vL}{D}$$



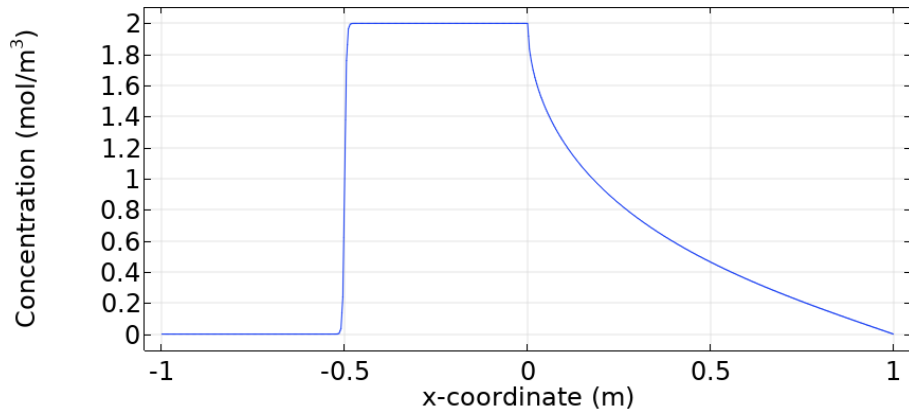
$Pe = 1$ - Strong diffusion



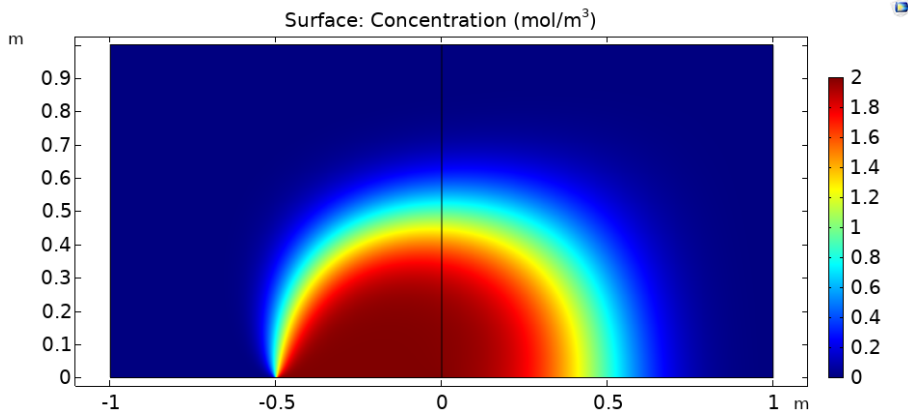
$Pe = 1$ - Strong diffusion



Line Graph: Concentration (mol/m³)

