

# EMCH 501: Engineering Analysis I

## Assignment 4

Enhanced Solutions with Python Implementation

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Due: December 8, 2025

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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:

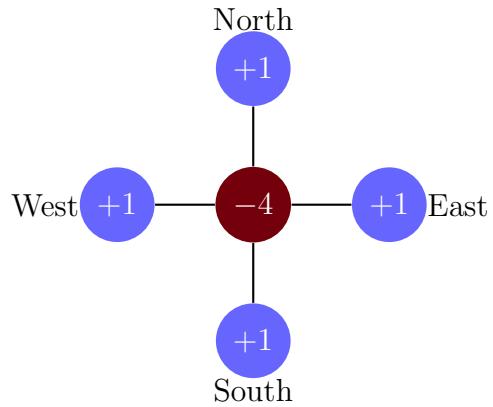
$$u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{i,j} = h^2 f_{i,j}$$

## 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**:

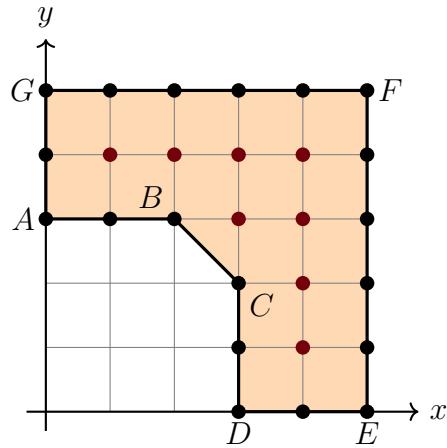


**Exercise 15.1: Problem 7 — Part (b)**

(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.

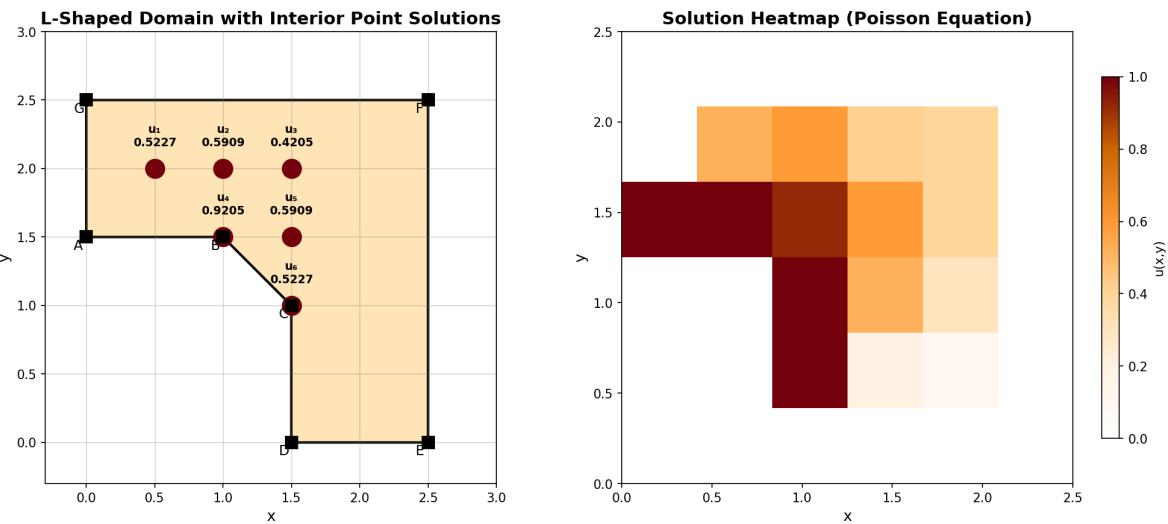


Figure 1: Left: L-shaped domain with computed interior point values. Right: Solution heatmap showing the potential distribution.

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

Consider the boundary-value problem from Example 2:

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

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

(a) Find  $\lambda$

**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  $\lambda$** 

The stability parameter  $\lambda$  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  $\lambda$  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  $\alpha$  and  $\beta$** 

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)
- $u_{2,1} = 0$  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:

