Basics of Applied Analysis A Report

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Problem 1. Let $N \in \mathbb{N}, N \geq 2$, and $\alpha > 0$ be given.

We define A = (aij) as finite difference discretization of:

$$\begin{cases} \alpha u - u'' = f \ in \ (0, 1) \\ u(0) = u(1) = 0 \end{cases} \tag{1}$$

with entries

$$a_{ij} = \begin{cases} \alpha + \frac{2}{h^2}, & i = j \\ -\frac{1}{h^2}, & |i - j| = 1 \\ 0 & otherwise \end{cases}$$
 (2)

where $h = \frac{1}{N}$

(a) To find eigenvalues λ s.t. $Av = \lambda v$, We define $v_i := \sin\left(\frac{m\pi i}{N}\right)$ for some $m \in \mathbb{N}$, hence the discrete form of eigen problem becomes:

$$\sum_{i=1}^{N} a_{ij} v_j = \lambda v_i$$

$$\left(\alpha + \frac{2}{h^2}\right) v_i - \frac{1}{h^2} v_{i+1} - \frac{1}{h^2} v_{i-1} = \lambda v_i$$

$$\left(\alpha + \frac{2}{h^2} - \lambda\right) v_i - \frac{1}{h^2} (v_{i+1} + v_{i-1}) = 0$$

substitute v_i to the discrete eigen problem, it becomes

$$\left(\alpha + \frac{2}{h^2} - \lambda\right) \sin\left(\frac{m\pi i}{N}\right) - \frac{1}{h^2} \left(\sin\left(\frac{m\pi(i+1)}{N}\right) + \sin\left(\frac{m\pi(i-1)}{N}\right)\right) = 0$$

$$\left(\alpha + \frac{2}{h^2} - \lambda\right) \sin\left(\frac{m\pi i}{N}\right) - \frac{1}{h^2} \left(2\sin\left(\frac{m\pi i}{N}\right)\cos\left(\frac{m\pi}{N}\right)\right) = 0$$

$$\sin\frac{m\pi i}{N} \left(\alpha + \frac{2}{h^2} - \lambda - \frac{1}{h^2}2\cos\left(\frac{m\pi}{N}\right)\right) = 0$$

to find λ , we set $\sin \frac{m\pi i}{N}$ equal to some constant C, so that:

$$\left(\alpha + \frac{2}{h^2} - \lambda - \frac{2}{h^2} \cos\left(\frac{m\pi}{N}\right)\right) = 0$$

$$\therefore \lambda = \alpha + \frac{2}{h^2} \left(1 - \cos\left(\frac{m\pi}{N}\right)\right) \tag{3}$$

With Eigenvectors:

$$\therefore v = \sin\left(\frac{m\pi i}{N}\right)$$

(b) We define spectral radius $\sigma(A) := \max\{|\lambda|\}$, using λ computed in (3), we get:

$$\sigma(A) = \max\{|\lambda|\}$$
$$= \alpha + \frac{2}{h^2} \left(1 - \cos\left(\frac{m\pi}{N}\right)\right)$$

then, thanks to the symmetricity of λ , we can use taylor expansion to approximate the value of

$$\cos\left(\frac{m\pi}{N}\right) \approx 1 - \frac{1}{2}\left(\frac{\pi}{N}\right)^2$$

, hence the spectral radius $\sigma(A)$ can be calculated by:

$$\sigma(A) \approx \alpha + \frac{2}{h^2} \left(1 - \left(1 - \frac{m^2 \pi^2}{2N^2} \right) \right)$$
$$\approx \alpha + \frac{2}{h^2} \frac{m^2 \pi^2}{2N^2}$$
$$\therefore \sigma(A) \approx \alpha + (m\pi)^2$$

(c) Let:

$$R := -D^{-1}(L + U)$$

be the Jacobi iteration matrix, we want to find the eigenvalues and eigenvectors s.t. $Rv = \lambda v$.

$$Rv = \lambda v$$
$$-D^{-1}(L+U)v = \lambda v$$
$$-(L+U)v = \lambda Dv$$

the discrete form of the eigen problem is:

$$\frac{1}{h^2}(v_{i-1} + v_{i+1}) = \lambda \left(\alpha + \frac{2}{h^2}\right) v_i$$

and then we set, $v_i := \sin\left(\frac{m\pi i}{N}\right)$ and substitute to the discrete form of the eigen problem, thus:

$$\frac{1}{h^2} \left(\sin\left(\frac{m\pi(i+1)}{N}\right) + \sin\left(\frac{m\pi(i-1)}{N}\right) \right) = \lambda \left(\alpha + \frac{2}{h^2}\right) \sin\left(\frac{m\pi i}{N}\right)$$

$$\frac{1}{h^2} \left(2\sin\left(\frac{m\pi i}{N}\right) \cos\left(\frac{m\pi}{N}\right) \right) = \lambda \left(\alpha + \frac{2}{h^2}\right) \sin\left(\frac{m\pi i}{N}\right)$$

$$\sin\frac{m\pi i}{N} \left(\frac{2}{h^2} \cos\left(\frac{m\pi}{N}\right) - \lambda \left(\alpha + \frac{2}{h^2}\right) \right) = 0$$

to find λ , we set $\sin \frac{m\pi i}{N}$ equal to some constant D, so that:

$$\left(\frac{2}{h^2}\cos\left(\frac{m\pi}{N}\right) - \lambda\left(\alpha + \frac{2}{h^2}\right)\right) = 0$$

$$\lambda\left(\alpha + \frac{2}{h^2}\right) = \frac{2}{h^2}\cos\left(\frac{m\pi}{N}\right)$$

$$with \ h = \frac{1}{N}$$

$$\lambda = \frac{\frac{2}{h^2}\cos\left(\frac{m\pi}{N}\right)}{\alpha + \frac{2}{h^2}}$$
(4)

with
$$h = \frac{1}{N}$$
,

$$\therefore \lambda = \frac{2N^2 \cos\left(\frac{m\pi}{N}\right)}{\alpha + 2N^2}$$

