

Exercise 5

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1 Implicit Euler scheme for diffusion equation

1.a Discretization

Recalling the general form of the implicit Euler method:

$$\phi_{i,j}^{n+1} = \phi_{i,j}^n + f(\phi^{n+1}, t^{n+1})\Delta t$$

we get for the diffusion equation:

$$\phi_{i,j}^{n+1} = \phi_{i,j}^n + \alpha \left(\frac{\partial^2 \phi_{i,j}^{n+1}}{\partial x^2} + \frac{\partial^2 \phi_{i,j}^{n+1}}{\partial y^2} \right) \Delta t$$

Then we use the second order central difference scheme for the spatial derivatives:

$$\begin{aligned} \frac{\partial^2 \phi_{i,j}^{n+1}}{\partial x^2} &= \frac{\phi_{i+1,j}^{n+1} - 2\phi_{i,j}^{n+1} + \phi_{i-1,j}^{n+1}}{\Delta x^2} \\ \frac{\partial^2 \phi_{i,j}^{n+1}}{\partial y^2} &= \frac{\phi_{i,j+1}^{n+1} - 2\phi_{i,j}^{n+1} + \phi_{i,j-1}^{n+1}}{\Delta y^2} \end{aligned}$$

Substituting the above equations into the diffusion equation, we get:

$$\phi_{i,j}^{n+1} = \phi_{i,j}^n + \alpha \left(\frac{\phi_{i+1,j}^{n+1} - 2\phi_{i,j}^{n+1} + \phi_{i-1,j}^{n+1}}{\Delta x^2} + \frac{\phi_{i,j+1}^{n+1} - 2\phi_{i,j}^{n+1} + \phi_{i,j-1}^{n+1}}{\Delta y^2} \right) \Delta t \quad (1)$$

where we can rearrange the terms to get the following expression:

$$\phi_{i,j}^{n+1} (1 + 2\alpha\Delta t (\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2})) - \alpha\Delta t \left(\frac{\phi_{i+1,j}^{n+1} + \phi_{i-1,j}^{n+1}}{\Delta x^2} + \frac{\phi_{i,j+1}^{n+1} + \phi_{i,j-1}^{n+1}}{\Delta y^2} \right) = \phi_{i,j}^n \quad (2)$$

1.b Consistency proof

In order to prove consistency of the discretization we have to show that the truncation error T goes to zero as the grid spacing $(\Delta x, \Delta y)$ goes to zero and the time step (Δt) goes to zero.

We start with the exact form of the diffusion equation:

$$\frac{\partial \phi}{\partial t} = \alpha \left(\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} \right). \quad (3)$$

The discretized scheme for the implicit Euler method is described by equation (??).

To analyze the truncation error, we expand $\phi_{i+1,j}^{n+1}, \phi_{i-1,j}^{n+1}, \phi_{i,j+1}^{n+1}, \phi_{i,j-1}^{n+1}$ using Taylor series around $\phi_{i,j}^{n+1}$. For example, the expansion for $\phi_{i+1,j}^{n+1}$ is given by:

$$\phi_{i+1,j}^{n+1} = \phi_{i,j}^{n+1} + \Delta x \frac{\partial \phi}{\partial x} + \frac{\Delta x^2}{2} \frac{\partial^2 \phi}{\partial x^2} + \frac{\Delta x^3}{6} \frac{\partial^3 \phi}{\partial x^3} + \dots,$$

and similar expansions hold for the other terms. Then we substitute these expansions into the discretized scheme. For example, the term

$$\frac{\phi_{i+1,j}^{n+1} - 2\phi_{i,j}^{n+1} + \phi_{i-1,j}^{n+1}}{\Delta x^2}$$

becomes:

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\Delta x^2}{12} \frac{\partial^4 \phi}{\partial x^4} + \dots$$

Substituting all the Taylor expansions into the discretized equation and rearranging terms, we obtain the residual or truncation error as:

$$T = \Delta t \frac{\partial^2 \phi}{\partial t^2} + \frac{\Delta x^2}{12} \frac{\partial^4 \phi}{\partial x^4} + \frac{\Delta y^2}{12} \frac{\partial^4 \phi}{\partial y^4} + \dots \quad (4)$$

As $\Delta t \rightarrow 0$, $\Delta x \rightarrow 0$, and $\Delta y \rightarrow 0$, the truncation error T approaches zero, demonstrating that the discretization is consistent.

