

06_tasks

October 28, 2025

Write a **Python** program that implements both the **Explicit Method (Forward Euler)** and the **Crank–Nicolson Method** to solve the following **parabolic partial differential equation (PDE)**:

$$\begin{aligned}\text{PDE: } \quad \frac{\partial u}{\partial t} &= \frac{\partial^2 u}{\partial x^2}, \quad -1 < x < 1, \quad t > 0, \\ \text{Initial Condition: } \quad u(x, 0) &= e^{-x^2/(0.2)^2}, \\ \text{Boundary Conditions: } \quad u(-1, t) &= 0, \\ &u(1, t) = 0.\end{aligned}$$

Implement both time-stepping schemes in Python.

```
[1]: import numpy as np
import matplotlib.pyplot as plt
import scipy.sparse as sparse
import scipy.sparse.linalg as linalg
```

```
# --- PDE Parameters ---
T_final = 0.03 # Final time

# Initial condition
def initial_condition(x):
    return np.exp(-x**2 / (0.2**2))

# Boundary conditions
g_0 = lambda t: 0.0
g_1 = lambda t: 0.0
```

```
[28]: def solve_explicit(m, T_final):
    # 1. Create the spatial grid x from -1.0 to 1.0 with m points.
    x = np.linspace(-1.0, 1.0, m)
    delta_x = x[1] - x[0]
    assert m >= 3, "m debe ser 3 para tener puntos interiores"

    # 2. Set the time step delta_t from the stability condition (dt <= 0.5 *
    ↪ delta_x**2).
```

```

delta_t_stable = 0.5 * (delta_x ** 2)
N = int(np.ceil(T_final / delta_t_stable))
#delta_t = T_final / N
delta_t = 0.5 * delta_x**2
t = np.linspace(0.0, T_final, N + 1)

# 3. Initialize the solution array U with the correct shape.
U = np.zeros((N + 1, m))
U[0, :] = initial_condition(x)

r = delta_t / (delta_x ** 2)

# enforce boundary at t=0
U[0, 0] = g_0(t[0])
U[0, -1] = g_1(t[0])

# 4. Time-stepping loop:
for n in range(N):
    # known boundary at time t[n]
    U[n, 0] = g_0(t[n])
    U[n, -1] = g_1(t[n])

    # interior update i = 1..m-2
    U[n + 1, 1:-1] = U[n, 1:-1] + r * (U[n, 0:-2] - 2.0 * U[n, 1:-1] + U[n, 1
↪2:])

    # apply boundary at time t[n+1]
    U[n + 1, 0] = g_0(t[n + 1])
    U[n + 1, -1] = g_1(t[n + 1])

# 5. Return x, t, and U as results.
return x, t, U

```

```

[29]: def solve_crank_nicolson(m, T_final):
    # 1. Create the spatial grid x from -1.0 to 1.0 with m points.
    x = np.linspace(-1.0, 1.0, m)
    delta_x = x[1] - x[0]

    # 2. Define the time step delta_t (e.g., proportional to delta_x for
↪stability).
    #delta_t = 0.1 * delta_x

    delta_t = 0.5 * delta_x**2
    N = int(np.ceil(T_final / delta_t))
    delta_t = T_final / N
    t = np.linspace(0.0, T_final, N + 1)

