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Project #1: Solving Linear Systems

Advanced Numerical Analysis

By

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1 Problem Statement

Consider a steady-state heat diffusion inside a 2D rectangular medium with the prescribed boundary conditions in Fig. 1(a). The governing equation is as follows:

$$\frac{\partial^2 T}{\partial x^2} + \frac{\partial^2 T}{\partial y^2} = 0$$

The temperature distribution for this problem is also depicted in Fig. 1(b). Equation (1) can be discretized using a central-difference scheme inside the domain. Considering a domain size of $(L_x, L_y) = (1.0, 1.0)$ and grid dimension of (n_x, n_y) with a uniform grid size of $(\Delta x, \Delta y)$ for each direction i and j , respectively. The algebraic equation for $1 \leq i \leq n_x - 1$ and $1 \leq j \leq n_y - 1$ becomes:

$$a_{i-1,j}T_{i-1,j} + a_{i,j-1}T_{i,j-1} + a_{i,j}T_{i,j} + a_{i+1,j}T_{i+1,j} + a_{i,j+1}T_{i,j+1} = b_{i,j}$$

where the coefficients a 's and b 's are as follows:

$$a_{i-1,j} = a_{i+1,j} = 1.0, \quad a_{i,j-1} = a_{i,j+1} = \left(\frac{\Delta x}{\Delta y}\right)^2$$

$$a_{i,j} = -2.0 \left[1.0 + \left(\frac{\Delta x}{\Delta y}\right)^2 \right], \quad b_{i,j} = 0$$

Note that in order to reduce the round-off-error, both sides of Eq. (2) was multiplied by $(\Delta x)^2$. To implement Dirichlet and Neumann conditions at boundary nodes, the following relations are used:

$$\begin{aligned} \text{at left side (i=0)} & : T_{i,j} = 0.0 \rightarrow a_{i,j} = 1.0, b_{i,j} = 0.0 \\ \text{at lower side (j=0)} & : T_{i,j} = 1.0 \rightarrow a_{i,j} = 1.0, b_{i,j} = 1.0 \\ \text{at upper side (j=ny)} & : T_{i,j} = 1.0 \rightarrow a_{i,j} = 1.0, b_{i,j} = 1.0 \\ \text{at right side (i=nx)} & : T_{i,j} = T_{i-1,j} \rightarrow a_{i-1,j} = -a_{i,j} = 1.0, b_{i,j} = 0.0 \end{aligned}$$

Accordingly, the linear system of equation with size of $(N; N)$, with $N = n_x \times n_y$, can be formed as $AX = B$ where $A = [A_{m,n}]$ and $B = [B_m]$ ($0 \leq m, n \leq N - 1$) and $X = [X_m]$ is the solution.

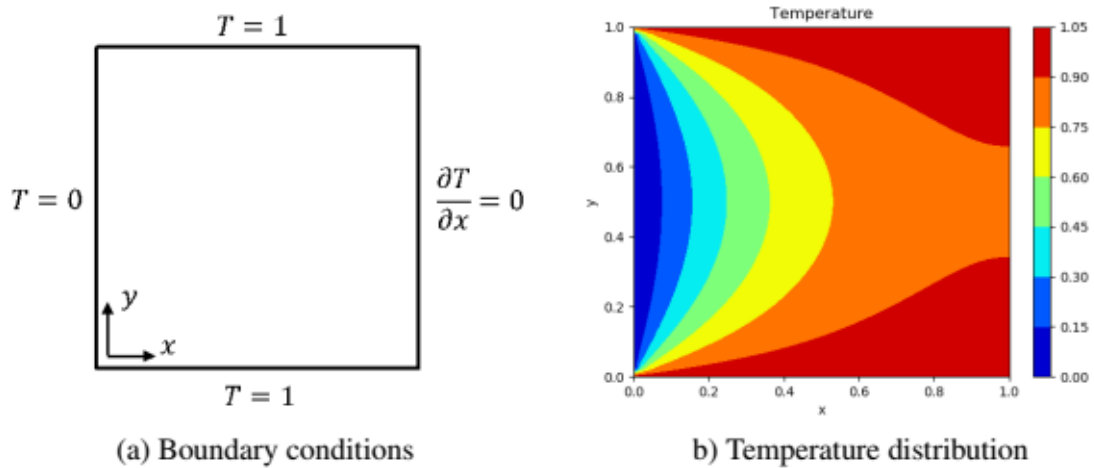


Figure 1: The problem sketch for 2D heat diffusion

2 Methodology

The system of equations described in Sec.1 is going to be solved using sparse matrix format and the uniform mesh with $n_x = n_y = \{10, 20, 50, 100\}$ and the residual tolerance of 10^{-5} and $X_0 = 0$. The code has written in Spyder (Python 3.6).

2.1 Constructing A and B for linear solver

In first step we define matrix A in dense form, then we convert it to sparse format using scipy.sparse module's coo_matrix attribute

```
1 # initialization
2 import numpy as np
3 import scipy.sparse as sp
4 nx = eval(input('x,y grid number: '))
5 ny = nx
6 dx = 1/nx
7 dy = 1/ny
8 N = (nx+1)*(ny+1)
9 x = np.linspace(0, 1, nx+1)
10 y = np.linspace(0, 1, ny+1)
11
12 # constructing A and b (coefficient matrix and right hand side vector)
13 A = np.zeros([N,N])
14 b = np.zeros(N)
```

```

15 for i in range(N):
16     if i==0:
17         A[i,i]=-4
18         A[i,i+1]=1
19         A[i,i+nx+1]=1
20         b[i]=-1
21     elif i==nx:
22         A[i,i]=-3
23         A[i,i-1]=1
24         A[i,2*i+1]=1
25         b[i]=-1
26     elif i==N-nx-1:
27         A[i,i]=-4
28         A[i,i-nx-1]=1
29         A[i,i+1]=1
30         b[i]=-1
31     elif i==N-1:
32         A[i,i]=-3
33         A[i,i-1]=1
34         A[i,i-nx-1]=1
35         b[i]=-1
36     elif i<nx+1:
37         A[i,i]=-4
38         A[i,i-1]=1
39         A[i,i+1]=1
40         A[i,i+nx+1]=1
41         b[i]=-1
42     elif i>N-nx-1:
43         A[i,i]=-4
44         A[i,i-1]=1
45         A[i,i+1]=1
46         A[i,i-nx-1]=1
47         b[i]=-1
48     elif i%(nx+1)==0:
49         A[i,i]=-4
50         A[i,i-nx-1]=1
51         A[i,i+nx+1]=1
52         A[i,i+1]=1
53     elif (i+1)%(nx+1)==0:
54         A[i,i]=-3
55         A[i,i-1]=1
56         A[i,i-nx-1]=1
57         A[i,i+nx+1]=1
58     else:
59         A[i,i]=-4
60         A[i,i-1]=1
61         A[i,i+1]=1
62         A[i,i-nx-1]=1
63         A[i,i+nx+1]=1
64
65 Asp=sp.coo_matrix(A)

```

2.2 sketching sparsity matrix for $nx = 6$

```

67 # Sparsity pattern for nx = 6
68 import matplotlib.pyplot as plt
69 print(plt.spy(Asp))

```

2.4 Investigate the convergence of SciPy Gmres solver

2.4.1 Without preconditioning:

```
71 # GMRES (without preconditioning)
72 import scipy.sparse.linalg as spla
73 def counter(rk=None):
74     counter.niter+=1
75     print("# iter {:3d}, residual = {}".format(counter.niter, str(rk)))
76 x0=np.zeros(N)
77 counter.niter=0
78 T = spla.gmres(A,b,x0=x0,callback=counter)
79 print(T[0])
```

2.4.2 With preconditioning:

```
71 # GMRES (with preconditioning)
72 import scipy.sparse.linalg as spla
73 def counter(rk=None):
74     counter.niter+=1
75     print("# iter {:3d}, residual = {}".format(counter.niter, str(rk)))
76 x0=np.zeros(N)
77 x = spla.gmres(A,b,x0=x0,callback=counter)
78 M2 = sp.linalg.spilu(Asp,fill_factor=1.0)
79 Mx = lambda x: M2.solve(x)
80 M = spla.LinearOperator((N,N),Mx)
81 counter.niter=0
82 T = spla.gmres(A,b,M=M,callback=counter,x0=x0)
83 print(T[0])
```