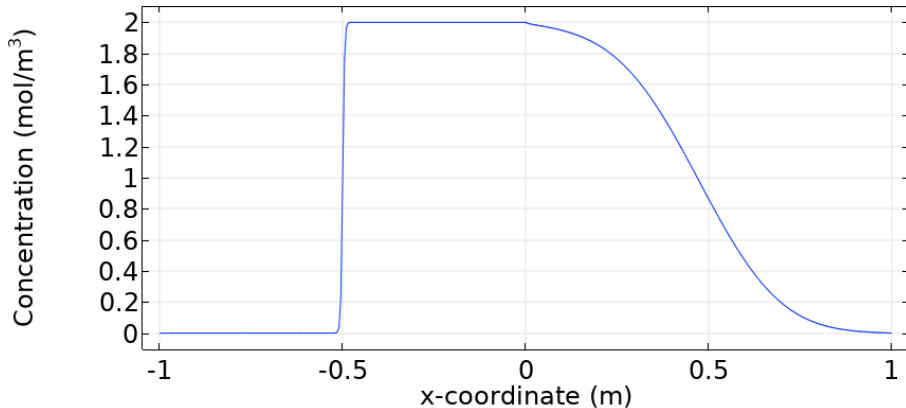


$Pe = 100$ - Balanced

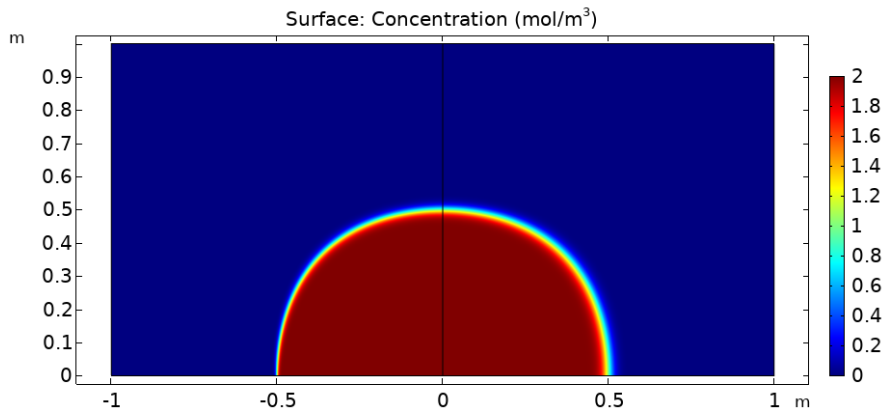




Line Graph: Concentration (mol/m³)



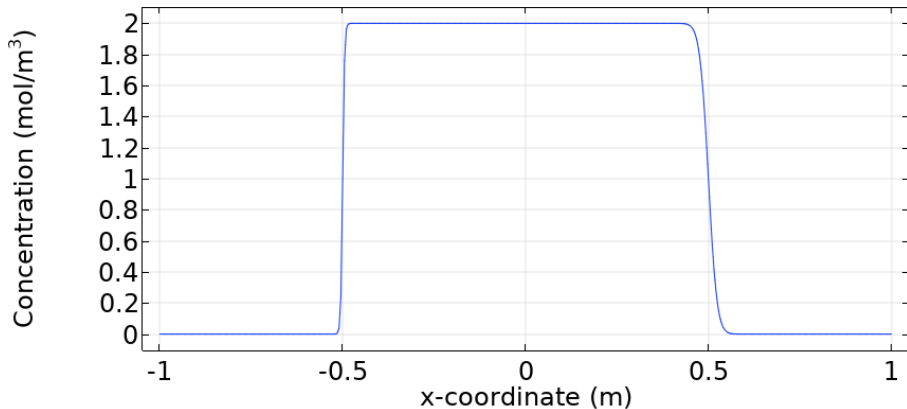
$Pe = 10^4$ - Advection dominates



$Pe = 10^4$ - **Advection dominates**



Line Graph: Concentration (mol/m³)



Toy problem 1 (TP1)



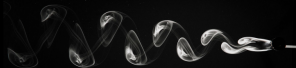
$$v \frac{\partial c}{\partial x} - D \frac{\partial^2 c}{\partial x^2} = 0$$

On domain $\Omega = [0, L]$ with the following boundary conditions:

$$c(x = 0) = 0$$

$$c(x = L) = 1$$

TP1: Dimensionless



$$\frac{\partial c}{\partial \bar{x}} - D \frac{\partial^2 c}{\partial \bar{x}^2} = 0$$

with

$$\bar{x} = \frac{x}{L}$$

$$c(\bar{x} = 0) = 0$$

$$c(\bar{x} = 1) = 1$$

$$D = \frac{1}{Pe}$$

$$Pe = \frac{vL}{D}$$

Which has the following analytical solution:

$$c = \frac{\exp(Pe \bar{x}) - 1}{\exp(Pe) - 1}$$



$$\frac{dc}{dx} - D \frac{d^2c}{dx^2} = 1 \quad (13)$$

On domain $\Omega = [0, 1]$ with $u = 1$ and $c(0) = c(1) = 0$.

