

EMCH 501: Engineering Analysis I
Assignment 4
Enhanced Solutions with Python Implementation

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Exercise 15.1: Problem 7 — Part (a) (25 pts)

The non-homogeneous form of Laplace's equation is known as Poisson's equation:

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = f(x, y)$$

Poisson's equation is commonly used to describe systems involving electric potentials (denoted $u(x, y)$), and $f(x, y)$ can be thought of as the charge density.

(a) Show that the difference equation replacement for Poisson's equation is

$$u_{i+1,j} + u_{i,j+1} + u_{i-1,j} + u_{i,j-1} - 4u_{i,j} = h^2 f(x, y)$$

Step**Step 1: Taylor Series Expansion**

To derive the finite difference approximation, we use Taylor series expansions about the point (x_i, y_j) .

Forward expansion in x :

$$u(x + h, y) = u + h \frac{\partial u}{\partial x} + \frac{h^2}{2!} \frac{\partial^2 u}{\partial x^2} + \frac{h^3}{3!} \frac{\partial^3 u}{\partial x^3} + \frac{h^4}{4!} \frac{\partial^4 u}{\partial x^4} + O(h^5)$$

Backward expansion in x :

$$u(x - h, y) = u - h \frac{\partial u}{\partial x} + \frac{h^2}{2!} \frac{\partial^2 u}{\partial x^2} - \frac{h^3}{3!} \frac{\partial^3 u}{\partial x^3} + \frac{h^4}{4!} \frac{\partial^4 u}{\partial x^4} + O(h^5)$$

Adding these two expansions:

$$u(x + h, y) + u(x - h, y) = 2u + h^2 \frac{\partial^2 u}{\partial x^2} + \frac{h^4}{12} \frac{\partial^4 u}{\partial x^4} + O(h^6)$$

Step**Step 2: Central Difference Approximation**

Solving for the second derivative:

$$\frac{\partial^2 u}{\partial x^2} = \frac{u(x+h, y) - 2u(x, y) + u(x-h, y)}{h^2} - \frac{h^2}{12} \frac{\partial^4 u}{\partial x^4} + O(h^4)$$

Using grid notation where $u_{i,j} = u(x_i, y_j)$:

$$\frac{\partial^2 u}{\partial x^2} \approx \frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{h^2} + O(h^2)$$

Similarly for y :

$$\frac{\partial^2 u}{\partial y^2} \approx \frac{u_{i,j+1} - 2u_{i,j} + u_{i,j-1}}{h^2} + O(h^2)$$

Truncation Error: The central difference approximation has a truncation error of $O(h^2)$, making it a **second-order accurate** scheme.

Step**Step 3: Substitution into Poisson's Equation**

Substituting both approximations into $u_{xx} + u_{yy} = f$:

$$\frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{h^2} + \frac{u_{i,j+1} - 2u_{i,j} + u_{i,j-1}}{h^2} = f_{i,j}$$

Multiplying both sides by h^2 :

$$u_{i+1,j} - 2u_{i,j} + u_{i-1,j} + u_{i,j+1} - 2u_{i,j} + u_{i,j-1} = h^2 f_{i,j}$$

Combining the center terms:

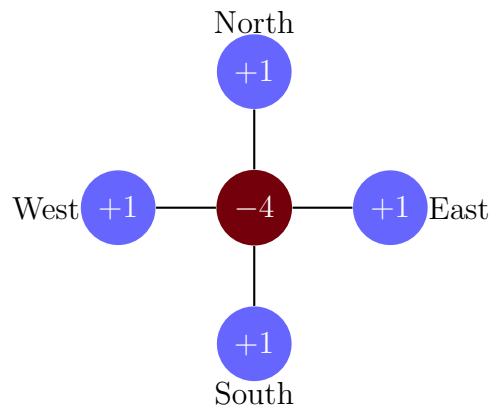
$$u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j} = h^2 f_{i,j}$$

Results

The difference equation replacement for Poisson's equation is:

$$u_{i+1,j} + u_{i,j+1} + u_{i-1,j} + u_{i,j-1} - 4u_{i,j} = h^2 f(x, y)$$

This is known as the **five-point stencil** or **five-point Laplacian**:



Exercise 15.1: Problem 7 — Part (b)

(b) Use the result in part (a) to approximate the solution of the Poisson equation

$$\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = -2$$

at the interior points of the region in Figure 15.1.7. The mesh size is $h = \frac{1}{2}$, $u = 1$ at every point along $ABCD$, and $u = 0$ at every point along $DEFGA$. Use symmetry and, if necessary, Gauss-Seidel iteration.