```

```

# 3. Loop over each time step n:
U = np.zeros((N + 1, m))
U[0, :] = initial_condition(x)

# interior points
M = m - 2
r = delta_t / (2.0 * delta_x**2)

# tridiagonal matrix
main = -2.0 * np.ones(M)
off = 1.0 * np.ones(M - 1)
Ttri = np.diag(main) + np.diag(off, 1) + np.diag(off, -1)

A = np.eye(M) - r * Ttri          # Lado implícito
B = np.eye(M) + r * Ttri          # Lado explícito

for n in range(N):
    # BC in time n and n+1
    gL_n, gR_n = g_0(t[n]),      g_1(t[n])
    gL_np1, gR_np1 = g_0(t[n+1]), g_1(t[n+1])

    # BC in time n
    U[n, 0] = gL_n
    U[n, -1] = gR_n

    # interior at time n
    U_int = U[n, 1:-1]

    # BC
    bc = np.zeros(M)
    bc[0] += r * (gL_np1 + gL_n)
    bc[-1] += r * (gR_np1 + gR_n)

    # RHS
    b = B @ U_int + bc
    U_next_int = np.linalg.solve(A, b)

    # capa n+1
    U[n+1, 0] = gL_np1
    U[n+1, -1] = gR_np1
    U[n+1, 1:-1] = U_next_int

# 4. Return x, t, and U as results.

return x, t, U

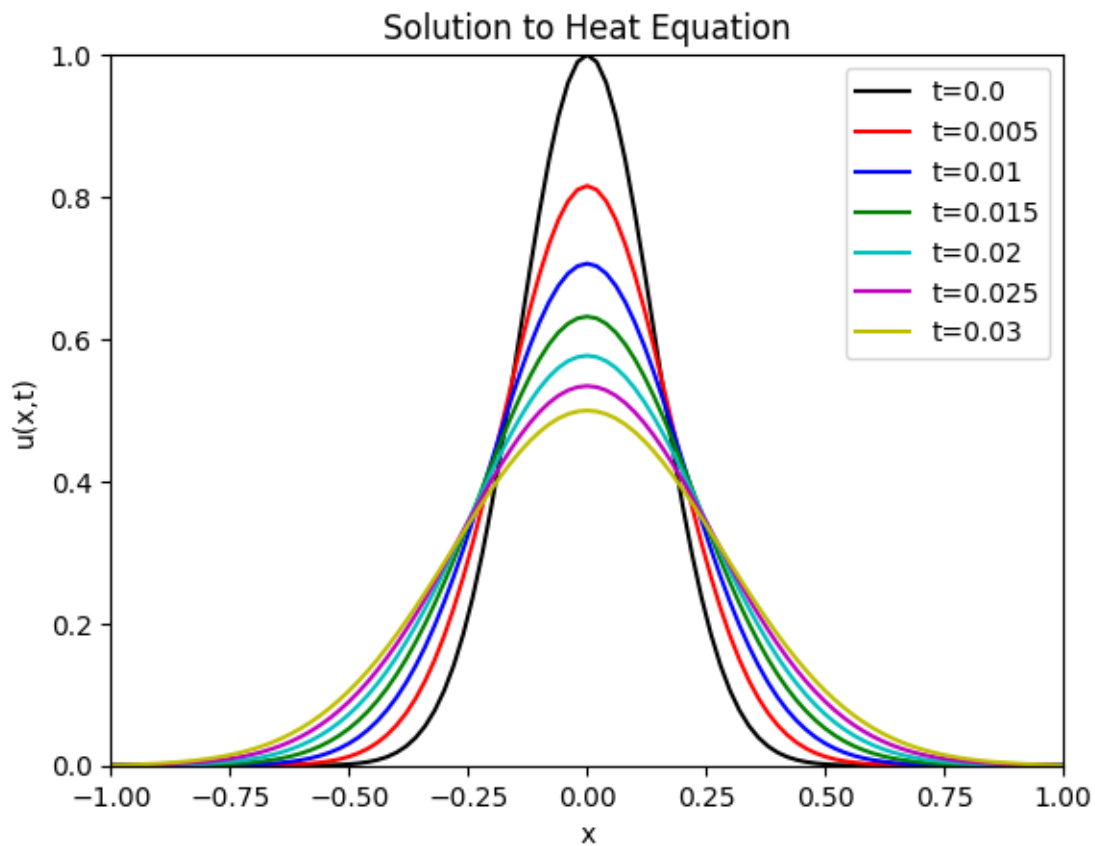
```

a) Plot the numerical solution at several time steps (e.g., $t = 0, 0.005, 0.01, 0.015, 0.02, 0.025, 0.03$) to visualize the diffusion process and compare the behavior of the **Explicit** and **Crank–Nicolson**