Which has the following analytical solution:

$$c = x - \frac{\exp(-Pe(1-x)) - \exp(-Pe)}{1 - \exp(-Pe)} \quad (14)$$

with:

$$Pe = \frac{vL}{D}$$



We consider the following problem with Dirichlet boundary conditions:

$$\mathbf{v} \cdot \nabla c - D \nabla^2 c = 0$$

with $D = \frac{1}{Pe}$.

Let's establish the weak form



Weak form:

$$\int_{\Omega} u \mathbf{v} \cdot \nabla c \, d\Omega + \int_{\Omega} D \nabla u \cdot \nabla c \, d\Omega = 0$$

Which from now on we will note:

$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} = 0$$

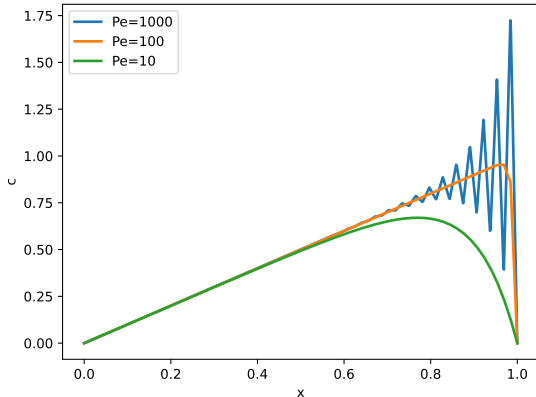
with $c, u \in H^1(\Omega)$.

TP2: Solution as function of Pe



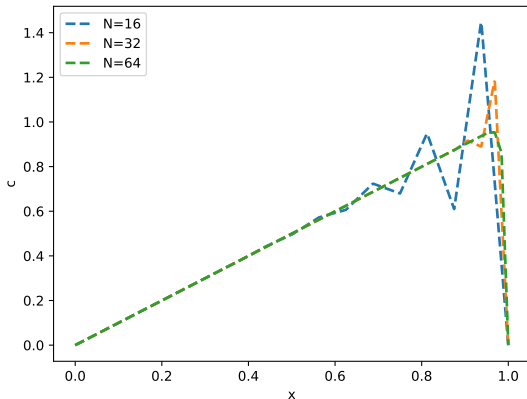
$$\frac{dc}{dx} - D \frac{d^2c}{dx^2} = 1$$

On domain $\Omega = [0, 1]$
with $c(0) = c(1) = 0$.

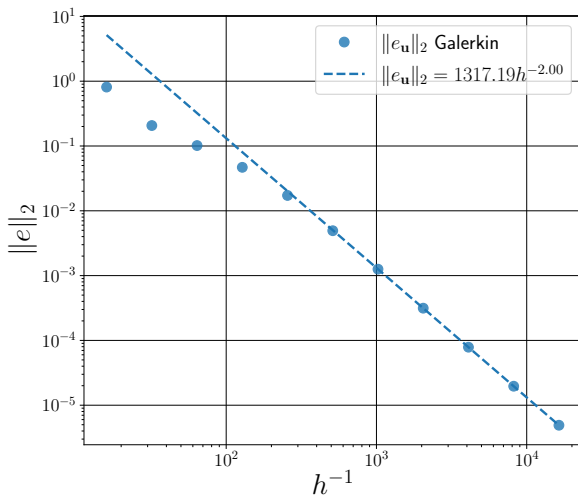


TP2: Solution as function of refinement

Refining the mesh fixes the issue. In this case, for $Pe = 100$.



Convergence?



What is happening here?



Avec les mains

As $\frac{1}{Pe}$ decreases, we lose control of the gradients of c . For small values of $\frac{1}{Pe}$ we have thin regions (layers) in which the solution changes rapidly. The Galerkin method has severe issues in handling layers. They tend to generate oscillations which will propagate throughout the domain. Once the diffusion parameter becomes smaller than the mesh size, this issue occurs.

Mathematically

$$\left\| c - c^h \right\|_{H^1} \leq (1 + C_p^{-2}) (1 + Pe) C_h h$$

with C_p and C_h constants. c is the analytical solution and c^h is the numerical solution.



