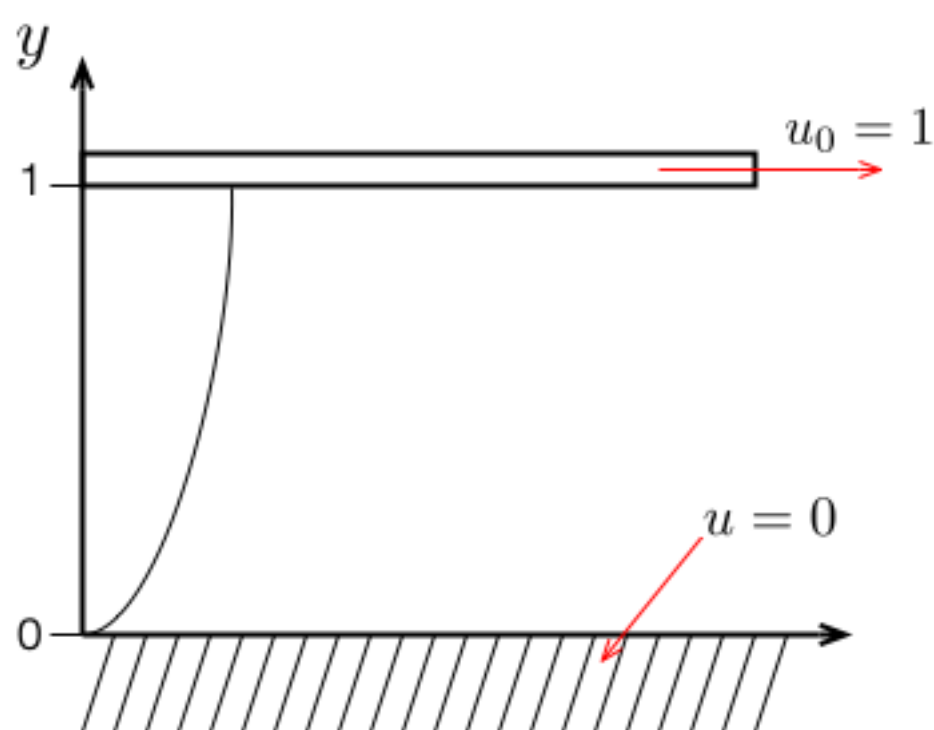


# PDE python exercise 5

In this exercise we are looking for the solution of a confined, unsteady couette flow. This problem is controlled by a parabolic PDE and is illustrated below. We consider the unsteady Couette flow, confined by two walls, which has the following governing equation:

$$\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial y^2}$$

The system we are looking at is described in the following picture, we are looking a fluid constrained between two rigid walls, one fixed and one moving with a constant velocity  $u_0$ . We are interested by the distribution of the velocity along the  $y$  axis within the time frame  $[0,0.4]$ .



where the velocity  $u$ :

- $u = 0$  when  $y = 0$
- $u = u_0$  when  $y = 1$
- driven by the PDE when  $0 < y < 1$

As initial conditions we have  $u = 0$  for  $0 < y < 1$

In this exercise, you have to:

- Discretize the PDE using the FTCS scheme (i.e forward in time and centered in space)
- Implement a function which solve the PDE using the FTCS scheme. As input we want the initial and boundary conditions as well as the number of points along the  $y$  axis and the number of time step. The diffusion number  $D$  will be computed inside this function. This function should return the complete history of the velocity  $u$  (i.e. a matrix). In this function you can use `for` loops to fill out the columns of  $u$ .
- Run the problem with 10 grid points along  $y$  and 1001 timesteps  $\Delta t$  and plot the distribution of the velocity along  $y$  for the timesteps 100, 500 and 1000
- Run the problem with 10 grid points along  $y$  and 11 timesteps  $\Delta t$  and plot the distribution of the velocity along  $y$  for the timesteps 1, 5 and 10. What can you observe?
- Define a new function where you solve the PDE problem using array slicing instead of for loops and repeat question 3.

## Question 1

For the FTCS scheme, we make use of centered finite difference for the space term:

$$\frac{\partial^2 u}{\partial y^2} \Big|_j^n \approx \frac{u_{j+1}^n - 2u_j^n + u_{j-1}^n}{(\Delta y)^2}$$

and a forward difference for the time term:

$$\frac{\partial u}{\partial t} \Big|_j^n \approx \frac{u_j^{n+1} - u_j^n}{\Delta t}$$

where the index  $j$  indicates the space coordinate and the index  $n$  represents the time.