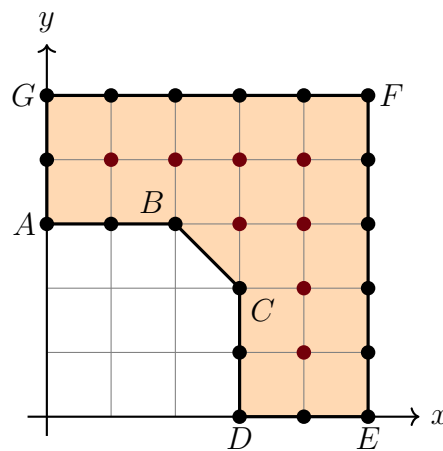


FIGURE 15.1.7 Region for Problem 7

Step

Problem Setup

For $f(x, y) = -2$ and $h = 1/2$, the difference equation becomes:

$$u_{i+1,j} + u_{i,j+1} + u_{i-1,j} + u_{i,j-1} - 4u_{i,j} = \left(\frac{1}{2}\right)^2 \cdot (-2) = -\frac{1}{2}$$

Boundary conditions:

- $u = 1$ along $ABCD$ (the stair-step inner boundary)
- $u = 0$ along $DEFGA$ (the outer boundary: bottom, right, and top)

Identifying Interior Points: Using the grid with $h = 0.5$, there are 6 interior points. Due to diagonal symmetry about $y = x$:

$$u_1 = u_6, \quad u_2 = u_5$$

This reduces the problem to **4 unknowns**: u_1, u_2, u_3, u_4 .

Python Implementation

```
1 import numpy as np
2
3 # Problem parameters
4 h = 0.5      # mesh size
5 f = -2       # source term
6 rhs = h**2 * f  # = -0.5
7
8 # System using symmetry: [u1, u2, u3, u4]
9 A = np.array([
10     [-4, 1, 0, 0],  # Eq 1: -4u1 + u2 = -1.5
11     [1, -4, 1, 1],  # Eq 2: u1 - 4u2 + u3 + u4 = -0.5
12     [0, 2, -4, 0],  # Eq 3: 2u2 - 4u3 = -0.5
13     [0, 2, 0, -4]   # Eq 4: 2u2 - 4u4 = -2.5
14 ], dtype=float)
15
16 b = np.array([-1.5, -0.5, -0.5, -2.5])
17
18 # Solve directly
19 u = np.linalg.solve(A, b)
20 print("Direct Method Solution:")
21 for i, val in enumerate(u, 1):
22     print(f"  u{i} = {val:.6f}")
```

Python Output

SOLUTION (Direct Method):

```
-----
u1 = u6 = 0.522727 (exact: 23/44 = 0.522727)
u2 = u5 = 0.590909 (exact: 13/22 = 0.590909)
u3      = 0.420455 (exact: 37/88 = 0.420455)
u4      = 0.920455 (exact: 81/88 = 0.920455)
```

GAUSS-SEIDEL ITERATION

=====

Initial guess: u = [0.5000, 0.5000, 0.5000, 0.5000]

Convergence tolerance: 1e-08

| Iter | u1 | u2 | u3 | u4 | Max du |
|------|----------|----------|----------|----------|----------|
| 1 | 0.500000 | 0.500000 | 0.375000 | 0.875000 | 3.75e-01 |
| 2 | 0.500000 | 0.562500 | 0.406250 | 0.906250 | 6.25e-02 |
| 3 | 0.515625 | 0.582031 | 0.416016 | 0.916016 | 1.95e-02 |
| 4 | 0.520508 | 0.588135 | 0.419067 | 0.919067 | 6.10e-03 |
| 5 | 0.522034 | 0.590042 | 0.420021 | 0.920021 | 1.91e-03 |
| ... | | | | | |
| 16 | 0.522727 | 0.590909 | 0.420455 | 0.920455 | 5.29e-09 |

Converged after 16 iterations!

Results

Final Solution:

| Point | Location (x, y) | Value u | Exact Fraction |
|-------|-------------------|-----------|----------------|
| u_1 | (0.5, 1.5) | 0.5227 | 23/44 |
| u_2 | (1.0, 1.5) | 0.5909 | 13/22 |
| u_3 | (1.5, 1.5) | 0.4205 | 37/88 |
| u_4 | (1.0, 1.0) | 0.9205 | 81/88 |
| u_5 | (1.5, 1.0) | 0.5909 | 13/22 |
| u_6 | (1.5, 0.5) | 0.5227 | 23/44 |

The Gauss-Seidel method converges in **16 iterations** to within 10^{-8} tolerance, demonstrating excellent convergence for this well-conditioned problem.