The issue we have is the following:

$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} = 0$$

with D too small. The natural solution is to add more diffusion to the problem, k , to stabilize the solution.

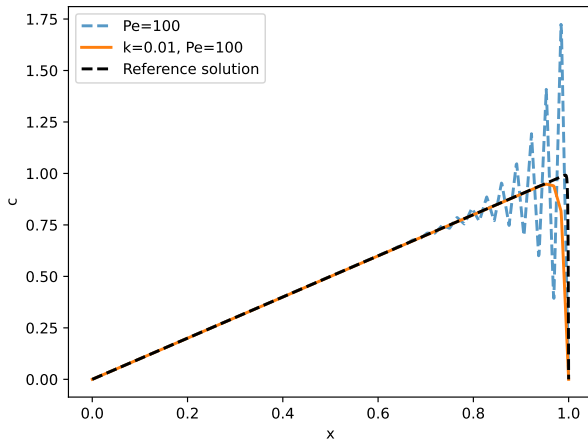
$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} + (k \nabla c, \nabla u)_{\Omega} = 0$$

This will fix our issue, because layers will be allowed to diffuse. However, how do we choose k ?

Solution obtained



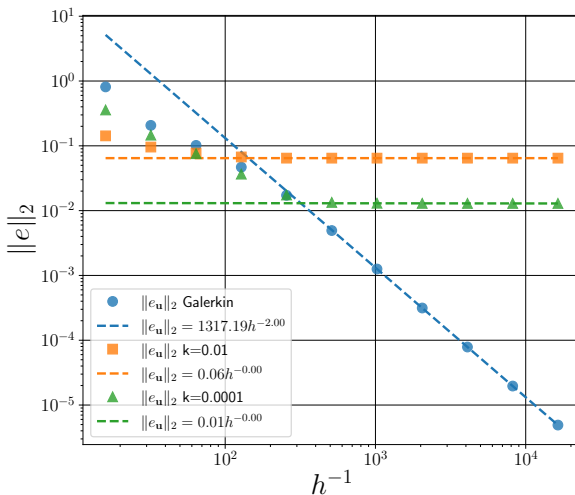
Adding k allows us to recover a smooth solution:



Convergence

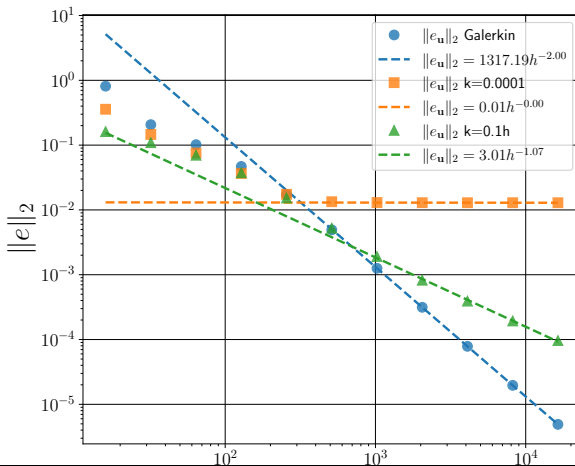


Choosing a finite value of k creates a variational inconsistency.



Convergence

A solution is to choose k such that $k = k^*h$ with h the element size. This recovers a form of variational consistency, but lowers the order of the underlying scheme.



Partial conclusion



Adding artificial diffusion has the following consequences:

- Regularizes the solution to ensure that the gradients are bounded
- Leads to a form of variational inconsistency (lowered order of convergence)
- **(not shown here)** Adds significant diffusion in the cross-wind direction (will not show up in 1D, but will in 2D and 3D).

We need a solution which adds diffusion in a more coherent way...

Streamline artificial diffusion



A solution to mitigate this problem is to add artificial diffusion in a single direction (the velocity direction).

$$\mathbf{v} \cdot \nabla c - \nabla \cdot (D\mathbf{I} \cdot \nabla c + \boldsymbol{\kappa} \cdot \nabla c) = 0$$

where $\boldsymbol{\kappa}$ is a diffusion tensor. What does it look like?