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

$$u_{2,1} = [0]$$

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

## 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

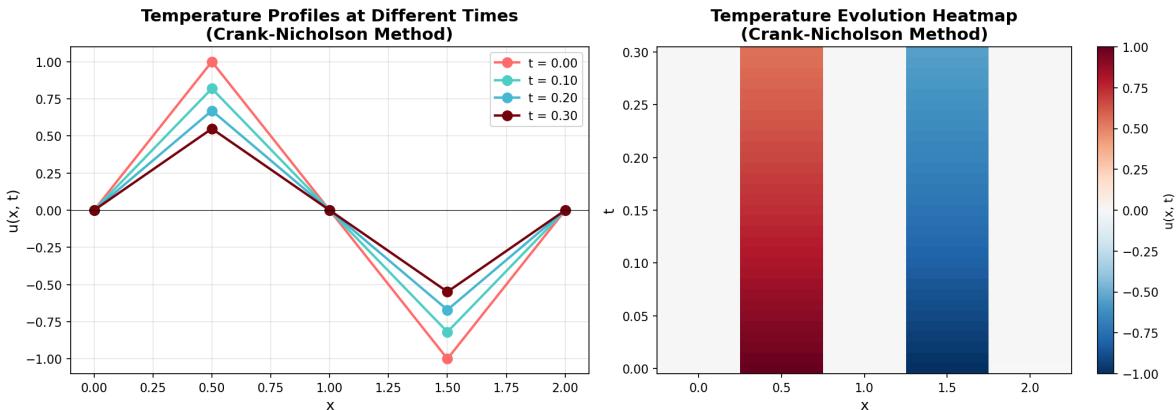


Figure 2: Left: Temperature profiles at different times showing antisymmetric decay. Right: Heatmap of the full space-time evolution.

**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

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

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)])

```

### Python Output

```
STEP 4: First Few Time Steps (Manual)
```

```
-----  
j = 0 -> j = 1 (t = 0.00 -> t = 0.02):  
u_{1,1} = 0.5 x (0.2400 + 0.0000) = 0.1200  
u_{2,1} = 0.5 x (0.2400 + 0.1600) = 0.2000  
u_{3,1} = 0.5 x (0.1600 + 0.2400) = 0.2000  
u_{4,1} = 0.5 x (0.0000 + 0.2400) = 0.1200
```