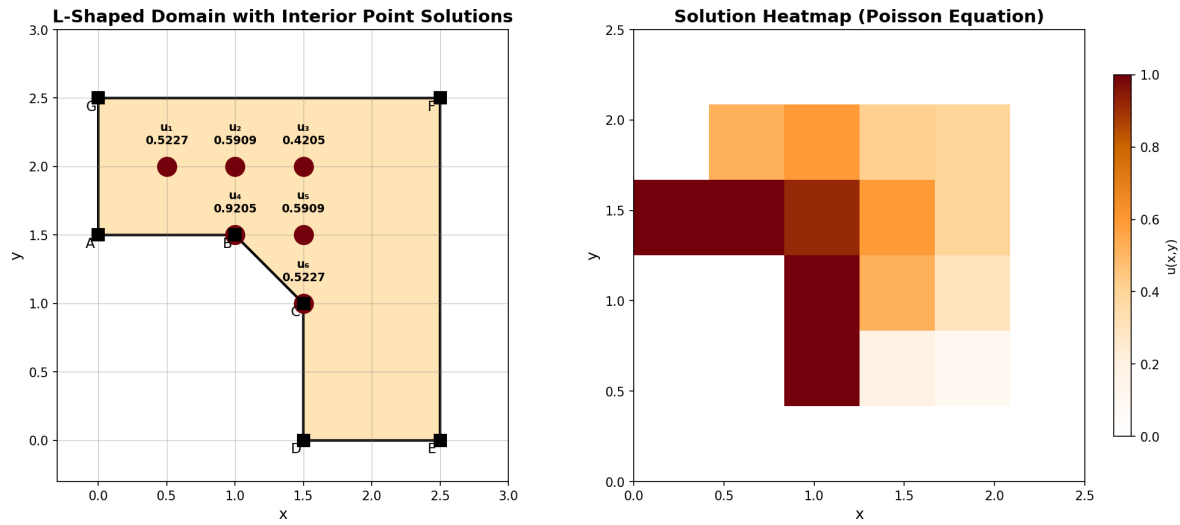


Figure 1: Left: L-shaped domain with computed interior point values. Right: Solution heatmap showing the potential distribution.

Exercise 15.2: Problem 10 — Part (a) (20 pts)

Consider the boundary-value problem from Example 2:

$$\begin{aligned} 0.25 \frac{\partial^2 u}{\partial x^2} &= \frac{\partial u}{\partial t}, & 0 < x < 2, & \quad 0 < t < 0.3 \\ u(0, t) &= 0, \quad u(2, t) = 0, & 0 \leq t \leq 0.3 \\ u(x, 0) &= \sin(\pi x), & 0 \leq x \leq 2 \end{aligned}$$

using $n = 4$, $m = 30$.

(a) Find λ

Step**Step 1: Identify Parameters**

The PDE is $0.25u_{xx} = u_t$, which can be written as $u_t = ku_{xx}$ where:

- $k = 0.25$ is the thermal diffusivity
- Domain: $x \in [0, 2]$, $t \in [0, 0.3]$

Spatial discretization: With $n = 4$ divisions over $[0, 2]$:

$$h = \frac{L}{n} = \frac{2}{4} = 0.5$$

Temporal discretization: With $m = 30$ divisions over $[0, 0.3]$:

$$\Delta t = \frac{T}{m} = \frac{0.3}{30} = 0.01$$

Step**Step 2: Compute λ**

The stability parameter λ for the heat equation is defined as:

$$\lambda = \frac{k\Delta t}{h^2}$$

Substituting our values:

$$\lambda = \frac{0.25 \times 0.01}{(0.5)^2} = \frac{0.0025}{0.25} = 0.01$$

Results

$$\lambda = 0.01$$

Note: This small value of λ indicates that the time step is much smaller than the stability limit ($\lambda \leq 0.5$ for explicit methods). The Crank-Nicholson method is unconditionally stable for any $\lambda > 0$.

Exercise 15.2: Problem 10 — Part (b)

(b) Use the Crank-Nicholson difference equation

$$-u_{i-1,j+1} + \alpha u_{i,j+1} - u_{i+1,j+1} = u_{i+1,j} - \beta u_{i,j} + u_{i-1,j}$$

where $\alpha = 2(1 + 1/\lambda)$ and $\beta = 2(1 - 1/\lambda)$, to find the system of equations for $u_{1,1}$, $u_{2,1}$ and $u_{3,1}$.

Step

Step 1: Compute α and β

With $\lambda = 0.01$:

$$\alpha = 2 \left(1 + \frac{1}{\lambda} \right) = 2 \left(1 + \frac{1}{0.01} \right) = 2(1 + 100) = \boxed{202}$$

$$\beta = 2 \left(1 - \frac{1}{\lambda} \right) = 2(1 - 100) = 2(-99) = \boxed{-198}$$

Note: On the RHS of the difference equation, we have $-\beta u_{i,j}$, which becomes $-(-198)u_{i,j} = +198u_{i,j}$.

