

# 作业 2: 数值求解泊松方程

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## 1 数值求解泊松方程

### 1.1 理论推导

现求解二维泊松方程:

$$\begin{cases} \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = f(x, y), & (x, y) \in \Omega \\ u = u_0 & \text{on } \partial\Omega \end{cases} \quad (1)$$

其中区域  $\Omega$  为矩形区域, 即  $\Omega = \{(x, y) : 0 < x < 1, 0 < y < 1\}$ . 首先对该矩形区域进行划分, 划分成若干个小矩形, 取  $x_i = i\Delta x$ ,  $i = 0, 1, \dots, N+1$ ,  $\Delta x = \frac{1}{N+1}$ ,  $y_j = j\Delta y$ ,  $j = 0, 1, \dots, M+1$ ,  $\Delta y = \frac{1}{M+1}$ . 由有限差分方法可推得:

$$\frac{1}{\Delta x^2}(u_{i-1,j} - 2u_{i,j} + u_{i+1,j}) + \frac{1}{\Delta y^2}(u_{i,j-1} - 2u_{i,j} + u_{i,j+1}) = f_{i,j}$$

$$\frac{\Delta y^2}{\Delta x^2}(u_{i-1,j} - 2u_{i,j} + u_{i+1,j}) + (u_{i,j-1} - 2u_{i,j} + u_{i,j+1}) = \Delta y^2 f_{i,j}$$

令  $\lambda = \frac{\Delta y}{\Delta x}$ , 化简可得:

$$\lambda^2 u_{i-1,j} + \lambda^2 u_{i+1,j} + u_{i,j-1} + u_{i,j+1} - 2(1 + \lambda^2)u_{i,j} = \Delta y^2 f_{i,j} \quad (2)$$

(2) 式中有  $M \times N$  个未知的  $u_{ij}$ , 将之写成矩阵形式  $AU = F$ , 其中

$$U = [u_{1,1}, \dots, u_{N,1}, u_{1,2}, \dots, u_{N,2}, \dots, u_{1,M}, \dots, u_{N,M}]^T.$$

$$A = \begin{bmatrix} T & D & & & \\ D & T & D & & \\ & & \ddots & \ddots & \ddots \\ & & & D & T & D \\ & & & & D & T \end{bmatrix},$$

$$T = \begin{bmatrix} -2(1+\lambda^2) & \lambda^2 & & & \\ \lambda^2 & -2(1+\lambda^2) & \lambda^2 & & \\ & & \ddots & \ddots & \ddots \\ & & & \lambda^2 & -2(1+\lambda^2) & \lambda^2 \\ & & & & \lambda^2 & -2(1+\lambda^2) \end{bmatrix}.$$

## 1.2 实例

求解如下方程：

$$\begin{cases} \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = 5e^{x+2y}, & 0 < x < 1, 0 < y < 1 \\ u = e^{2y} & x = 0, 0 < y < 1 \\ u = e^x & y = 0, 0 < x < 1 \\ u = e^{1+2y} & x = 1, 0 < y < 1 \\ u = e^{x+2} & y = 1, 0 < x < 1 \end{cases} \quad (3)$$

其精确解为  $u(x, y) = e^{x+2y}$ 。

## 1.3 编程实现

根据上述推导，用 python 编写程序，代码如下：

---

```

1  import numpy as np
2  import matplotlib.pyplot as plt
3  import math
4
5  # 生成矩阵T、D，为生成矩阵A做准备
6  def generate_TD(N, dx, dy):
7      T = np.zeros([N,N])
8      D = np.zeros([N,N])
9      a = (dy/dx)**2