Building it



Thus we obtain

$$\mathbf{v} \cdot \nabla c - \nabla \cdot \left(D \mathbf{I} \cdot \nabla c + \frac{\kappa}{\|\mathbf{v}\|^2} \mathbf{v} \otimes \mathbf{v} \cdot \nabla c \right) = 0$$

Or in Einstein's notation:

$$v_i \partial_i c - \partial_i \left(D \partial_j c + \frac{\kappa}{\|\mathbf{v}\|^2} v_i v_j \partial_j c \right) = 0$$

we use this to define:

$$\tau = \frac{\kappa}{\|\mathbf{v}\|^2} = C \sqrt{\left(\frac{h^2}{v^2} \right)}$$

Where C is now a unitless constant. Let's now do the weak form and figure things out.

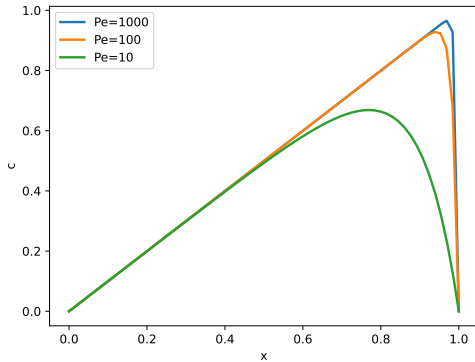


After integrating by parts we obtain:

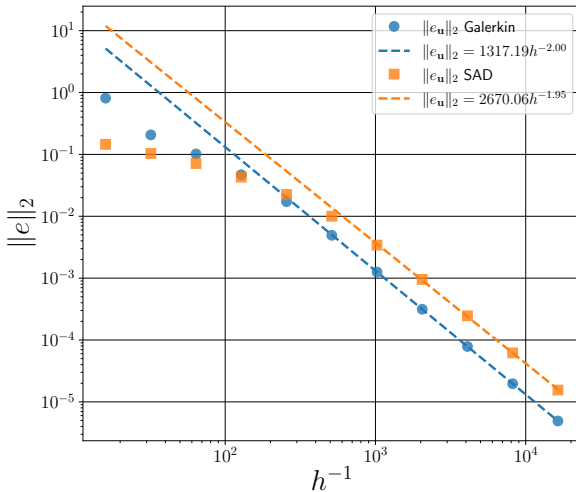
$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} + (\tau \mathbf{v} \cdot \nabla c, \mathbf{v} \cdot \nabla u)_{\Omega} = 0$$

The consequence of this is now that artificial diffusion is only applied in the direction of the velocity vector. This is a streamline artificial diffusion method.

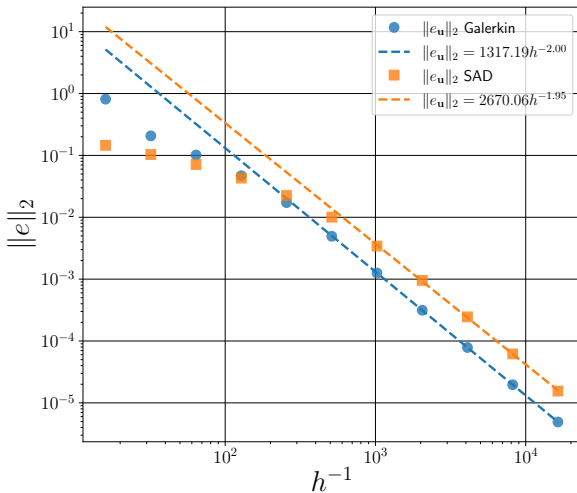
TP2: Result



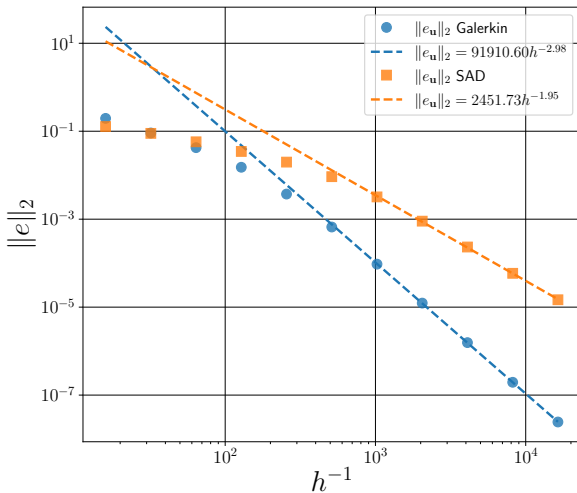
TP1: Convergence



TP2: Convergence



TP1: High-order convergence



Variational consistency



Consider TP1:

$$\mathbf{v} \cdot \nabla c - D \nabla^2 c = 0$$

For which the weak form is:

$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} = 0$$

and with streamline artificial diffusion

$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} + (\tau \mathbf{v} \cdot \nabla c, \mathbf{v} \cdot \nabla u)_{\Omega} = 0$$

The main issue is that until $(\tau \mathbf{v} \cdot \nabla c, \mathbf{v} \cdot \nabla u)_{\Omega}$ vanishes, variational consistency is not recovered. So what happened when we moved from Q1 to Q2 ?

Variational consistency



Consider a generalized form of TP2:

$$\mathbf{v} \cdot \nabla c - D \nabla^2 c = f$$

For which the weak form is:

$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} - (f, u)_{\Omega} = 0$$

and with streamline artificial diffusion

$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} - (f, u)_{\Omega} + (\tau \mathbf{v} \cdot \nabla c, \mathbf{v} \cdot \nabla u)_{\Omega} = 0$$

The same issue could also occur at Q1 if f is a non-trivial function... (I thought it would actually occur in the present case, it did not.)



The main idea behind SUPG (which is a brilliant one) is to transform streamline artificial diffusion into something that is variationally consistent. This is achieved by using the residual of the equation. Starting from our PDE:

$$\mathbf{v} \cdot \nabla c - D \nabla^2 c = 1$$

We define the strong residual $R(c)$ as:

$$R(c) = \mathbf{v} \cdot \nabla c - D \nabla^2 c - 1$$

This will be used for our upwinding.



$$(\mathbf{v} \cdot \nabla c, u)_{\Omega} + (D \nabla c, \nabla u)_{\Omega} - (1, u)_{\Omega} + (\tau(\mathbf{v} \cdot \nabla c - 1 - D \nabla^2 c), \mathbf{v} \cdot \nabla u)_{\Omega} = 0$$

This is the Streamline-Upwind / Petrov-Galerkin (SUPG) method which was first published by Brookes and Hughes in 1982. This was a revolution in the world of the finite element method.

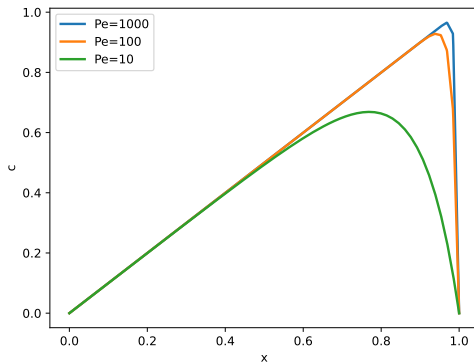
It is a very interesting approach because it is an upwinding method that preserves the order of the underlying scheme and just adds the right amount of numerical diffusion.

Results



It works... (luckily for me heh?)

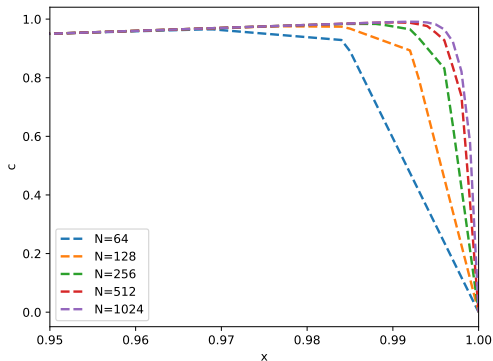
The solution appears less continuous as the Péclet number increases. It has more jagged edges. We will understand why in what follows...