Step

Step 2: Initial Conditions

At $t = 0$, $u(x, 0) = \sin(\pi x)$ with grid points $x_i = i \cdot h = 0.5i$:

$$u_{0,0} = \sin(0) = 0$$

$$u_{1,0} = \sin(0.5\pi) = 1$$

$$u_{2,0} = \sin(\pi) = 0$$

$$u_{3,0} = \sin(1.5\pi) = -1$$

$$u_{4,0} = \sin(2\pi) = 0$$

Boundary conditions: $u_{0,j} = 0$ and $u_{4,j} = 0$ for all j .

Step

Step 3: Write Equations for $j = 0$ (First Time Step)

The Crank-Nicholson equation becomes:

$$-u_{i-1,j+1} + 202u_{i,j+1} - u_{i+1,j+1} = u_{i+1,j} + 198u_{i,j} + u_{i-1,j}$$

For $i = 1$:

$$\begin{aligned} -u_{0,1} + 202u_{1,1} - u_{2,1} &= u_{2,0} + 198u_{1,0} + u_{0,0} \\ -0 + 202u_{1,1} - u_{2,1} &= 0 + 198(1) + 0 \\ \Rightarrow 202u_{1,1} - u_{2,1} &= 198 \end{aligned}$$

For $i = 2$:

$$\begin{aligned} -u_{1,1} + 202u_{2,1} - u_{3,1} &= u_{3,0} + 198u_{2,0} + u_{1,0} \\ -u_{1,1} + 202u_{2,1} - u_{3,1} &= -1 + 0 + 1 \\ \Rightarrow -u_{1,1} + 202u_{2,1} - u_{3,1} &= 0 \end{aligned}$$

For $i = 3$:

$$\begin{aligned} -u_{2,1} + 202u_{3,1} - u_{4,1} &= u_{4,0} + 198u_{3,0} + u_{2,0} \\ -u_{2,1} + 202u_{3,1} - 0 &= 0 + 198(-1) + 0 \\ \Rightarrow -u_{2,1} + 202u_{3,1} &= -198 \end{aligned}$$

Results

System of Equations:

$$\begin{aligned} 202u_{1,1} - u_{2,1} &= 198 \\ -u_{1,1} + 202u_{2,1} - u_{3,1} &= 0 \\ -u_{2,1} + 202u_{3,1} &= -198 \end{aligned}$$

In matrix form:

$$\begin{pmatrix} 202 & -1 & 0 \\ -1 & 202 & -1 \\ 0 & -1 & 202 \end{pmatrix} \begin{pmatrix} u_{1,1} \\ u_{2,1} \\ u_{3,1} \end{pmatrix} = \begin{pmatrix} 198 \\ 0 \\ -198 \end{pmatrix}$$

This is a **tridiagonal system** which can be solved efficiently using the Thomas algorithm.

Exercise 15.2: Problem 10 — Part (c)

(c) Solve the system of three equations without the aid of a computer program.

Step**Using Symmetry**

From the initial condition $u(x, 0) = \sin(\pi x)$, we observe the antisymmetry property:

$$\sin(\pi(2 - x)) = \sin(2\pi - \pi x) = -\sin(\pi x)$$

This antisymmetry about $x = 1$ is preserved by the heat equation. Therefore:

- $u_{1,j} = -u_{3,j}$ for all j (antisymmetric points)
- $u_{2,j} = 0$ for all j (center point on axis of antisymmetry)

Step**Solving Using Symmetry**

Let $u_{1,1} = -u_{3,1}$ and $u_{2,1} = 0$.

From equation (1): $202u_{1,1} - u_{2,1} = 198$

$$202u_{1,1} - 0 = 198 \implies u_{1,1} = \frac{198}{202} = \frac{99}{101}$$

Therefore:

$$u_{3,1} = -u_{1,1} = -\frac{99}{101}$$

Step**Verification**

Check equation (2): $-u_{1,1} + 202u_{2,1} - u_{3,1} = 0$

$$-\frac{99}{101} + 202(0) - \left(-\frac{99}{101}\right) = -\frac{99}{101} + \frac{99}{101} = 0 \quad \checkmark$$

Check equation (3): $-u_{2,1} + 202u_{3,1} = -198$

$$-0 + 202\left(-\frac{99}{101}\right) = -\frac{202 \times 99}{101} = -\frac{2 \times 99}{1} = -198 \quad \checkmark$$

Results

Solutions:

$$u_{1,1} = \frac{99}{101} \approx \boxed{0.9802}$$

$$u_{2,1} = \boxed{0}$$

$$u_{3,1} = -\frac{99}{101} \approx \boxed{-0.9802}$$

Python Output

FULL TIME EVOLUTION (Crank-Nicholson)