methods.

```
[30]: x, t, U = solve_explicit(m=101, T_final=T_final)

# Plot a few solutions
colors = ['k', 'r', 'b', 'g', 'c', 'm', 'y']
fig = plt.figure()
axes = fig.add_subplot(1, 1, 1)
for (i, n) in enumerate((0, 25, 50, 75, 100, 125, 150)):
    axes.plot(x, U[n, :], colors[i], label='t=%s' % np.round(t[n], 4))
    axes.set_xlabel("x")
    axes.set_ylabel("u(x,t)")
    axes.set_title("Solution to Heat Equation")
    axes.set_xlim([-1,1])
    axes.set_ylim([0.0, 1.0])
axes.legend()
plt.show()
```



```
[31]: from matplotlib import cm
```

```

# Create meshgrid for plotting
X, T = np.meshgrid(x, t)

# Create figure and 3D axis
fig = plt.figure(figsize=(8, 6))
ax = fig.add_subplot(111, projection='3d')

# Plot the surface
surf = ax.plot_surface(X, T, U, cmap=cm.viridis, edgecolor='none')

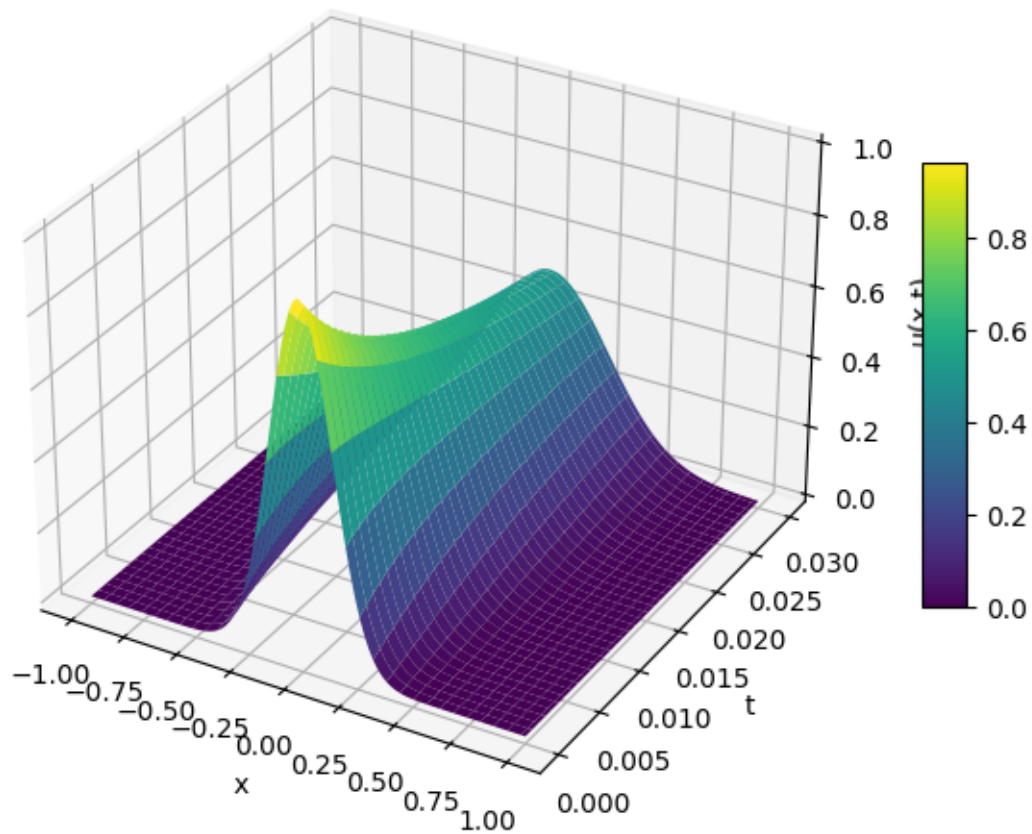
# Labels and title
ax.set_xlabel('x')
ax.set_ylabel('t')
ax.set_zlabel('u(x,t)')
ax.set_title('Heat Equation Solution (Crank-Nicolson Method)')

# Optional color bar
fig.colorbar(surf, shrink=0.5, aspect=10)

plt.show()

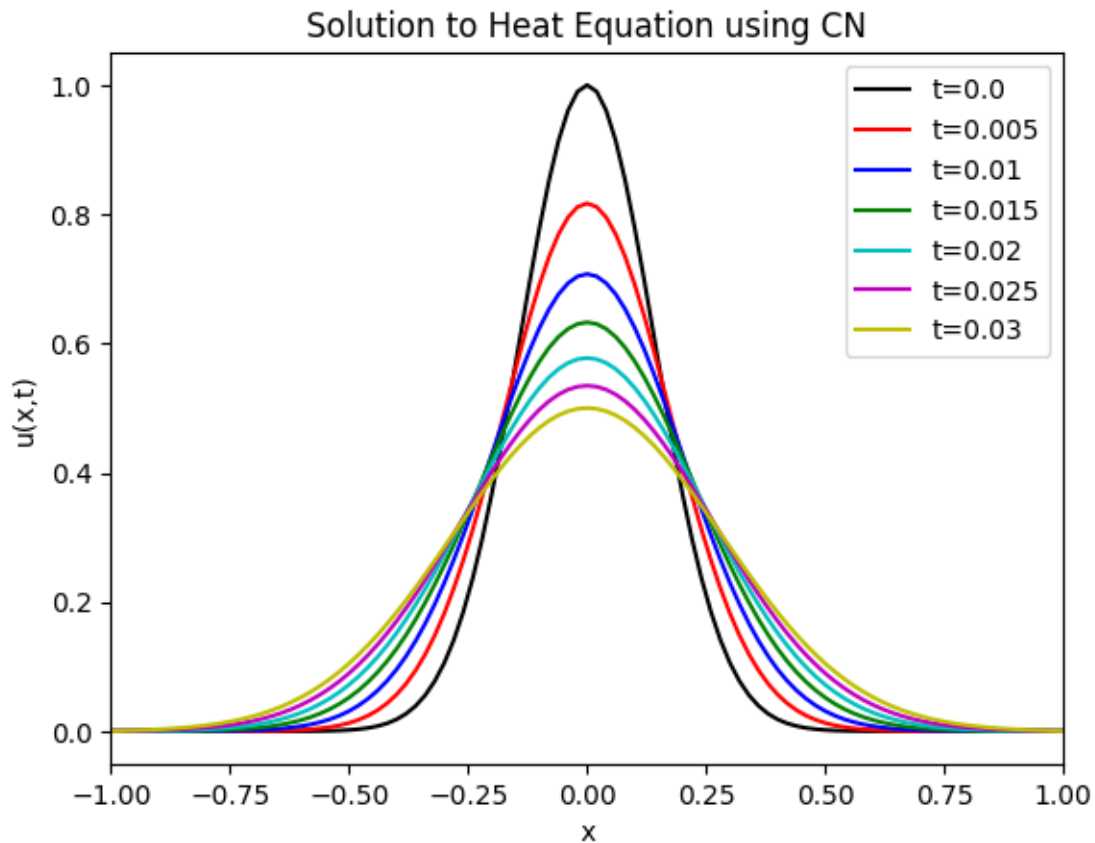
```