Once combined we expressed the FTCS scheme as:

$$u_j^{n+1} = D(u_{j+1}^n + u_{j-1}^n) + (1 - 2D)u_j^n$$

where  $D$  is the diffusion number and is defined as:

$$D = \frac{\Delta t}{(\Delta y)^2}$$

```
In [ ]: import numpy as np
import matplotlib.pyplot as plt
```

In the next cell we define a function which computes the complete history of the velocity  $u$  with as input the discretization along the  $y$  axis and the time  $t$ . We have defined the initial and boundary conditions into the function.

```
In [ ]: def compute_u(Ny,Nt):
    y = np.linspace(0, 1, Ny)
    time = np.linspace(0,0.4,Nt)
    dy = y[1] - y[0]
    dt = time[1]-time[0]
    # Compute the diffusion number
    D = dt/dy**2.0
    U = np.zeros([Nt,Ny])
    # Enforce boundary and initial conditions
    U[:,0] = 0
    U[:, -1] = 1
    # Loop over time
    for n in range(1,Nt):
        # Loop over space
        for j in range(1,Ny-1):
            U[n,j] = D*(U[n-1,j+1]+U[n-1,j-1])+(1-2*D)*U[n-1,j]
    return U
```

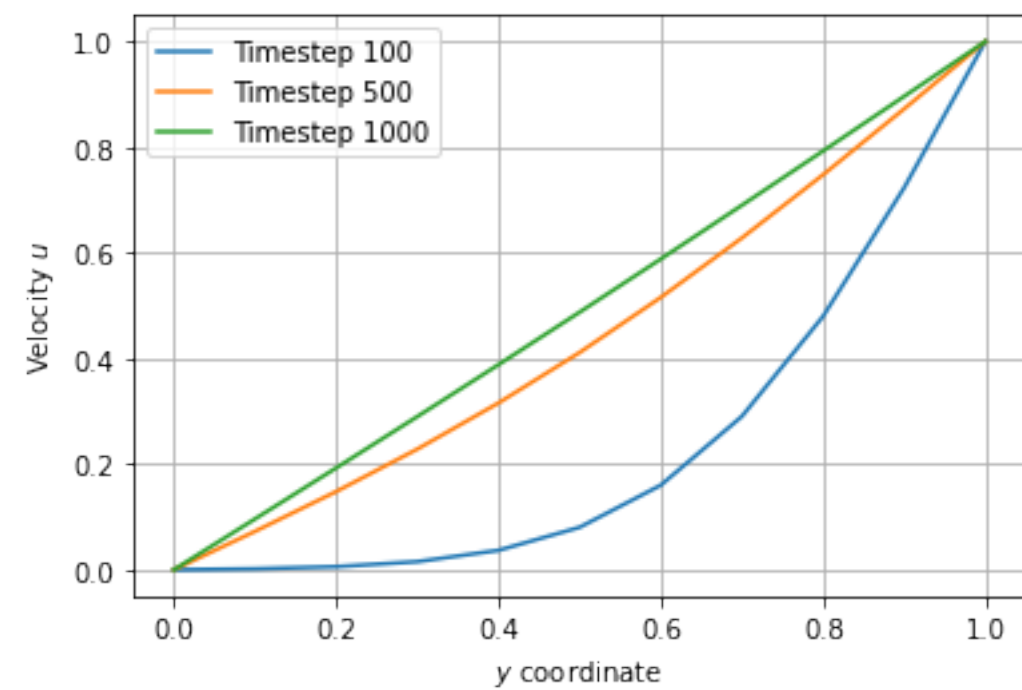
We first run the simulations with 1001 timesteps for 11 points along the  $y$  axis and plot the results.

```
In [ ]: U = compute_u(11,1001)
```

```
In [ ]: y = np.linspace(0,1,11)

plt.plot(y,U[100,:],label='Timestep 100')
plt.plot(y,U[500,:],label='Timestep 500')
plt.plot(y,U[1000,:],label='Timestep 1000')
plt.grid()
plt.legend()
plt.xlabel(r'$y$ coordinate')
plt.ylabel(r'VeLOCITY $u$')
```

```
Out [ ]: Text(0, 0.5, 'Velocity $u$')
```

