1. PART A - NUMERICAL METHOD

The one dimensional steady state transport equation

$$U\frac{\partial \phi}{\partial x} = \frac{\partial}{\partial x} \left(\Gamma \frac{\partial \phi}{\partial x} \right) + Q$$

can be discretized with central differences as

$$U\frac{\phi_{i+1} - \phi_{i-1}}{2\Delta x} - \underbrace{\frac{\left(\Gamma\frac{\partial\phi}{\partial x}\right)_{i+1/2} - \left(\Gamma\frac{\partial\phi}{\partial x}\right)_{i-1/2}}_{b}}_{D} = Q_{i}.$$
(1)

Ben Wilfong

Username: bwilfong3

The first derivatives in the numerator of b can be written as

$$\left(\Gamma \frac{\partial \phi}{\partial x}\right)_{i+1/2} = \Gamma_{i+1/2} \frac{\phi_{i+1} - \phi_i}{\Delta x} \quad \text{and} \quad \left(\Gamma \frac{\partial \phi}{\partial x}\right)_{i-1/2} = \Gamma_{i-1/2} \frac{\phi_i - \phi_{i-1}}{\Delta x}.$$
(2)

Substituting eq. (2) into eq. (1) and rearanging gives

$$U \frac{\phi_{i+1} - \phi_{i-1}}{2\Delta x} - \frac{\Gamma_{i+1/2} \frac{\phi_{i+1} - \phi_{i}}{\Delta x} - \Gamma_{i-1/2} \frac{\phi_{i} - \phi_{i-1}}{\Delta x}}{\Delta x} = Q_{i}$$

$$U \frac{\phi_{i+1} - \phi_{i-1}}{2\Delta x} - \frac{\Gamma_{i+1/2} \phi_{i+1} - (\Gamma_{i+1/2} + \Gamma_{i-1/2}) \phi_{i} + \Gamma_{i-1/2} \phi_{i-1}}{(\Delta x)^{2}} = Q_{i}$$

$$\left(\frac{U}{2\Delta x} - \frac{\Gamma_{i+1/2}}{(\Delta x)^{2}}\right) \phi_{i+1} + \left(\frac{\Gamma_{i+1/2} + \Gamma_{i-1/2}}{(\Delta x)^{2}}\right) \phi_{i} - \left(\frac{U}{2\Delta x} + \frac{\Gamma_{i-1/2}}{(\Delta x)^{2}}\right) \phi_{i-1} = Q_{i}.$$
(3)

Equation (3) can be cast as the system of linear equations

$$\underbrace{\begin{bmatrix} 1 \\ -\left(\frac{U}{2\Delta x} + \frac{\Gamma_{1/2}}{(\Delta x)^2}\right) & \frac{\Gamma_{3/2} + \Gamma_{1/2}}{(\Delta x)^2} & \frac{U}{2\Delta x} - \frac{\Gamma_{3/2}}{(\Delta x)^2} \\ \vdots & \vdots & \vdots \\ -\left(\frac{U}{2\Delta x} + \frac{\Gamma_{N-3/2}}{(\Delta x)^2}\right) & \frac{\Gamma_{N-1/2} + \Gamma_{N-3/2}}{(\Delta x)^2} & \frac{U}{2\Delta x} - \frac{\Gamma_{i-1/2}}{(\Delta x)^2} \end{bmatrix}}_{\Delta} \underbrace{\begin{pmatrix} \phi_0 \\ \phi_1 \\ \vdots \\ \phi_{N-1} \\ \phi_N \end{pmatrix}}_{\Phi} = \underbrace{\begin{pmatrix} 1 \\ Q_1 \\ \vdots \\ Q_{N-1} \\ 0 \end{pmatrix}}_{b}$$

for a discretization at $\{x_0, x_1, \dots x_N\} = \{0, \Delta x, 2\Delta x, \dots, N\Delta x\}$. The boundary conditions are given by $\phi_{-1/2} = 1$ and $\phi_{N+1/2} = 0$.

For the cases in which U, Γ , and Q are constant, the solution is found via the following steps:

- (1) Populate the matrix A
- (2) Populate the vector b
- (3) Solve $A\Phi = b$ for the vector Φ using TDMA

When U, Γ , or Q depend on the value of ϕ the problem becomes nonlinear and an iterative approach must be taken. For nonlinear problems, the solution is found via the following steps:

- (1) Make an initial guess for the vector $\Phi^{(0)}$
- (2) Calculate $U(\Phi^{(n)})$, $\Gamma(\Phi^{(n)})$, and $Q(\Phi^{(n)})$
- (3) Populate the matrix A
- (4) Populate the vector b
- (5) Solve $A\Phi^{(n+1)} = b$ for the vector $\Phi^{(n+1)}$ using TDMA
- (6) Repeat steps 2 though 5 until convergence

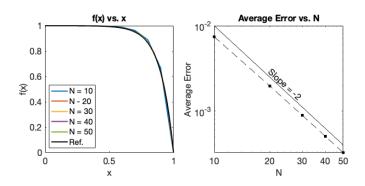


FIGURE 1. Solution and average error for N = 10, 20, 30, 40, 50

For this problem I will use a discretization of $\phi(x) = 1 - x/L$ as an initial guess.

The tridiagonal matrix A is stored in an $N + 1 \times 3$ to take advantage of it's sparsity. The lower, center, and upper diagonal are stored in $A_{1:N,0}$, $A_{0:N,1}$, and $A_{0:N-1,2}$. The tridiagonal linear solve using this data structure is given in algorithm 1.

Algorithm 1 Tridiagonal Matrix Algorithm	
Given: A	$\triangleright N + 1 \times N + 1$
Given: x	$\triangleright N + 1 \times 1$
Given: b	$\triangleright N + 1 \times 1$
$\begin{aligned} &\textbf{for } i = 1: N \textbf{ do} \\ &w = A_{i,0}/A_{i-1,1} \\ &A_{i,0} = A_{i,0} - wA_{i-1,1} \\ &A_{i,1} = A_{i,0} - wA_{i-1,2} \\ &A_{i,0} = w \\ &\textbf{end for} \end{aligned}$	▷ In place LU factorization
$egin{aligned} y_0 &= b_0 \ & ext{for}\ i &= 1: N\ ext{do} \ y_i &= b_i - A_{i,0)y_{i-1}} \ & ext{end for} \end{aligned}$	▷ Forward Substitution
$x_N = y_N/A_{N,1}$ for $i = N-1:-1:0$ do $x_i = \frac{y_i - A_{i,2}x_{i+1}}{A_{i,1}}$ end for	

2. Part B - Linear Problems

Figure 1 shows the solution, average error, and runtime for the case U = 1, $\Gamma = 0.1$, and Q = 0 with N = 10, 20, 30, 40, 50. The error decreases with increasing N as expected. The plot of average error versus N shows that the method is second-order-accurate. This is because the error decreases with a slope of two in loglog space. This order of accuracy is expected given that central differences were used to discretize the derivatives.

3. Part C - Non-Linear Problems