Heat Equation Solution (Crank-Nicolson Method)



```
[32]: x, t, U = solve_crank_nicolson(m=101, T_final=T_final)

# Plot a few solutions
colors = ['k', 'r', 'b', 'g', 'c', 'm', 'y']
fig = plt.figure()
axes = fig.add_subplot(1, 1, 1)
for (i, n) in enumerate((0, 25, 50, 75, 100, 125, 150)):
    axes.plot(x, U[n, :], colors[i], label='t=%s' % np.round(t[n], 4))
    axes.set_xlabel("x")
    axes.set_ylabel("u(x,t)")
    axes.set_title("Solution to Heat Equation using CN")
    axes.set_xlim([-1,1])
axes.legend()
plt.show()
```



```
[33]: from matplotlib import cm

# Create meshgrid for plotting
X, T = np.meshgrid(x, t)

# Create figure and 3D axis
fig = plt.figure(figsize=(8, 6))
ax = fig.add_subplot(111, projection='3d')

# Plot the surface
surf = ax.plot_surface(X, T, U, cmap=cm.viridis, edgecolor='none')

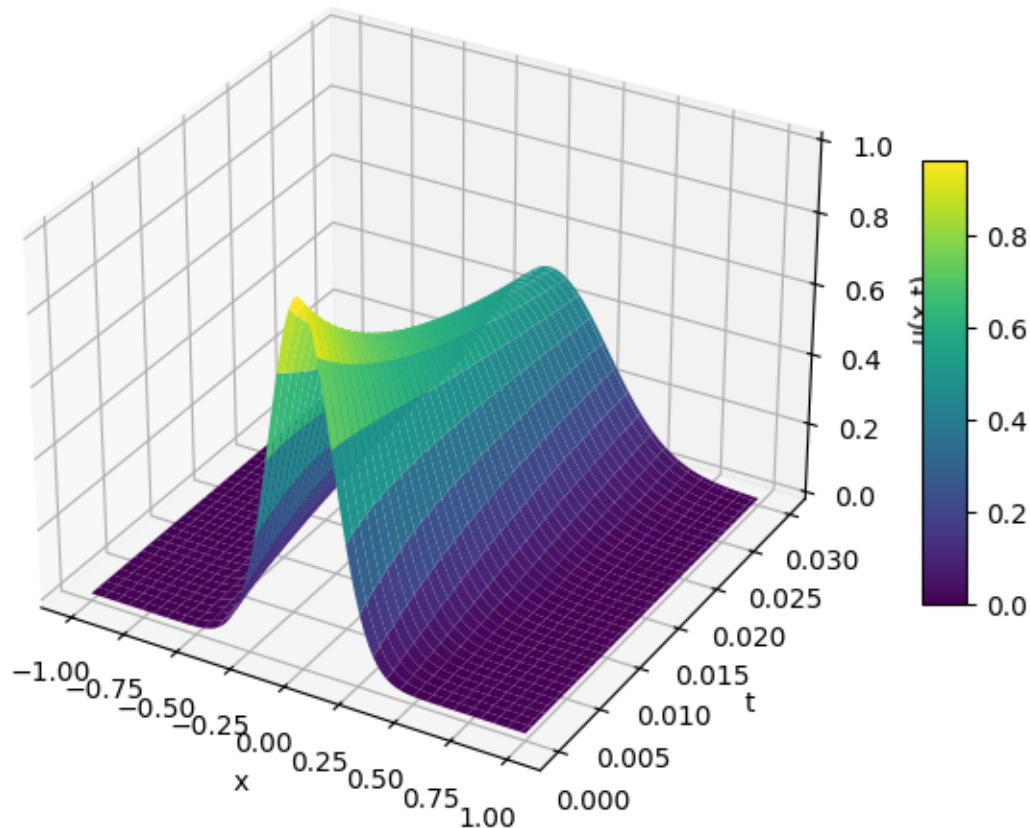
# Labels and title
ax.set_xlabel('x')
ax.set_ylabel('t')
ax.set_zlabel('u(x,t)')
ax.set_title('Heat Equation Solution (Crank-Nicolson Method)')

# Optional color bar
```

```
fig.colorbar(surf, shrink=0.5, aspect=10)

plt.show()
```