2.5 Plotting 2D temperature field

```
81 # Plotting 2D temperature field
82 T2 = T[0].reshape(ny+1,nx+1)
83 x,y = np.meshgrid(x,y)
84 plt.figure(figsize=(8,6))
85 plt.contourf(x,y,T2,10)
86 plt.set_cmap('jet')
87 plt.colorbar()
88 plt.ylabel('y')
89 plt.xlabel('x')
90 plt.show()
```

3 Results and Discussions

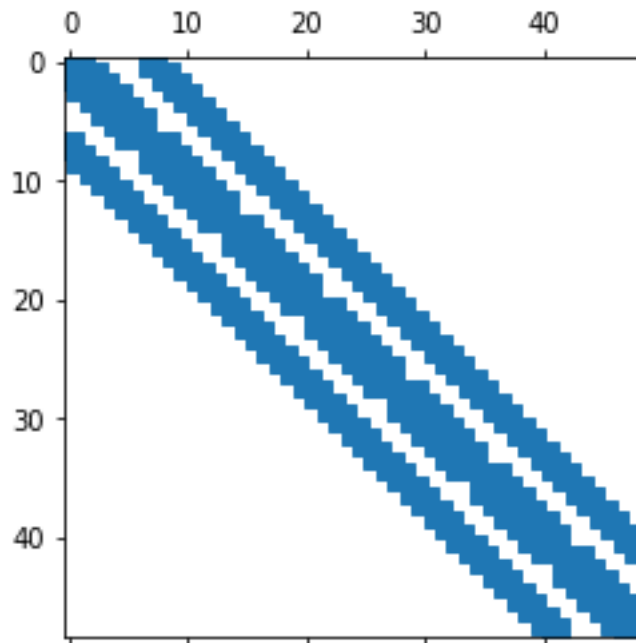
3.1 Matrix A and b

For $n_x = 3$:

```
In [12]: print(A)
[[-4.  1.  0.  0.  1.  0.  0.  0.  0.  0.  0.  0.  0.  0.  0.  0.]
 [ 1. -4.  1.  0.  0.  1.  0.  0.  0.  0.  0.  0.  0.  0.  0.  0.]
 [ 0.  1. -4.  1.  0.  0.  1.  0.  0.  0.  0.  0.  0.  0.  0.  0.]
 [ 0.  0.  1. -3.  0.  0.  0.  1.  0.  0.  0.  0.  0.  0.  0.  0.]
 [ 1.  0.  0.  0. -4.  1.  0.  0.  1.  0.  0.  0.  0.  0.  0.  0.]
 [ 0.  1.  0.  0.  1. -4.  1.  0.  0.  1.  0.  0.  0.  0.  0.  0.]
 [ 0.  0.  1.  0.  0.  1. -4.  1.  0.  0.  1.  0.  0.  0.  0.  0.]
 [ 0.  0.  0.  1.  0.  0.  1. -3.  0.  0.  0.  1.  0.  0.  0.  0.]
 [ 0.  0.  0.  0.  1.  0.  0.  0. -4.  1.  0.  0.  1.  0.  0.  0.]
 [ 0.  0.  0.  0.  0.  1.  0.  0.  1. -4.  1.  0.  0.  1.  0.  0.]
 [ 0.  0.  0.  0.  0.  0.  1.  0.  0.  1. -4.  1.  0.  0.  1.  0.]
 [ 0.  0.  0.  0.  0.  0.  0.  1.  0.  0.  1. -3.  0.  0.  0.  1.]
 [ 0.  0.  0.  0.  0.  0.  0.  0.  1.  0.  0.  0. -4.  1.  0.  0.]
 [ 0.  0.  0.  0.  0.  0.  0.  0.  0.  1.  0.  0.  1. -4.  1.  0.]
 [ 0.  0.  0.  0.  0.  0.  0.  0.  0.  0.  1.  0.  0.  1. -4.  1.]
 [ 0.  0.  0.  0.  0.  0.  0.  0.  0.  0.  0.  1.  0.  0.  1. -3.]]

In [13]: print(b)
[-1. -1. -1. -1.  0.  0.  0.  0.  0.  0.  0.  0. -1. -1. -1. -1.]
```

