

Kazakh-British Technical University

Report

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Assignment #9

1) Boundary and Initial Conditions

No time dependency → No initial Conditions

Dirichlet Boundary Conditions

BC for OX: $T(0, y) = 100$ and $T(1, y) = 100$

BC for OY: $T(x, 0) = 100$ and $T(x, 1) = 100$

2) Computational Grid

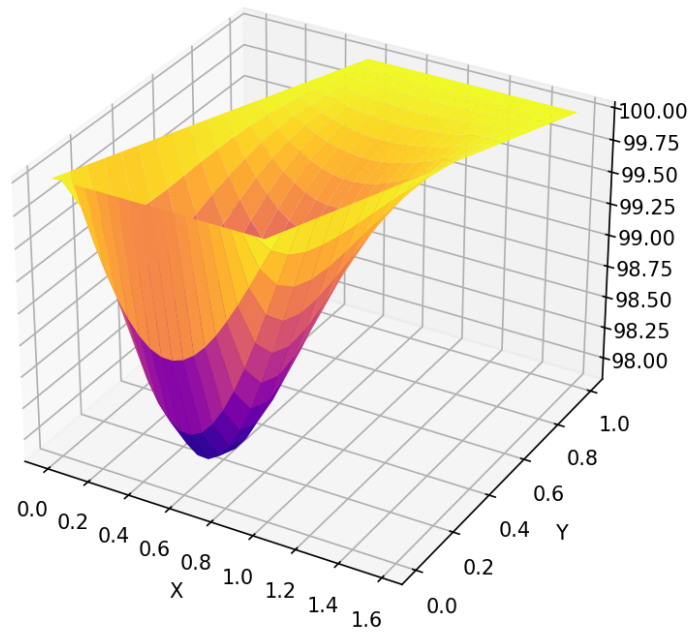
Grid Type: nodes are nonevenly spaced in a rectangular shapes in XOY (which is linearly transformed by 30 degrees to a parallelogram)

Spacings ($\Delta x, \Delta y$) are chosen randomly

3) FVM (Finite Volume Method):

Similar to the Jacobi method, but specifically incorporates volume integrals to approximate flux and source terms, ensuring conservation laws are satisfied within each control volume.

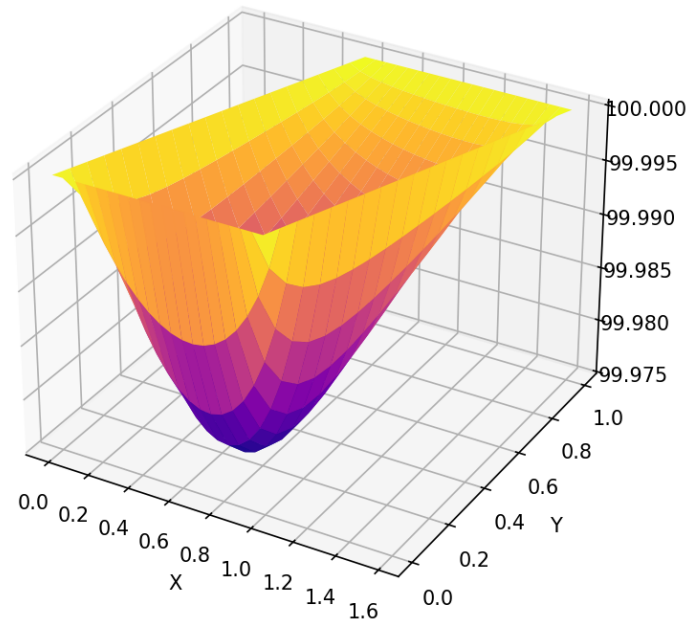
Finite Volume Method



4) Jacobi Method

$$T_{i,j}^{(k+1)} = \frac{S_{21}T_{i,j-1}^{(k)} + S_{31}T_{i-1,j}^{(k)} + S_{41}T_{i,j+1}^{(k)} + S_{51}T_{i+1,j}^{(k)} - f(x_{i,j}, y_{i,j}) \cdot \gamma}{S_{21} + S_{31} + S_{41} + S_{51}}$$