Heat Equation Solution (Crank-Nicolson Method)



b) Perform a **convergence analysis in time and space** for $T = 0.03$:

- **Convergence:** Fix $k = 0.5h^2$ for both methods. Compute the numerical solution for successively refined spatial grids (e.g., $m = [11, 21, 41, 81, 161, 321]$). Compare each result to a reference “exact” or highly resolved numerical solution ($m=641$) to compute the **spatial error**, typically using the L_2 or L_∞ norm. Estimate the **order of accuracy in space** by fitting the slope of the log-log plot of error vs. h .

```
[34]: import numpy as np

# Calculate the fine-grid solution once
mfine = 641
x_fine, _, U_fine = solve_explicit(m=mfine, T_final=T_final)
U_fine_final = U_fine[-1, :]
```



```

errors = []
ms = [11, 21, 41, 81, 161, 321]

print(f"{'m':>5} {'error':>15} {'order':>10}")
print("-" * 32)

for m in ms:
    x, t, U = solve_explicit(m=m, T_final=T_final)

    # Step to subsample the fine solution
    step = int((mfine - 1) / (m - 1))

    # Compute max error at final time
    error = np.max(np.abs(U[-1, :] - U_fine_final[:, step]))
    errors.append(error)

    # Compute order if not first row
    if len(errors) == 1:
        order = "-"
    else:
        h1 = 1 / (ms[len(errors)-2] - 1)
        h2 = 1 / (ms[len(errors)-1] - 1)
        order = np.log(errors[-2] / errors[-1]) / np.log(h1 / h2)
        order = f"{order:.2f}"

    print(f"{'m':5d} {'error':15.6e} {'order':>10}")

```

| m | error | order |
|-----|--------------|-------|
| 11 | 1.134592e-01 | - |
| 21 | 3.928654e-02 | 1.53 |
| 41 | 2.939683e-03 | 3.74 |
| 81 | 7.223410e-04 | 2.02 |
| 161 | 6.588966e-04 | 0.13 |
| 321 | 1.563368e-04 | 2.08 |

```

[35]: import numpy as np

# Calculate the fine-grid solution once
mfine = 641
x_fine, _, U_fine = solve_crank_nicolson(m=mfine, T_final=T_final)
U_fine_final = U_fine[-1, :]

errors = []
ms = [11, 21, 41, 81, 161, 321]

```

```

print(f"{'m':>5} {'error':>15} {'order':>10}")
print("-" * 32)

for m in ms:
    x, t, U = solve_crank_nicolson(m=m, T_final=T_final)

    # Step to subsample the fine solution
    step = int((mfine - 1) / (m - 1))

    # Compute max error at final time
    error = np.max(np.abs(U[-1, :] - U_fine_final[:, :step]))
    errors.append(error)

    # Compute order if not first row
    if len(errors) == 1:
        order = "-"
    else:
        h1 = 1 / (ms[len(errors)-2] - 1)
        h2 = 1 / (ms[len(errors)-1] - 1)
        order = np.log(errors[-2] / errors[-1]) / np.log(h1 / h2)
        order = f"{order:.2f}"

    print(f"{m:5d} {error:15.6e} {order:>10}")

```

| m | error | order |
|-----|--------------|-------|
| 11 | 2.232954e-02 | - |
| 21 | 5.468667e-03 | 2.03 |
| 41 | 1.419306e-03 | 1.95 |
| 81 | 3.580007e-04 | 1.99 |
| 161 | 8.567692e-05 | 2.06 |
| 321 | 1.715704e-05 | 2.32 |

[]: