Fundamentals of Numerical Thermo-Fluid Dynamics 322.061

Exercise 7: Burger's Equation

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In this exercise we propose to use different techniques to solve the following non linear problem:

$$\begin{cases} \partial_t u + u \,\partial_x u - \nu \partial_{xx} u = 0 \quad , \text{ with } t \in \mathbb{R}^+, x \in [0, 1], \nu \in \mathbb{R}^+ \\ u(t = 0, x) = 1 - x + e^{-\frac{(x - 0.5)^2}{0.02}} \\ u(t, x = 0) = 1 \quad \frac{\partial}{\partial x} u(t, x = 1) = 0 \end{cases}$$
(1)

7.1) Solve numerically the problem by treating explicitly the non linear term, and implicitly the linear terms. (Functions from previous exercises can be used ...)

$$\frac{U_i^{n+1} - U_i^n}{\Delta t} - \nu \frac{U_{i-1}^{n+1} - 2U_i^{n+1} + U_{i+1}^{n+1}}{\Delta x^2} = -U_i^n \frac{U_{i+1}^n - U_{i-1}^n}{2\Delta x}$$
(2)

7.2) We propose here to solve this problem by solving a non-linear problem through a fixed point algorithm.

$$\frac{V_i^{k+1} - U_i^n}{\Delta t} - \nu \frac{V_{i-1}^k - 2V_i^k + V_{i+1}^k}{\Delta x^2} = -V_i^k \frac{V_{i+1}^k - V_{i-1}^k}{2\Delta x}$$
(3)

- Put this problem in a fixed point function *i.e.* $V^{k+1} = f(V^k)$.
- Find the solution, implementing Algorithm 1 to solve one time step.
- See what happens for different relaxation factor θ (usually $\theta \approx 0.01$).

Algorithm 1 My fixed point

1: procedure GetStepFixedPoint(U^n , tol, k_{max} , θ) $V_1 \leftarrow U^n; k \leftarrow 0$ 2: while $tol < \delta$ and $k < k_{max}$ do 3: $V^{k+1} \leftarrow f(V^k, U^n)$ 4: $\delta \leftarrow norm(V^{k+1} - V^k)/norm(V^k)$ 5: $V^{k+1} \leftarrow \theta V^{k+1} + (1 - \theta) V^k$ 6: $k \leftarrow k+1$ 7: end do 8: return V^{k+1} 9:

7.3) We propose in the following part to solve the Burger equation using the Newton method. The non-linear function that one consider is then:

$$F_i(V, U^n) = \frac{V_i - U_i^n}{\Delta t} - \nu \frac{V_{i-1} - 2V_i^+ V_{i+1}}{\Delta x^2} + V_i \frac{V_{i+1} - V_{i-1}}{2\Delta x}.$$
 (4)

• Differentiate (4) with respect to V_{i-1}, V_i and V_{i+1} . Find that the jacobian J is a tridiagonal matrix. We recall that

$$J(V, U^n) = \begin{pmatrix} \partial_{V_1} F_1 & \partial_{V_2} F_1 & \cdots & \partial_{V_i} F_1 & \cdots & \partial_{V_N} F_1 \\ \vdots & & \vdots & & \vdots \\ \partial_{V_1} F_i & \partial_{V_2} F_i & \cdots & \partial_{V_i} F_i & \cdots & \partial_{V_N} F_i \\ \vdots & & & \vdots & & \vdots \\ \partial_{V_1} F_N & \partial_{V_2} F_N & \cdots & \partial_{V_i} F_N & \cdots & \partial_{V_N} F_N \end{pmatrix}.$$

• Solve numerically the problem, by implement a Newton method solving U for a time step (*cf.* Algorithm 2).

Algorithm 2 My Newton

```
1: procedure GETSTEPNEWTON(U^n, tol, k_{max})
 2:
         V_1 \leftarrow U^n
         k \leftarrow 1
 3:
         while tol > \delta and k < k_{max} do
 4:
             F \leftarrow createVector_F(V^k, U^n)
 5:
             J \leftarrow createJacobian(V^k, U^n)
 6:
             V^{k+1} = V^k - J^{-1}F
 7:
             \delta \leftarrow norm(V^{k+1} - V^k)/norm(V^k)
 8:
             k \leftarrow k+1
9:
         end do
10:
         U^{n+1} \leftarrow V^{k+1}
11:
         return U^{n+1}
12:
```

7.4) What can you conclude concerning the convergence of each method, and how fast each method converges to a solution ? Can you think of a case in which the Newton method cannot be applied ?