```
j = 1 -> j = 2 (t = 0.02 -> t = 0.04):  
u_{1,2} = 0.5 x (0.2000 + 0.0000) = 0.1000  
u_{2,2} = 0.5 x (0.2000 + 0.1200) = 0.1600  
u_{3,2} = 0.5 x (0.1200 + 0.2000) = 0.1600  
u_{4,2} = 0.5 x (0.0000 + 0.2000) = 0.1000
```

#### NUMERICAL RESULTS

t	x=0.0	x=0.2	x=0.4	x=0.6	x=0.8	x=1.0
0.00	0.0000	0.1600	0.2400	0.2400	0.1600	0.0000
0.20	0.0000	0.0182	0.0295	0.0295	0.0182	0.0000
0.40	0.0000	0.0022	0.0035	0.0035	0.0022	0.0000
0.60	0.0000	0.0003	0.0004	0.0004	0.0003	0.0000
0.80	0.0000	0.0000	0.0001	0.0001	0.0000	0.0000
1.00	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000

## 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.

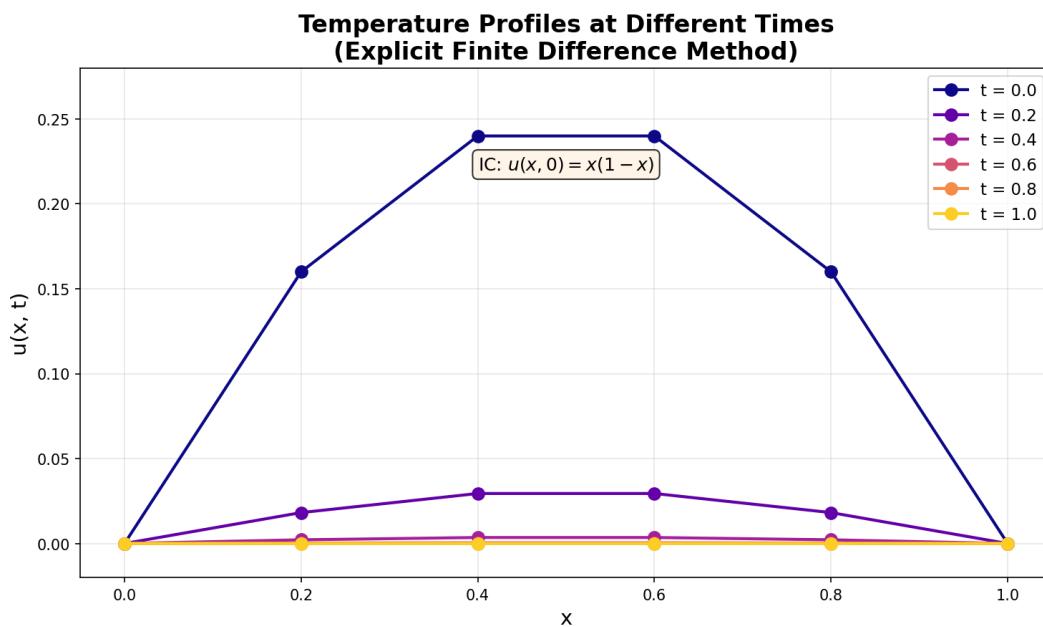


Figure 3: Temperature profiles at different times showing exponential decay from the initial parabolic distribution  $u(x, 0) = x(1 - x)$ .

**Heat Equation Solution Surface**  
 $u_{xx} = u_t$  with  $u(x,0) = x(1-x)$

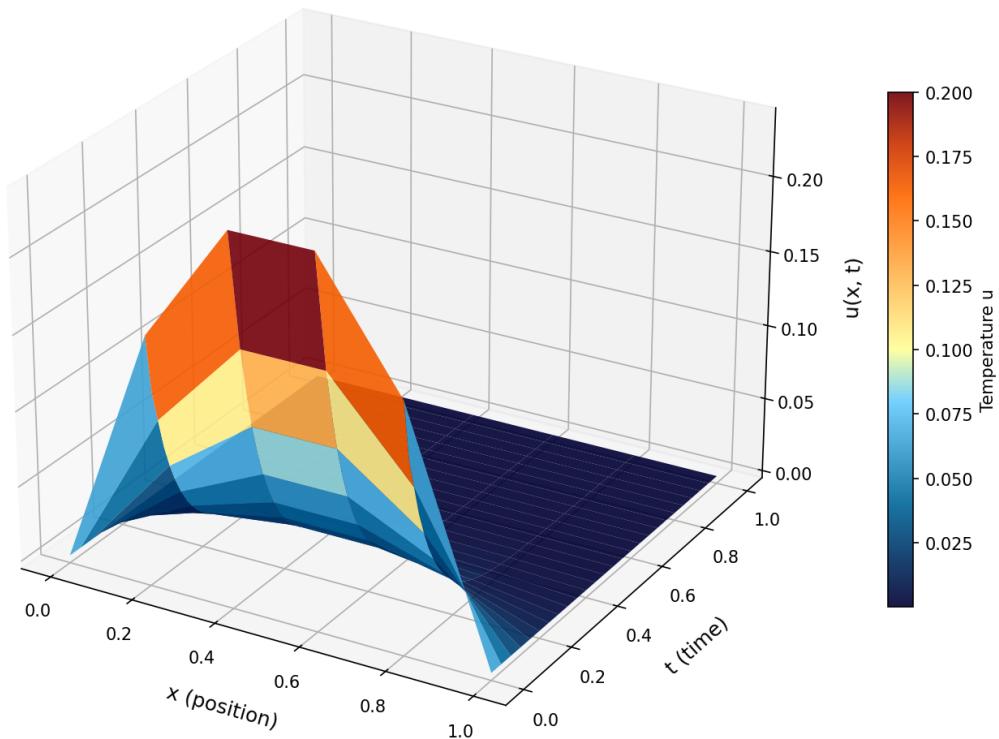


Figure 4: 3D surface plot of the heat equation solution  $u(x, t)$  showing the complete spatiotemporal evolution.

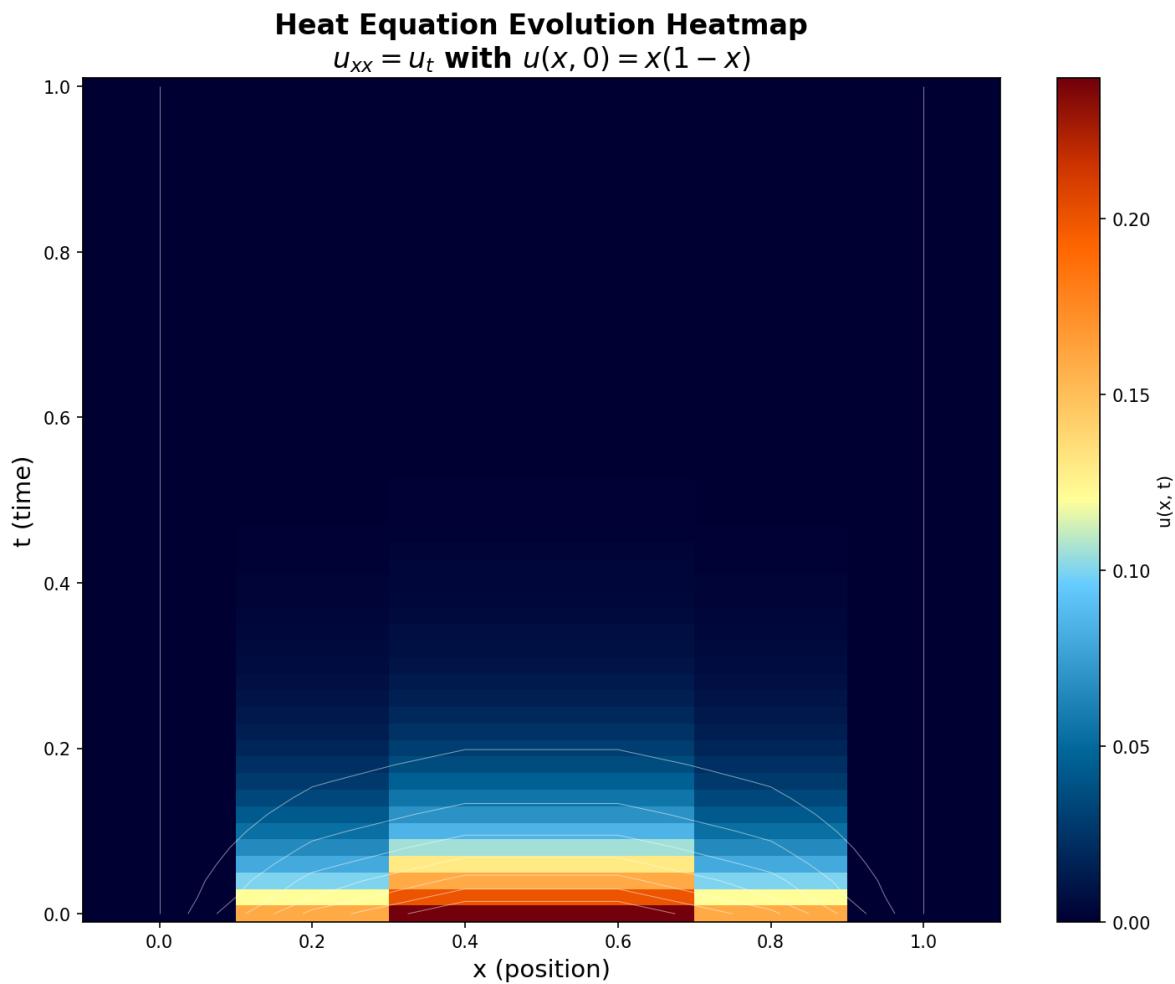


Figure 5: Heatmap representation of the temperature evolution. The rapid decay from the initial condition is clearly visible.

## 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) integral_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