```

```
10     for i in range(N):
11         T[i,i] = -2*(1+a)
12         D[i,i] = 1
13         if (i < N-1):
14             T[i,i+1] = a
15         if (i > 0):
16             T[i,i-1] = a
17     return T, D
18
19 # 生成矩阵A
20 def assemble_A(N, M, dx, dy):
21     T, D = generate_TD(N, dx, dy)
22     A = np.zeros([N*M, N*M])
23     for j in range(M):
24         A[j*N:(j+1)*N, j*N:(j+1)*N] = T
25         if (j < M-1):
26             A[j*N:(j+1)*N, (j+1)*N:(j+2)*N] = D
27         if (j > 0):
28             A[j*N:(j+1)*N, (j-1)*N:(j)*N] = D
29     return A
30
31
32 def f(x, y):
33     return 5 * math.exp(x + 2 * y)
34
35 # 精确解
36 def exact_f(x, y):
37     return math.exp(x + 2 * y)
38
39 def gL(y):
40     return math.exp(2 * y)
41
42 def gR(y):
43     return math.exp(1 + 2 * y)
44
45 def gB(x):
46     return math.exp(x)
47
48 def gT(x):
```

```

49         return math.exp(x + 2)
50
51     def assemble_F(x, y, dx, dy, N, M, gL, gR, gB, gT):
52         F = np.zeros(N*M)
53
54         a = (dy/dx)**2
55
56         # dy^2 * f(i,j)
57         for j in range(M):
58             for i in range(N):
59                 F[j * N + i] += ((dy) ** 2) * f(x[i + 1], y[j + 1])
60
61         # left BCs
62         for j in range(M):
63             F[j*N] += -a*gL(y[j+1])
64
65         # right BCs
66         for j in range(M):
67             F[(j+1)*N - 1] += -a*gR(y[j+1])
68
69         # top BCs
70         for i in range(N):
71             F[N * (M - 1) + i] += -gT(x[i+1])
72
73         # bottom BCs
74         for i in range(N):
75             F[i] += -gB(x[i + 1])
76
77         return F
78
79     def exact_solution(N, M, x, y):
80         U_exact = np.zeros(N * M)
81         for j in range(M):
82             for i in range(N):
83                 U_exact[j * N + i] = exact_f(x[i + 1], y[j + 1])
84         return U_exact
85
86     def Poission_solver(N, M, gL, gR, gB, gT):
87         dx = 1./(N+1)

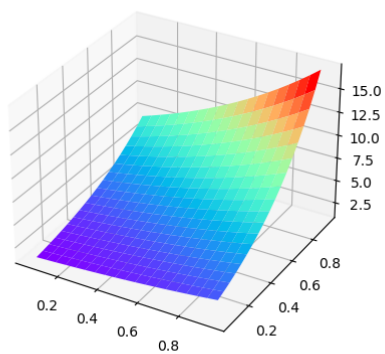
```

```
88     dy = 1./(M+1)
89     x = np.linspace(0, 1, N+2)
90     y = np.linspace(0, 1, M+2)
91
92     A = assemble_A(N, M, dx, dy)
93
94     F = assemble_F(x, y, dx, dy, N, M, gL, gR, gB, gT)
95
96     U = np.linalg.solve(A, F)
97     U_exact = exact_solution(N, M, x, y)
98     error = max(abs(U-U_exact))
99
100    u = np.reshape(U, (N,M))
101    u_exact = np.reshape(U_exact, (N,M))
102
103
104    X, Y = np.meshgrid(x[1:N+1], y[1:M+1])
105    fig = plt.figure()
106    ax = fig.add_subplot(1, 1, 1, projection='3d')
107
108    ax.plot_surface(X, Y, u, cmap='rainbow')
109
110    # print (u)
111    print(error)
112    plt.show()
113
114    Possion_solver(19, 19, gL, gR, gB, gT)
```

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## 1.4 结果分析

当取  $h = 0.05$  时, 此时误差为 0.0013997692884775148, 结果如下图所示:

图 1:  $h = 0.05$  结果图

当取不同  $h$ ，得到的误差如下表所示：

表 1: 不同  $h$  的误差表

$h$	误差
$\frac{1}{10}$	0.005519939625335368
$\frac{1}{20}$	0.0013997692884775148
$\frac{1}{40}$	0.0003511099859592193
$\frac{1}{80}$	8.787626974093854e-05