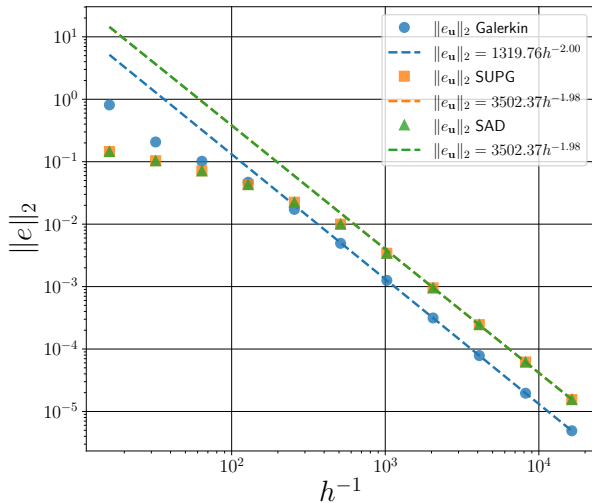
Influence of the mesh



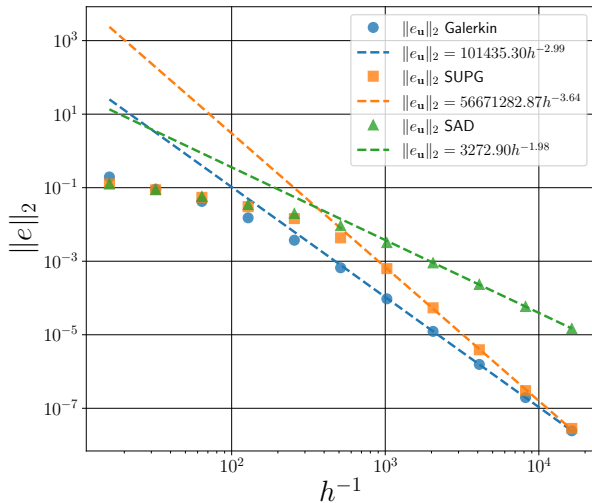
Stabilization introduces just enough numerical diffusion to ensure that the solution is adequate.



Convergence Q1



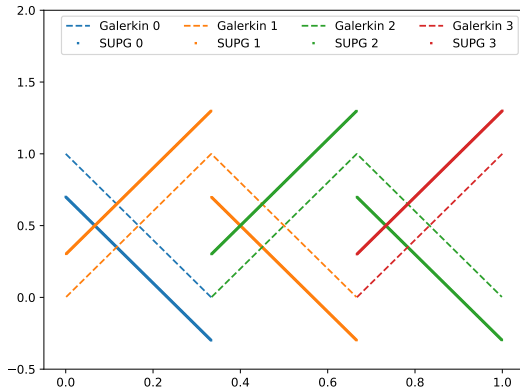
Convergence Q2



Interpretation



The Streamline Upwind Petrov-Galerkin method skews the test function of the FEM scheme to be larger *upwind* than *downwind* by using the strong form of the integral.





Transient problems

- Motivation

- Euler's methods

- Alternatives

Advection-Diffusion problem

- Motivation

- Two toy problems

- Regular formulation

- On the need for stabilization

- Artificial diffusion

- Streamline artificial diffusion

- SUPG: Regaining consistency

Conclusions

Two important topics



Transient problems

Transient problems are critical in engineering. The FEM is quite able to solve them easily. Extending a solver to solve a transient problem is relatively straightforward, depending on the time-stepping scheme you use.

Hyperbolic problems

Hyperbolic problems are very subtle. Solving them with FEM is quite challenging and the literature on stabilized method is very difficult to understand. FEM works correctly for hyperbolic problems, but it is less straightforward to implement than it is for the Finite Volume Method for example. Hence the latter being more common in CFD. The literature on this is huge! There are many alternatives such as **Discontinuous Galerkin** methods.