3.2 Sketching sparsity pattern for $n_x = 6$:



3.4 Investigation of the convergence of SciPy Gmres

Without preconditioning:

`nx=10`

`# iter 39, residual = 7.4786858594249985e-06`

`nx=20`

`# iter 108, residual = 9.704270569453499e-06`

`nx=50`

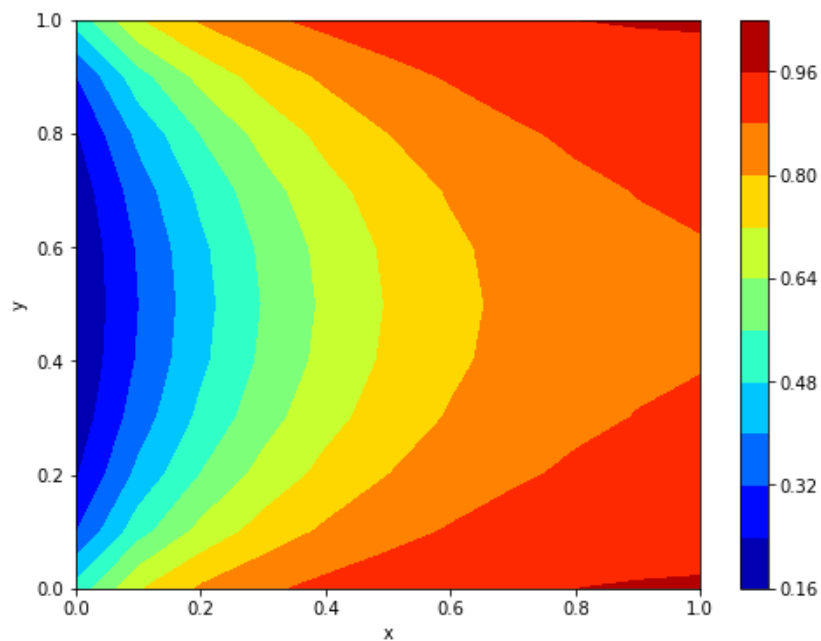
`# iter 304, residual = 9.839744279751739e-06`

`nx=100`

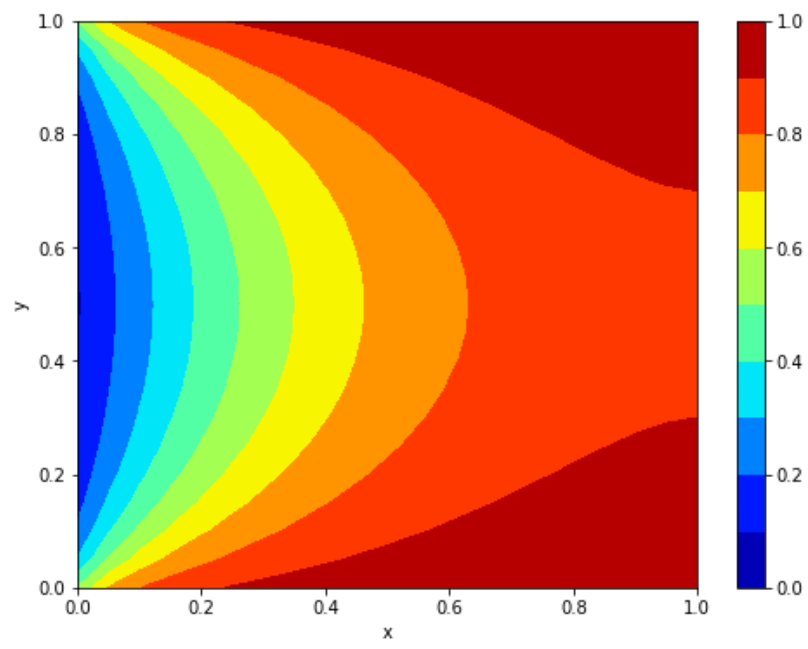
`# iter 1111, residual = 9.939962381073695e-06`