1.c Stability criteria

To determine the stability criterion for the implicit Euler scheme using the Von Neumann method, we start with the discretized equation (??) and we introduce the parameters $r_x = \frac{\alpha \Delta t}{\Delta x^2}$ and $r_y = \frac{\alpha \Delta t}{\Delta y^2}$, then the equation becomes:

$$\phi_{i,j}^{n+1} = \phi_{i,j}^n + r_x (\phi_{i+1,j}^{n+1} - 2\phi_{i,j}^{n+1} + \phi_{i-1,j}^{n+1}) + r_y (\phi_{i,j+1}^{n+1} - 2\phi_{i,j}^{n+1} + \phi_{i,j-1}^{n+1}).$$

Rearranging terms, we write:

$$\phi_{i,j}^{n+1} (1 + 2r_x + 2r_y) = \phi_{i,j}^n + r_x (\phi_{i+1,j}^{n+1} + \phi_{i-1,j}^{n+1}) + r_y (\phi_{i,j+1}^{n+1} + \phi_{i,j-1}^{n+1}).$$

The Von Neumann method assumes a Fourier mode solution of the form:

$$\phi_{i,j}^n = G^n e^{I(k_x i \Delta x + k_y j \Delta y)},$$

where G is the amplification factor, k_x and k_y are the wave numbers in the x - and y -directions, and $I = \sqrt{-1}$.

We substituting this Fourier mode into the discretized equation, and the neighboring terms become:

$$\begin{aligned} \phi_{i+1,j}^{n+1} &= G e^{I k_x \Delta x} \phi_{i,j}^{n+1}, & \phi_{i-1,j}^{n+1} &= G e^{-I k_x \Delta x} \phi_{i,j}^{n+1}, \\ \phi_{i,j+1}^{n+1} &= G e^{I k_y \Delta y} \phi_{i,j}^{n+1}, & \phi_{i,j-1}^{n+1} &= G e^{-I k_y \Delta y} \phi_{i,j}^{n+1}. \end{aligned}$$

Using these expressions, substitute into the equation. Simplify the exponentials using $e^{I\theta} + e^{-I\theta} = 2 \cos \theta$, leading to:

$$G \phi_{i,j}^{n+1} (1 + 2r_x + 2r_y) = \phi_{i,j}^n + 2r_x G \phi_{i,j}^{n+1} \cos(k_x \Delta x) + 2r_y G \phi_{i,j}^{n+1} \cos(k_y \Delta y).$$

Factoring G from the left-hand side:

$$G (1 + 2r_x + 2r_y - 2r_x \cos(k_x \Delta x) - 2r_y \cos(k_y \Delta y)) = 1.$$

We solve for G :

$$G = \frac{1}{1 + 2r_x + 2r_y - 2r_x \cos(k_x \Delta x) - 2r_y \cos(k_y \Delta y)}.$$

The scheme is stable if $|G| \leq 1$. For the implicit Euler method, the denominator in G is always greater than 1 for all k_x and k_y . So we have that:

$$|G| \leq 1, \tag{5}$$

which implies unconditional stability.

1.d Convergence proof

Since the scheme is both consistent and stable, for the Lax equivalence theorem, the scheme is convergent.

2 Explicit Euler Scheme for the Diffusion Equation

2.a Discretization

The two-dimensional diffusion equation is given as:

$$\frac{\partial \Phi}{\partial t} = \alpha \left(\frac{\partial^2 \Phi}{\partial x^2} + \frac{\partial^2 \Phi}{\partial y^2} \right). \tag{6}$$

Using the explicit Euler method, we approximate the time derivative as:

$$\frac{\partial \Phi}{\partial t} \approx \frac{\Phi_{i,j}^{n+1} - \Phi_{i,j}^n}{\Delta t}. \tag{7}$$

Here, $\Phi_{i,j}^n$ represents the value of Φ at grid point (i, j) at time step n , and Δt is the time step size.