```

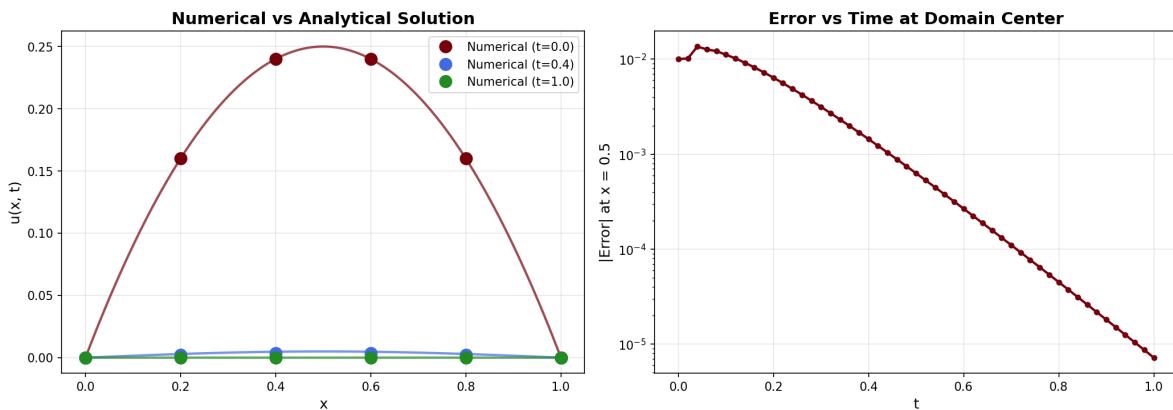


Figure 6: Left: Comparison of numerical (markers) and analytical (lines) solutions at different times. Right: Error decay at the center point  $x = 0.5$ .