3.6 Plotting 2D temperature field

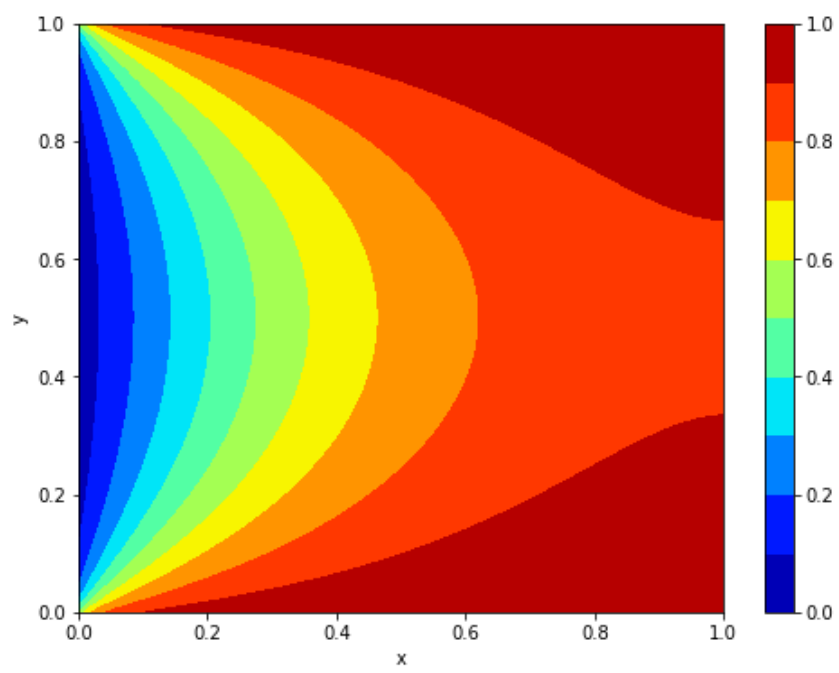
`nx=10`



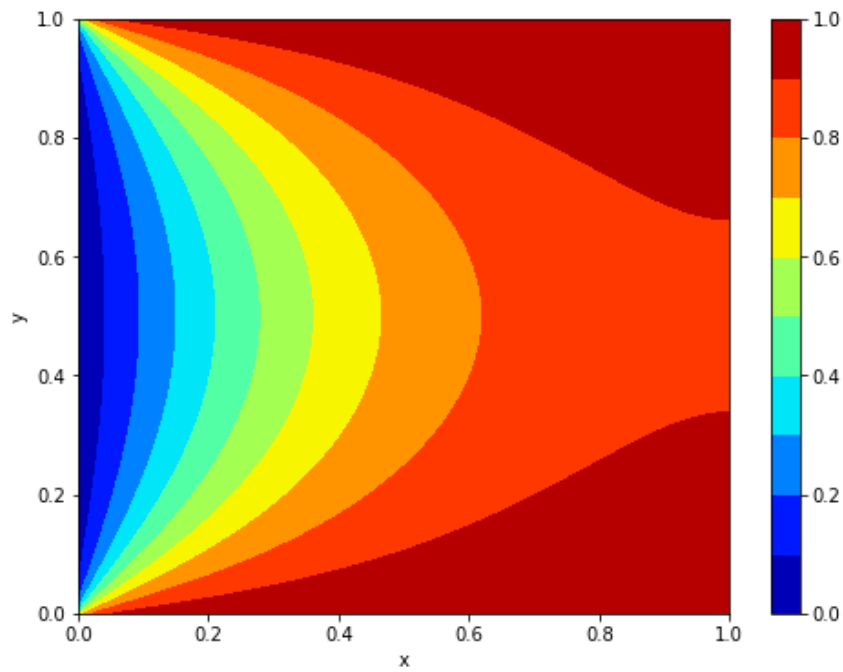
$nx=20$



$nx=50$



$nx=100$



By increasing the number of grid points, the temperature field's contour lines become smoother, temperatures of boundary nodes become closer to prescribed values in problem so solution become more accurate.