Rewriting the time derivative, we get:

$$\Phi_{i,j}^{n+1} = \Phi_{i,j}^n + \Delta t \cdot \alpha \left(\frac{\partial^2 \Phi}{\partial x^2} + \frac{\partial^2 \Phi}{\partial y^2} \right). \tag{8}$$

For the second-order derivatives in space, we use central finite differencing. The second derivative in the x -direction is approximated as:

$$\frac{\partial^2 \Phi}{\partial x^2} \approx \frac{\Phi_{i+1,j}^n - 2\Phi_{i,j}^n + \Phi_{i-1,j}^n}{\Delta x^2}. \tag{9}$$

Similarly, the second derivative in the y -direction is approximated as:

$$\frac{\partial^2 \Phi}{\partial y^2} \approx \frac{\Phi_{i,j+1}^n - 2\Phi_{i,j}^n + \Phi_{i,j-1}^n}{\Delta y^2}. \quad (10)$$

Substitute the spatial discretizations into the equation for $\Phi_{i,j}^{n+1}$:

$$\Phi_{i,j}^{n+1} = \Phi_{i,j}^n + \alpha \Delta t \left(\frac{\Phi_{i+1,j}^n - 2\Phi_{i,j}^n + \Phi_{i-1,j}^n}{\Delta x^2} + \frac{\Phi_{i,j+1}^n - 2\Phi_{i,j}^n + \Phi_{i,j-1}^n}{\Delta y^2} \right). \quad (11)$$

2.b Consistency

The goal is to prove that the explicit Euler discretization for the two-dimensional diffusion equation is consistent. This involves showing that the truncation error T vanishes as $\Delta t \rightarrow 0$, $\Delta x \rightarrow 0$, and $\Delta y \rightarrow 0$.

Rewriting the discretized form, we have:

$$\frac{\Phi_{i,j}^{n+1} - \Phi_{i,j}^n}{\Delta t} = \alpha \left(\frac{\Phi_{i+1,j}^n - 2\Phi_{i,j}^n + \Phi_{i-1,j}^n}{\Delta x^2} + \frac{\Phi_{i,j+1}^n - 2\Phi_{i,j}^n + \Phi_{i,j-1}^n}{\Delta y^2} \right). \quad (12)$$

Expand the neighboring points $\Phi_{i+1,j}^n$, $\Phi_{i-1,j}^n$, $\Phi_{i,j+1}^n$, $\Phi_{i,j-1}^n$ using Taylor series about $\Phi_{i,j}^n$:

$$\Phi_{i+1,j}^n = \Phi_{i,j}^n + \Delta x \frac{\partial \Phi}{\partial x} + \frac{\Delta x^2}{2} \frac{\partial^2 \Phi}{\partial x^2} + \frac{\Delta x^3}{6} \frac{\partial^3 \Phi}{\partial x^3} + \dots, \quad (13)$$

$$\Phi_{i-1,j}^n = \Phi_{i,j}^n - \Delta x \frac{\partial \Phi}{\partial x} + \frac{\Delta x^2}{2} \frac{\partial^2 \Phi}{\partial x^2} - \frac{\Delta x^3}{6} \frac{\partial^3 \Phi}{\partial x^3} + \dots. \quad (14)$$

Similarly, for the y -direction:

$$\Phi_{i,j+1}^n = \Phi_{i,j}^n + \Delta y \frac{\partial \Phi}{\partial y} + \frac{\Delta y^2}{2} \frac{\partial^2 \Phi}{\partial y^2} + \frac{\Delta y^3}{6} \frac{\partial^3 \Phi}{\partial y^3} + \dots, \quad (15)$$

$$\Phi_{i,j-1}^n = \Phi_{i,j}^n - \Delta y \frac{\partial \Phi}{\partial y} + \frac{\Delta y^2}{2} \frac{\partial^2 \Phi}{\partial y^2} - \frac{\Delta y^3}{6} \frac{\partial^3 \Phi}{\partial y^3} + \dots. \quad (16)$$

Substitute the Taylor expansions into the spatial discretization terms:

$$\frac{\Phi_{i+1,j}^n - 2\Phi_{i,j}^n + \Phi_{i-1,j}^n}{\Delta x^2} = \frac{\partial^2 \Phi}{\partial x^2} + \frac{\Delta x^2}{12} \frac{\partial^4 \Phi}{\partial x^4} + \mathcal{O}(\Delta x^4), \quad (17)$$

$$\frac{\Phi_{i,j+1}^n - 2\Phi_{i,j}^n + \Phi_{i,j-1}^n}{\Delta y^2} = \frac{\partial^2 \Phi}{\partial y^2} + \frac{\Delta y^2}{12} \frac{\partial^4 \Phi}{\partial y^4} + \mathcal{O}(\Delta y^4). \quad (18)$$

The residual (truncation error T) when substituting the discretized equation into the original PDE is:

$$T = \frac{\Phi_{i,j}^{n+1} - \Phi_{i,j}^n}{\Delta t} - \alpha \left(\frac{\partial^2 \Phi}{\partial x^2} + \frac{\partial^2 \Phi}{\partial y^2} \right). \quad (19)$$

From the Taylor expansions, the leading terms of T are:

$$T = \Delta t \frac{\partial^2 \Phi}{\partial t^2} + \frac{\Delta x^2}{12} \frac{\partial^4 \Phi}{\partial x^4} + \frac{\Delta y^2}{12} \frac{\partial^4 \Phi}{\partial y^4} + \mathcal{O}(\Delta t^2, \Delta x^4, \Delta y^4). \quad (20)$$

As $\Delta t \rightarrow 0$, $\Delta x \rightarrow 0$, and $\Delta y \rightarrow 0$, the truncation error $T \rightarrow 0$. Thus, the scheme is consistent.

2.c Stability Analysis using the Von Neumann Method

We assume a Fourier mode solution of the form:

$$\Phi_{i,j}^n = G^n e^{I(k_x i \Delta x + k_y j \Delta y)}, \quad (21)$$

We substitute the assumed solution into the discretized equation. For neighboring terms in the x -direction:

$$\Phi_{i+1,j}^n = G^n e^{I k_x (i+1) \Delta x} = G^n e^{I k_x \Delta x} e^{I k_x i \Delta x}, \quad (22)$$

$$\Phi_{i-1,j}^n = G^n e^{I k_x (i-1) \Delta x} = G^n e^{-I k_x \Delta x} e^{I k_x i \Delta x}. \quad (23)$$

For the y -direction:

$$\Phi_{i,j+1}^n = G^n e^{Ik_y(j+1)\Delta y} = G^n e^{Ik_y\Delta y} e^{Ik_y j\Delta y}, \quad (24)$$

$$\Phi_{i,j-1}^n = G^n e^{Ik_y(j-1)\Delta y} = G^n e^{-Ik_y\Delta y} e^{Ik_y j\Delta y}. \quad (25)$$

Substituting these into the discretized equation, we have:

$$G\Phi_{i,j}^n = \Phi_{i,j}^n + \alpha\Delta t \left[\frac{\Phi_{i,j}^n (e^{Ik_x\Delta x} + e^{-Ik_x\Delta x} - 2)}{\Delta x^2} + \frac{\Phi_{i,j}^n (e^{Ik_y\Delta y} + e^{-Ik_y\Delta y} - 2)}{\Delta y^2} \right]. \quad (26)$$

Using the identity $e^{I\theta} + e^{-I\theta} = 2\cos\theta$, we simplify:

$$G\Phi_{i,j}^n = \Phi_{i,j}^n + \alpha\Delta t \Phi_{i,j}^n \left[\frac{2\cos(k_x\Delta x) - 2}{\Delta x^2} + \frac{2\cos(k_y\Delta y) - 2}{\Delta y^2} \right]. \quad (27)$$

Divide through by $\Phi_{i,j}^n$ (which is non-zero):

$$G = 1 + 2\alpha\Delta t \left[\frac{\cos(k_x\Delta x) - 1}{\Delta x^2} + \frac{\cos(k_y\Delta y) - 1}{\Delta y^2} \right]. \quad (28)$$

The method is stable if $|G| \leq 1$. Since $\cos\theta \leq 1$, the terms $\cos(k_x\Delta x) - 1$ and $\cos(k_y\Delta y) - 1$ are always ≤ 0 . To ensure stability, the magnitude of the amplification factor G must not exceed 1. This leads to the stability criterion:

$$\alpha\Delta t \left(\frac{1}{\Delta x^2} + \frac{1}{\Delta y^2} \right) \leq \frac{1}{2}. \quad (29)$$

2.d Convergence

Since the scheme is both consistent and stable, for the Lax equivalence theorem, the scheme is convergent.

3 Finite-volume method

3.a Co-located grid

In the co-located finite volume method, all variables are stored at the same location, typically the cell center. This approach uses the same control volumes for all conservation equations, which minimizes programming and storage effort, especially for geometric quantities. It is particularly beneficial in handling complex geometries, as control volumes can easily conform to complicated boundaries. However, a significant challenge is the pressure-velocity coupling, which can lead to oscillations in the pressure field. This issue is often addressed using "momentum interpolation" to ensure stability.

3.b Staggered grid

A staggered grid is a numerical technique in which different variables are stored at different locations within the control volume. Specifically, pressure is stored at the center of the cell, while velocity components are stored at the faces of the cell. This results in different control volumes for different equations. One of the primary advantages of the staggered grid is that many terms can be calculated without the need for interpolation, maintaining second-order accuracy. It provides strong coupling between the pressure and velocity fields, which helps prevent oscillations in the solution, ensuring stability. However, a significant disadvantage of the staggered grid is the difficulty and computational expense of extending it to arbitrary, curvilinear, and non-orthogonal grids, which limits its flexibility in handling complex geometries.

3.c Variable arrangements

In the picture reported below, the staggered variable arrangement is illustrated as (1). In contrast, (2) shows the co-located variable arrangement.

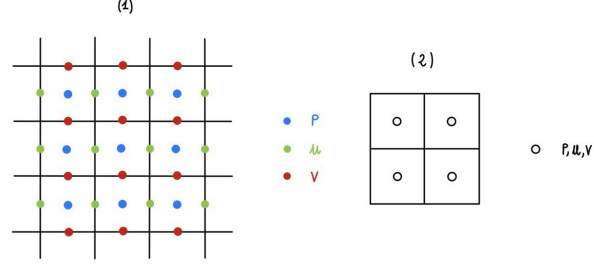


Figure 1: Variable arrangements

3.d Integral form of the equation

To derive the integral form of the partial differential equation, we integrate over a control volume:

$$\int_V \frac{\partial(\rho\Phi)}{\partial t} dV + \int_V \nabla \cdot (\rho\vec{u}\Phi - \alpha\nabla\Phi) dV = \int_V S_\Phi dV$$

Applying the divergence theorem to the second term, we obtain:

$$\int_V \frac{\partial(\rho\Phi)}{\partial t} dV + \int_{\partial V} (\rho\vec{u}\Phi - \alpha\nabla\Phi) \cdot \vec{n} dA = \int_V S_\Phi dV$$

3.e Approximation of the volume and surface integral

Using the mid-point rule, volume integrals over are approximated by evaluating the integrand at the center of the control volume P .

$$\int_V \frac{\partial(\rho\Phi)}{\partial t} dV \approx \frac{\partial(\rho\Phi)_P}{\partial t} \Delta x \Delta y$$

Surface integrals over are approximated by evaluating fluxes through face at the midpoints of the faces of the control volume.

$$\int_{\partial V} (\rho u \Phi - \alpha \nabla \Phi) \cdot \vec{n} dA \approx \sum_f F_f A_f$$

3.f Approximation of the source term and time derivative

The source term S_Φ is assumed constant within the control volume:

$$\int_V S_\Phi dV \approx S_{\Phi,P} \Delta x \Delta y$$

The time derivative is discretized using a finite difference method. For a first-order explicit scheme:

$$\frac{\partial(\rho\Phi)_P}{\partial t} \approx \frac{(\rho\Phi)_P^{n+1} - (\rho\Phi)_P^n}{\Delta t}$$

3.g Approximation of the Flux terms

Convective Fluxes

The upwind scheme uses the value of Φ from the upstream cell based on the flow direction:

$$(\rho u \Phi)_f = \rho_f u_f \Phi_{\text{upwind}} \quad (30)$$

For example at the east face, if $u_e > 0$, $\Phi_{\text{upwind}} = \Phi_P$, so:

$$(\rho u \Phi)_e = \rho_e u_e \Phi_P \quad (31)$$

The central scheme uses the average value of Φ from adjacent cells on either side of the face:

$$(\rho u \Phi)_f = \rho_f u_f \frac{\Phi_L + \Phi_R}{2} \quad (32)$$

For example, at the east face, $\Phi_L = \Phi_P$ and $\Phi_R = \Phi_E$, so:

$$(\rho u \Phi)_e = \rho_e u_e \frac{\Phi_P + \Phi_E}{2} \quad (33)$$

Diffusive Fluxes

A central difference scheme is used to approximate the gradient of Φ at each face:

$$(\nabla\Phi)_f = \frac{\Phi_R - \Phi_L}{\Delta x_f} \quad (34)$$

For example, at the east face:

$$(\nabla\Phi)_e = \frac{\Phi_E - \Phi_P}{\Delta x_e} \quad (35)$$

3.h Final Discretized Equation

For a uniform grid, the final discretized equation it's given by:

$$\begin{aligned} \frac{(\rho\Phi)_P^{n+1} - (\rho\Phi)_P^n}{\Delta t} \Delta x \Delta y + \Delta y \left[(\rho u\Phi)_e - (\rho u\Phi)_w - \alpha \frac{\Phi_E - \Phi_P}{\Delta x} + \alpha \frac{\Phi_P - \Phi_W}{\Delta x} \right] \\ + \Delta x \left[(\rho v\Phi)_n - (\rho v\Phi)_s - \alpha \frac{\Phi_N - \Phi_P}{\Delta y} + \alpha \frac{\Phi_P - \Phi_S}{\Delta y} \right] = S_{\Phi,P} \Delta x \Delta y \end{aligned}$$

3.i Flux blending

Flux blending is used to combine the advantages of both schemes. The CDS is second-order accurate in space, providing higher accuracy when the flow is smooth however, it can cause numerical oscillations, especially in cases with high Peclet numbers. The UDS is first-order accurate in space but is stable and avoids oscillations in high Peclet number flows. Flux blending provides the stability of the UDS with higher accuracy of the CDS. The blending ratio β allows control over the relative contributions of these two schemes. A good blending ratio can be derived based on the Peclet number:

$$\beta = \frac{Pe}{1 + Pe}$$

This formula ensures:

$$\beta \rightarrow 0 \quad \text{as} \quad Pe \rightarrow 0 \quad (\text{diffusion-dominated})$$

$$\beta \rightarrow 1 \quad \text{as} \quad Pe \rightarrow \infty \quad (\text{convection-dominated})$$

3.j Boundary conditions

At a velocity inlet, a specified velocity boundary condition is given. The mass flux and convective flux are computed using the given velocity and other variables such as density and velocity components. To compute the diffusive flux the gradient of the scalar quantities is computed using one-sided finite differences. At a solid (impermeable) wall, there is no mass flux because the normal velocity component is zero along the wall. Similarly, due to zero velocity along the wall, the convective flux is zero as well. In contrast, the diffusive flux inside the momentum equation is not equal to zero and contributes to momentum transfer due to shear stress. At a symmetry boundary, the boundary normal velocity component is zero, so the mass flux through the boundary is equal to zero as well as the convective flux of all variables is equal to zero. The diffusive fluxes of all scalar quantities is also equal to zero since the scalar gradients normal to the symmetry boundary are zero.

3.k Advantages of a fully coupled and a sequential solution of NSE

A fully coupled sequential solution of the Navier-Stokes equations involves solving the equations step by step, with feedback between the pressure and velocity fields at each time step. The advantage of this approach is that coupling ensures that the pressure and velocity fields are adjusted iteratively to maintain physical consistency, which can improve stability compared to uncoupled methods where only individual variables are solved separately.

4 Correspondence between finite-volume and finite-difference methods

4.a Constant α :

Given the one-dimensional unsteady diffusion equation:

$$\frac{\partial u}{\partial t} = \frac{\partial}{\partial x} \left(\alpha \frac{\partial u}{\partial x} \right) \quad (36)$$

we can obtain its semi-discretization using the finite-volume method with CDS formulation as shown below.

First we integrate over a control volume $[x_{i-1/2}, x_{i+1/2}]$:

$$\int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial u}{\partial t} dx = \int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial}{\partial x} \left(\alpha \frac{\partial u}{\partial x} \right) dx. \quad (37)$$

The time derivative term can be simplified as follows:

$$\int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial u}{\partial t} dx \approx \Delta x \frac{\partial u_i}{\partial t}$$

while for the diffusion term we use the divergence theorem:

$$\int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial}{\partial x} \left(\alpha \frac{\partial u}{\partial x} \right) dx \approx \left[\alpha \frac{\partial u}{\partial x} \right]_{x_{i+1/2}} - \left[\alpha \frac{\partial u}{\partial x} \right]_{x_{i-1/2}}.$$

Then we discretize the fluxes using a centrale difference scheme for $\frac{\partial u}{\partial x}$:

$$\left[\alpha \frac{\partial u}{\partial x} \right]_{x_{i+1/2}} \approx \alpha \frac{u_{i+1} - u_i}{\Delta x}, \quad (38)$$

$$\left[\alpha \frac{\partial u}{\partial x} \right]_{x_{i-1/2}} \approx \alpha \frac{u_i - u_{i-1}}{\Delta x}. \quad (39)$$

In the end we substitute these expressions into the diffusion term and we obtain:

$$\Delta x \frac{\partial u_i}{\partial t} = \frac{\alpha}{\Delta x} (u_{i+1} - 2u_i + u_{i-1}). \quad (40)$$

Now we perform the semi discretization with the finite-difference method. The time derivative is approximated simply as:

$$\frac{\partial u}{\partial t} \approx \frac{\partial u_i}{\partial t}. \quad (41)$$

The second derivative in space is approximated using the central difference scheme:

$$\frac{\partial^2 u}{\partial x^2} \approx \frac{u_{i+1} - 2u_i + u_{i-1}}{\Delta x^2}. \quad (42)$$

Substituting these approximations into the original equation, we obtain:

$$\frac{\partial u_i}{\partial t} = \alpha \frac{u_{i+1} - 2u_i + u_{i-1}}{\Delta x^2}. \quad (43)$$

In order to compare the finite-volume and finite-difference methods, we divide the equation (43) by Δx :

$$\frac{\partial u_i}{\partial t} = \alpha \frac{u_{i+1} - 2u_i + u_{i-1}}{\Delta x^2}. \quad (44)$$

This shows that the finite-volume and finite-difference methods yield the same discretized form of the one-dimensional unsteady diffusion equation for constant α and uniform grid spacing.

4.b Variable $\alpha(x)$:

We start with the one-dimensional unsteady diffusion equation:

$$\frac{\partial u}{\partial t} = \frac{\partial}{\partial x} \left(\alpha(x) \frac{\partial u}{\partial x} \right), \quad (45)$$

where $\alpha = \alpha(x)$ is a spatially varying diffusion coefficient.

We get the semi-discretization using the finite-volume method as follows. First we integrate the equation over a control volume $[x_{i-1/2}, x_{i+1/2}]$:

$$\int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial u}{\partial t} dx = \int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial}{\partial x} \left(\alpha(x) \frac{\partial u}{\partial x} \right) dx. \quad (46)$$

The time derivative term is approximated as:

$$\int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial u}{\partial t} dx \approx \Delta x \frac{\partial u_i}{\partial t}. \quad (47)$$

For the diffusion term we use the divergence theorem:

$$\int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial}{\partial x} \left(\alpha(x) \frac{\partial u}{\partial x} \right) dx = \left[\alpha \frac{\partial u}{\partial x} \right]_{x_{i+1/2}} - \left[\alpha \frac{\partial u}{\partial x} \right]_{x_{i-1/2}}. \quad (48)$$

Then we approximate the fluxes using a central difference scheme as described below. In particular the flux at $x_{i+1/2}$ is approximated as:

$$\left[\alpha \frac{\partial u}{\partial x} \right]_{x_{i+1/2}} \approx \alpha_{i+1/2} \frac{u_{i+1} - u_i}{\Delta x}, \quad (49)$$

and at $x_{i-1/2}$:

$$\left[\alpha \frac{\partial u}{\partial x} \right]_{x_{i-1/2}} \approx \alpha_{i-1/2} \frac{u_i - u_{i-1}}{\Delta x}. \quad (50)$$

Here, $\alpha_{i+1/2}$ and $\alpha_{i-1/2}$ are interpolated as:

$$\alpha_{i+1/2} = \frac{\alpha_i + \alpha_{i+1}}{2}, \quad \alpha_{i-1/2} = \frac{\alpha_i + \alpha_{i-1}}{2}. \quad (51)$$

We combine these expressions into the diffusion term and obtain:

$$\int_{x_{i-1/2}}^{x_{i+1/2}} \frac{\partial}{\partial x} \left(\alpha(x) \frac{\partial u}{\partial x} \right) dx \approx \frac{\alpha_{i+1/2}(u_{i+1} - u_i) - \alpha_{i-1/2}(u_i - u_{i-1})}{\Delta x}. \quad (52)$$

Finally, we substitute these approximations into the original equation, we divide by Δx and obtain:

$$\frac{\partial u_i}{\partial t} = \frac{1}{\Delta x^2} [\alpha_{i+1/2}(u_{i+1} - u_i) - \alpha_{i-1/2}(u_i - u_{i-1})]. \quad (53)$$

Now we perform the semi-discretization with the finite-difference method. The second derivative is approximated using central differences:

$$\left. \frac{\partial}{\partial x} \left(\alpha(x) \frac{\partial u}{\partial x} \right) \right|_i \approx \frac{\alpha_{i+1/2} \frac{u_{i+1} - u_i}{\Delta x} - \alpha_{i-1/2} \frac{u_i - u_{i-1}}{\Delta x}}{\Delta x}.$$

Simplifying, the semi-discretized FDM equation becomes:

$$\frac{\partial u_i}{\partial t} = \frac{1}{\Delta x^2} [\alpha_{i+1/2}(u_{i+1} - u_i) - \alpha_{i-1/2}(u_i - u_{i-1})]. \quad (54)$$

Thus, the FVM and FDM equations are equivalent for second-order accuracy.

In order to show how the flux terms in FVM reduce to the central difference approximations used in FDM we can exploit the given identity: $a - b \equiv \frac{(a-2b+c)}{2} + \frac{(a-c)}{2}$. We first use apply it to $\alpha_{i+1/2}(u_{i+1} - u_i)$:

$$u_{i+1} - u_i = \frac{(u_{i+1} - 2u_i + u_{i-1})}{2} + \frac{(u_{i+1} - u_{i-1})}{2}.$$

Then we substitute this expression into the flux term:

$$\alpha_{i+1/2}(u_{i+1} - u_i) = \alpha_{i+1/2} \left[\frac{(u_{i+1} - 2u_i + u_{i-1})}{2} + \frac{(u_{i+1} - u_{i-1})}{2} \right].$$

We repeat the same procedure for $\alpha_{i-1/2}(u_i - u_{i-1})$.

Using the suggested equivalence, the FVM and FDM formulations reduce to the same semi-discretized equation for second-order accuracy:

$$\frac{\partial u_i}{\partial t} = \frac{1}{\Delta x^2} [\alpha_{i+1/2}(u_{i+1} - u_i) - \alpha_{i-1/2}(u_i - u_{i-1})].$$

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