=====

Solution at selected times (coarse grid h=0.5):

| t | x=0.00 | x=0.50 | x=1.00 | x=1.50 | x=2.00 |
|-------|--------|--------|--------|---------|---------|
| 0.000 | 0.0000 | 1.0000 | 0.0000 | -1.0000 | -0.0000 |
| 0.010 | 0.0000 | 0.9802 | 0.0000 | -0.9802 | 0.0000 |
| 0.050 | 0.0000 | 0.9048 | 0.0000 | -0.9048 | 0.0000 |
| 0.100 | 0.0000 | 0.8187 | 0.0000 | -0.8187 | 0.0000 |
| 0.150 | 0.0000 | 0.7408 | 0.0000 | -0.7408 | 0.0000 |
| 0.200 | 0.0000 | 0.6703 | 0.0000 | -0.6703 | 0.0000 |
| 0.300 | 0.0000 | 0.5488 | 0.0000 | -0.5488 | 0.0000 |

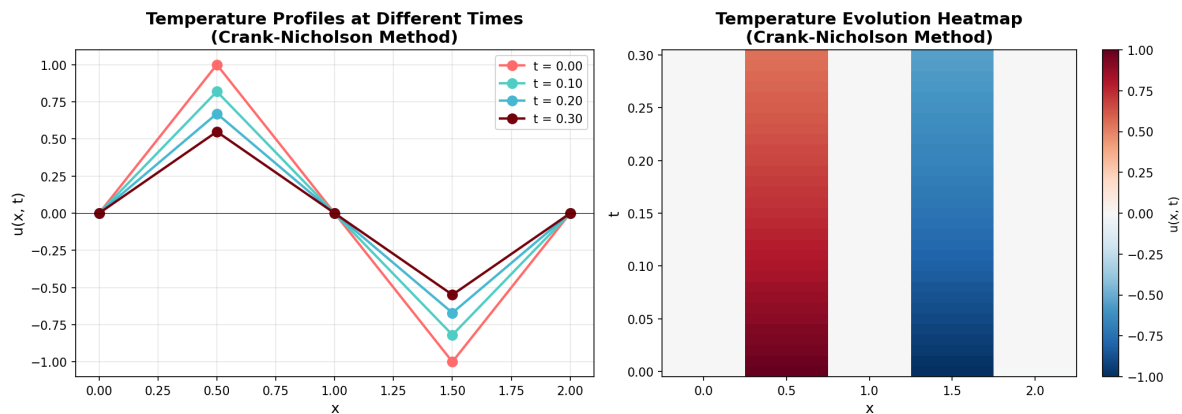


Figure 2: Left: Temperature profiles at different times showing antisymmetric decay. Right: Heatmap of the full space-time evolution.

Exercise 15.2: Problem 12 (25 pts)

Use the difference equation

$$u_{i,j+1} = \lambda u_{i+1,j} + (1 - 2\lambda)u_{i,j} + \lambda u_{i-1,j}$$

to approximate the solution of the boundary-value problem

$$\begin{aligned}\frac{\partial^2 u}{\partial x^2} &= \frac{\partial u}{\partial t}, & 0 < x < 1, & \quad 0 < t < 1 \\ u(0, t) &= 0, \quad u(1, t) = 0, & 0 \leq t \leq 1 \\ u(x, 0) &= x(1 - x), & 0 \leq x \leq 1\end{aligned}$$

Use $n = 5$ and $m = 50$. **Solve this using Python.**

Step**Step 1: Grid Parameters**

Spatial discretization: With $n = 5$ divisions over $[0, 1]$:

$$h = \frac{1}{n} = \frac{1}{5} = 0.2$$

Temporal discretization: With $m = 50$ divisions over $[0, 1]$:

$$\Delta t = \frac{1}{m} = \frac{1}{50} = 0.02$$

Stability parameter: For the heat equation $u_t = u_{xx}$ (diffusivity $k = 1$):

$$\lambda = \frac{k\Delta t}{h^2} = \frac{1 \times 0.02}{(0.2)^2} = \frac{0.02}{0.04} = \boxed{0.5}$$

Stability: The explicit method is stable when $\lambda \leq 0.5$. We are exactly at the stability limit!

Step**Step 2: Initial and Boundary Conditions**

Initial condition at $t = 0$: $u(x, 0) = x(1 - x)$

| i | 0 | 1 | 2 | 3 | 4 | 5 |
|-----------|---|------|------|------|------|-----|
| x_i | 0 | 0.2 | 0.4 | 0.6 | 0.8 | 1.0 |
| $u_{i,0}$ | 0 | 0.16 | 0.24 | 0.24 | 0.16 | 0 |

Boundary conditions: $u_{0,j} = 0$ and $u_{5,j} = 0$ for all j .