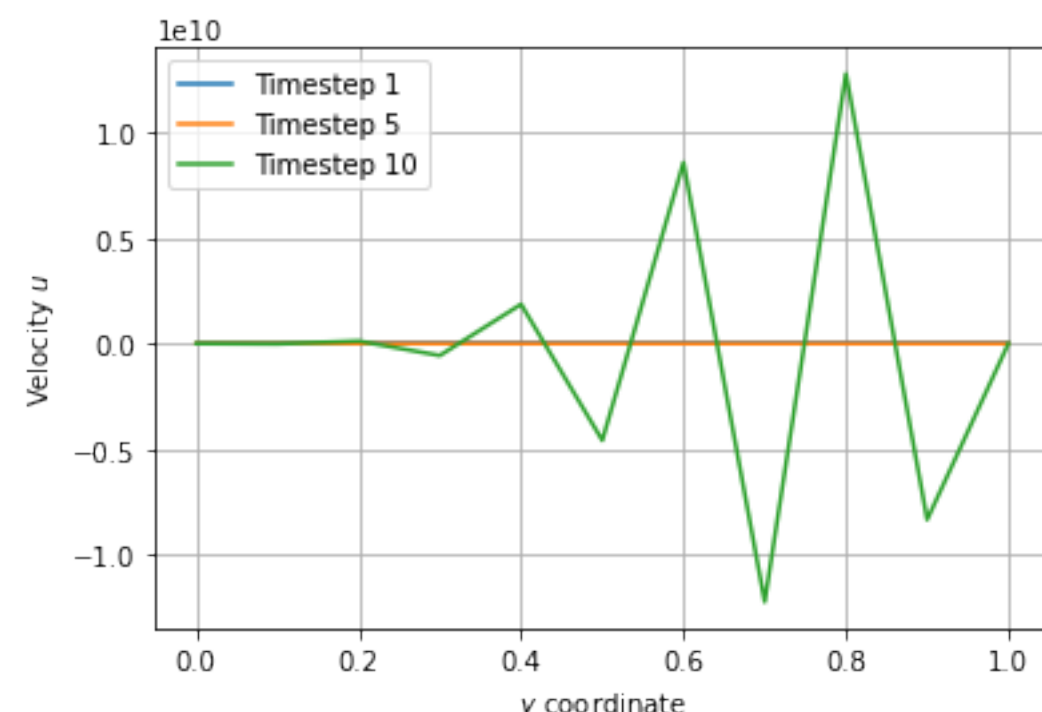


We then re-run the same simulation using 11 timesteps and 11 points along the  $y$  coordinate. The results are very different from the previous run showing an unstable behaviour from the numerical scheme.

```
In [ ]: U = compute_u(11,11)
y = np.linspace(0,1,11)

plt.plot(y,U[1,:],label='Timestep 1')
plt.plot(y,U[5,:],label='Timestep 5')
plt.plot(y,U[10,:],label='Timestep 10')
plt.grid()
plt.legend()
plt.xlabel(r'$y$ coordinate')
plt.ylabel(r'VeLOCITY $u$')
```

```
Out [ ]: Text(0, 0.5, 'Velocity $u$')
```



For the last question we re-code our function using array slicing instead of a `for` loop to compute the space part of our numerical scheme.

```
In [ ]: def compute_u_array(Ny,Nt):
    y = np.linspace(0, 1, Ny)
    time = np.linspace(0,0.4,Nt)
    dy = y[1] - y[0]
    dt = time[1]-time[0]
    # Compute the diffusion number
    D = dt/dy**2.0
    U = np.zeros([Nt,Ny])
    # Enforce boundary and initial conditions
    U[:,0] = 0
    U[:, -1] = 1
    # Loop over time
    for n in range(1,Nt):
        U[n,1:-1] = D*(U[n-1,2:]+U[n-1,-2:])+(1-2*D)*U[n-1,1:-1]
    return U
```

```
In [ ]: y = np.linspace(0,1,11)
U = compute_u(11,1001)
plt.plot(y,U[100,:],label='Timestep 100')
plt.plot(y,U[500,:],label='Timestep 500')
plt.plot(y,U[1000,:],label='Timestep 1000')
plt.grid()
plt.legend()
plt.xlabel(r'$y$ coordinate')
plt.ylabel(r'VeLOCITY $u$')
```

```
Out [ ]: Text(0, 0.5, 'Velocity $u$')
```