Jacobi Method



```
import matplotlib.pyplot as plt
import numpy as np
from mpl_toolkits.mplot3d import Axes3D

def f(x, y):
    return np.sin(x)*np.cos(y)

def volume(neighbour1, neighbour2):
    return (neighbour1+neighbour1)*(neighbour2+neighbour2)/10

def Sij(neighbour1, neighbour2, neighbour3, neighbour4, neigh
    S_31 = 1/neighbour5*(neighbour1+neighbour3)/2
    s_21 = 1/(2*neighbour1)*(neighbour2+neighbour4)/2
    S_41 = 1/(2*neighbour3)*(neighbour2+neighbour4)/2
    S_51 = 1/neighbour2*(neighbour1+neighbour3)/2
    return s_21, S_31, S_41, S_51

def FVM(X, Y, N, T, max_step=1000):
    for it in range(max_step):
        for i in range(1, N[1] - 1):
```

```

        for j in range(1, N[0] - 1):
            neighbour1 = X[i,j]-X[i,j-1]
            neighbour2 = Y[i,j]-Y[i-1,j]
            neighbour3 = X[i,j+1]-X[i,j]
            neighbour4 = Y[i+1,j]-Y[i,j]
            neighbour5 = X[i+1,j]-X[i,j-1]

            gamma = volume(neighbour1, neighbour2)
            S21, S31, S41, S51 = Sij(neighbour1, neighbour2, neighbour3, neighbour4, neighbour5)

            f_xy = f(X[i, j], Y[i, j])
            T[i,j]=(S21*T[i,j-1]+S31*T[i-1,j]+S41*T[i,j+1]+S51*T[i+1,j]+f_xy*gamma)/4

    return T

def Jacobi(X, Y, N, T, max_step=1000):
    for it in range(max_step):
        for i in range(1, N[1] - 1):
            for j in range(1, N[0] - 1):
                dx_2 = X[i,j]-X[i,j-1]
                dx_4 = X[i,j+1]-X[i,j]
                dy_5 = Y[i,j]-Y[i-1,j]
                dy_3 = Y[i+1,j]-Y[i,j]
                dx_3 = X[i+1,j]-X[i,j-1]

                S21, S31, S41, S51 = Sij(dx_2, dx_4, dy_5, dy_3, dx_3)
                gamma = volume(dx_2, dy_5)

                f_xy = f(X[i, j], Y[i, j])

                T[i, j] = (S21*T[i,j-1]+S31*T[i-1,j]+S41*T[i,j+1]+S51*T[i+1,j]+f_xy*gamma)/4

    return T

N = [20, 20]
L = [1, 1]
initial_temp = 100

```

```

angle = 30
inrad = np.radians(angle)

x_discrete = np.linspace(0, L[0], N[0])
y_discrete = np.linspace(0, L[1], N[1])
X, Y = np.meshgrid(x_discrete, y_discrete)
X += Y * np.tan(inrad)

T = np.zeros((N[0], N[1]))
T[:, 0] = initial_temp
T[:, -1] = initial_temp
T[0, :] = initial_temp
T[-1, :] = initial_temp

fvm_matrix = FVM(X, Y, N, T)
# plot FVM
fig = plt.figure(figsize=(10, 6))
ax = fig.add_subplot(111, projection='3d')
surf = ax.plot_surface(X, Y, fvm_matrix, cmap='plasma', edgecolor='k')
plt.colorbar(surf, label='Temperature')
ax.set_xlabel('X')
ax.set_ylabel('Y')
ax.set_title('Finite Volume Method')
plt.show()

jacobi_matrix = Jacobi(X, Y, N, T)
# plot Jacobi
fig = plt.figure(figsize=(10, 6))
ax = fig.add_subplot(111, projection='3d')
surf = ax.plot_surface(X, Y, jacobi_matrix, cmap='plasma', edgecolor='k')
plt.colorbar(surf, label='Temperature')
ax.set_xlabel('X')
ax.set_ylabel('Y')
ax.set_title('Jacobi Method')
plt.show()

```

Conclusion:

- **Jacobi Method:**

- Simple and easy to implement for **linear** systems but might struggle with complex problems.
- More useful for **steady-state solutions** and **linear equations**.

- **Finite Volume Method:**

- More complex but suitable for **PDEs**, especially with **conservation laws**, and for systems with **non-linear behavior** or **complicated boundaries**.
- Likely to be more accurate in your case since it's specifically designed to handle physical conservation principles.