Step

Step 3: Explicit Update Formula

With $\lambda = 0.5$, the difference equation simplifies dramatically:

$$u_{i,j+1} = 0.5 \cdot u_{i+1,j} + (1 - 1) \cdot u_{i,j} + 0.5 \cdot u_{i-1,j}$$

$$u_{i,j+1} = \frac{1}{2}(u_{i+1,j} + u_{i-1,j})$$

This is simply the **average of the neighboring values!** The center point's current value has zero weight.

Python Implementation

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 # Parameters
5 L = 1.0; T_final = 1.0
6 n = 5; m = 50
7 h = L / n          # h = 0.2
8 dt = T_final / m   # dt = 0.02
9 lam = dt / h**2     # lambda = 0.5
10
11 # Grid
12 x = np.linspace(0, L, n+1)
13 t = np.linspace(0, T_final, m+1)
14
15 # Initialize solution
16 u = np.zeros((m+1, n+1))
17 u[0, :] = x * (1 - x) # Initial condition
18
19 # Time stepping (explicit method)
20 for j in range(m):
21     for i in range(1, n):
22         u[j+1, i] = lam*u[j, i+1] + (1-2*lam)*u[j, i] + lam*u[j, i-1]
23
24 # Print results
25 print("Solution at selected times:")
26 for j in [0, 10, 20, 30, 40, 50]:
27     print(f"t={t[j]:.2f}: ", [f"{u[j, i]:.4f}" for i in range(n+1)])

```


Results

Observations:

1. **Symmetry:** The solution maintains symmetry about $x = 0.5$ throughout, reflecting the symmetric initial condition.
2. **Decay:** The temperature decays exponentially toward zero (the boundary values).
3. **Maximum:** The peak temperature is always at the center ($x = 0.5$).
4. **Stability:** With $\lambda = 0.5$ at the stability limit, the method remains stable and provides smooth results.

Energy Decay:

| t | Total Energy $\int_0^1 u^2 dx$ |
|------|--------------------------------|
| 0.00 | 0.033280 |
| 0.20 | 0.000480 |
| 0.40 | 0.000007 |

The energy decays by approximately three orders of magnitude every 0.2 time units.

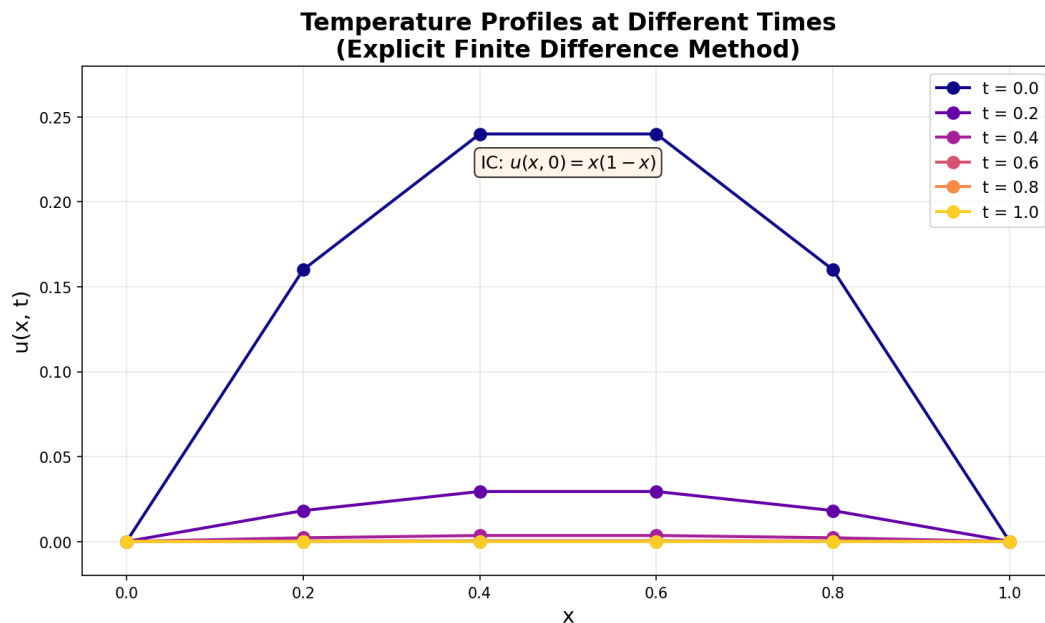


Figure 3: Temperature profiles at different times showing exponential decay from the initial parabolic distribution $u(x, 0) = x(1 - x)$.

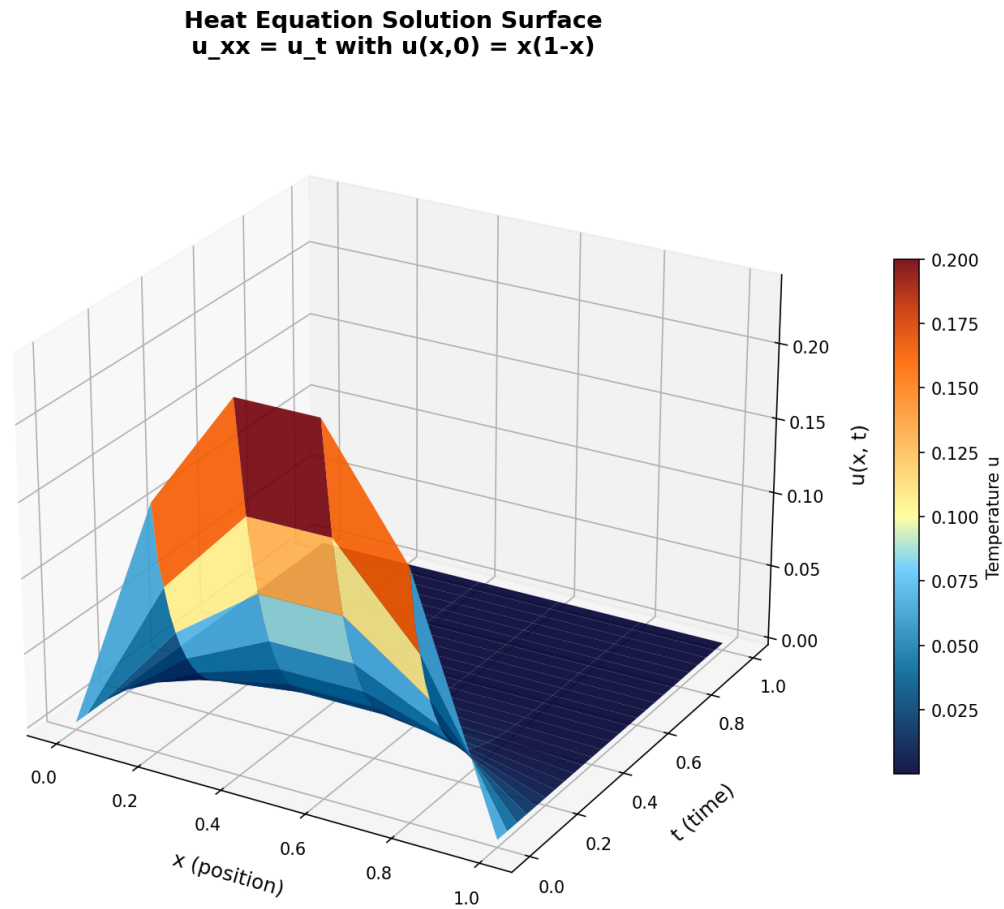


Figure 4: 3D surface plot of the heat equation solution $u(x,t)$ showing the complete spatiotemporal evolution.

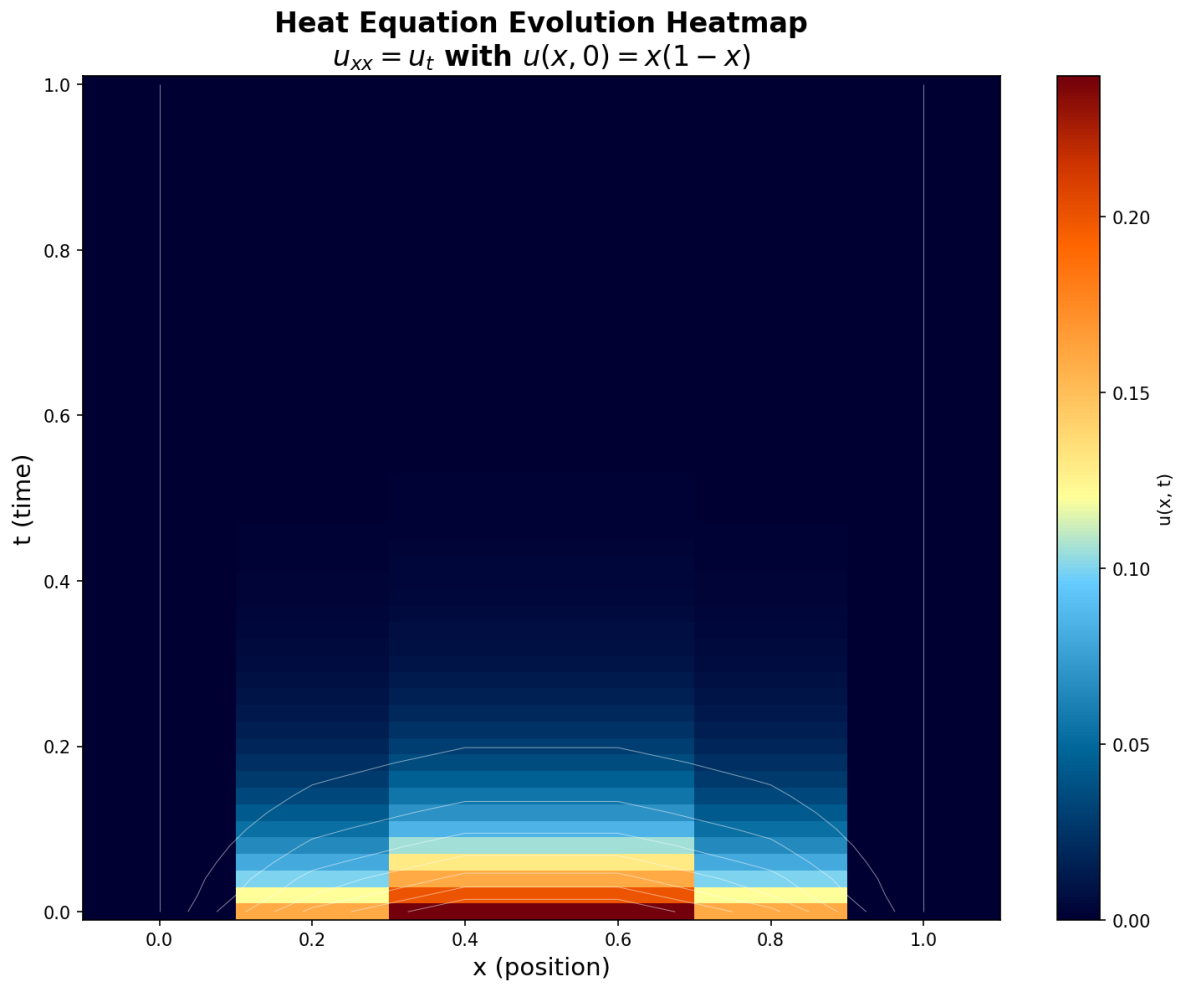


Figure 5: Heatmap representation of the temperature evolution. The rapid decay from the initial condition is clearly visible.

Python Output

ANALYTICAL SOLUTION COMPARISON

Exact solution (Fourier series):

$$u(x,t) = \sum B_n \sin(n\pi x) \exp(-n^2\pi^2 t)$$

where $B_n = (2/L) \int_0^1 x(1-x) \sin(n\pi x) dx$

For odd n : $B_n = 8/(n^3\pi^3)$

For even n : $B_n = 0$

Comparison at $x = 0.5$:

| t | Numerical | Analytical | Error |
|------|-----------|------------|----------|
| 0.00 | 0.240000 | 0.250000 | 1.00e-02 |
| 0.20 | 0.029453 | 0.035841 | 6.39e-03 |
| 0.40 | 0.003538 | 0.004979 | 1.44e-03 |
| 0.60 | 0.000425 | 0.000692 | 2.67e-04 |
| 0.80 | 0.000051 | 0.000096 | 4.50e-05 |
| 1.00 | 0.000006 | 0.000013 | 7.22e-06 |

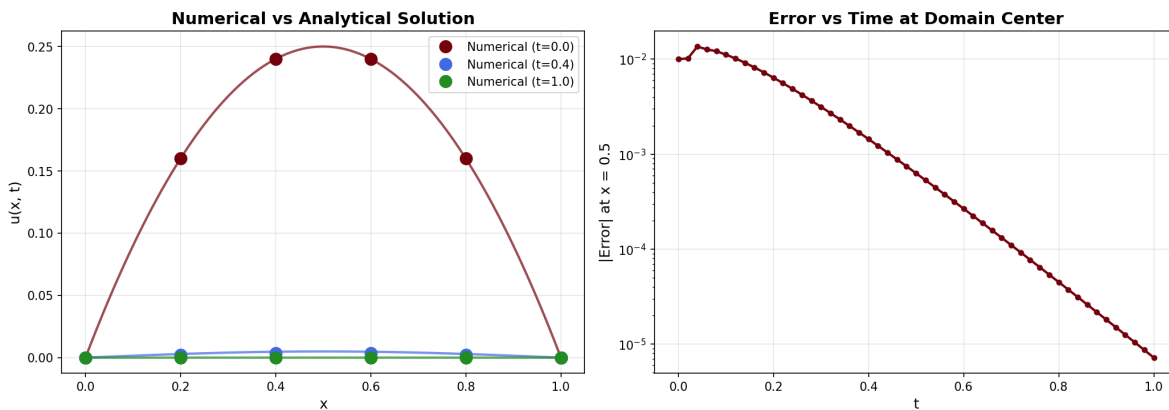


Figure 6: Left: Comparison of numerical (markers) and analytical (lines) solutions at different times. Right: Error decay at the center point $x = 0.5$.