(d) We define spectral radius $\sigma(R) := \max\{|\lambda|\}$, using λ computed in (4), we get:

$$\begin{split} \sigma(R) &= \max\{|\lambda|\} \\ &= \frac{2N^2 \cos\left(\frac{m\pi}{N}\right)}{\alpha + 2N^2} \end{split}$$

again, thanks to the symmetricity of λ , we can use taylor expansion to approximate the value of

$$\cos\left(\frac{m\pi}{N}\right) \approx 1 - \frac{1}{2}\left(\frac{m\pi}{N}\right)^2$$

, hence the spectral radius $\sigma(R)$ can be calculated by:

$$\therefore \sigma(R) \approx \frac{2N^2 \left(1 - \frac{m^2 \pi^2}{2N^2}\right)}{\alpha + 2N^2}$$

Problem 2. Let $N, h = \frac{1}{N}, \alpha$ and A be as in Problem 1. We define:

$$\begin{cases} \alpha u - u'' = \sin(\pi x) \ in \ (0, 1) \\ u(0) = u(1) = 0 \end{cases}$$
 (5)

(a) To find the exact solution of u, we use general solution for Ordinary Differential Equation (ODE):

$$u(x) = A\sin(\pi x) + B\cos(\pi x) \tag{6}$$

and take the second derivative of u

$$u''(x) = -A\pi^2 \sin(\pi x) + B\pi^2 \cos(\pi x) \tag{7}$$

then substitute (6) and (7) to (5), we get:

$$\alpha(A\sin(\pi x) + B\cos(\pi x)) + A\pi^2\sin(\pi x) + B\pi^2\cos(\pi x) = \sin(\pi x)$$

$$A(\alpha + \pi^2)\sin(\pi x) + B(\alpha + \pi^2)\cos(\pi x) = \sin(\pi x)$$
(8)

from (8) we know that

$$A(\alpha + \pi^2)\sin(\pi x) = \sin(\pi x)$$

and

$$B(\alpha + \pi^2)\cos(\pi x) = 0$$

$$\therefore A = \frac{1}{(\alpha + \pi^2)} \quad \therefore B = 0 \tag{9}$$

then we substitute (9) to (6), we get the exact solution as:

$$u(x) = \frac{\sin(\pi x)}{(\alpha + \pi^2)} \tag{10}$$

(b) To find exact solution s.t. Av = b, with $b_i = \sin(\pi hi)$, we set

$$v_i = C\sin(\pi hi)$$

thus,

$$Av = b$$

$$\sum_{i=1}^{N-1} a_{ij}v_i = b_i$$

$$\left(\left(\alpha + \frac{2}{h^2}\right)v_i - \frac{1}{h^2}\left(v_{i+1} + v_{i-1}\right)\right) = \sin(\pi h i)$$

$$\left(\left(\alpha + \frac{2}{h^2}\right)C\sin(\pi h i) - \frac{1}{h^2}\left(C\sin(\pi h (i+1)) + C\sin(\pi h (i-1))\right)\right) = \sin(\pi h i)$$

$$\left(\left(\alpha + \frac{2}{h^2}\right)C\sin(\pi h i) - C\frac{1}{h^2}\left(2\sin(\pi h i)\cos(\pi h i)\right) = \sin(\pi h i)$$

divide by $\sin(\pi hi)$ on the both side we get:

$$C\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}\cos(\pi h)\right) = 1$$

$$\therefore C = \frac{1}{\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}\cos(\pi h)\right)}$$

so the exact solution in terms of v_i is:

$$v_i = \frac{\sin(\pi hi)}{\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}\cos(\pi h)\right)}$$

(c) We define

$$\epsilon(h) := \max_{1 \le i \le N-1} |u(hi) - v_i|$$

then we use u(x) and v_i that we derived before to find the explicit formula of $\epsilon(h)$:

$$\epsilon(h) := \max_{1 \le i \le N-1} \left| \frac{\sin(\pi hi)}{(\alpha + \pi^2)} - \frac{\sin(\pi hi)}{\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}\cos(\pi h)\right)} \right|$$

$$\epsilon(h) := \max_{1 \le i \le N-1} \left| \sin(\pi hi) \left(\frac{1}{(\alpha + \pi^2)} - \frac{1}{\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}\cos(\pi h)\right)} \right) \right|$$

for simplicity, we rewrite:

$$D := \sin(\pi h i)$$

$$E := \left(\frac{1}{(\alpha + \pi^2)} - \frac{1}{\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}\cos(\pi h)\right)}\right)$$

$$\max_{1 \le i \le N-1} |DE| = \max_{1 \le i \le N-1} |D| \max_{1 \le i \le N-1} |E|$$

$$= 1\left(\frac{1}{(\alpha + \pi^2)} - \frac{1}{\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}\cos(\pi h)\right)}\right)$$

then, the explicit formula for $\epsilon(h)$ is:

$$\epsilon(h) = \left(\frac{1}{(\alpha + \pi^2)} - \frac{1}{\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}\cos(\pi h)\right)}\right) \tag{11}$$

around h = 0, leading order of taylor expansion of $\epsilon(h)$ can be obtained by expand $\cos(\pi h)$, so that

$$\cos(\pi h) \approx \cos(a) + \frac{\cos'(a)(\pi h - a)}{1!} + \frac{\cos''(a)(\pi h - a)^2}{2!} + \cdots$$

then we take a = 0, we get

$$\cos(\pi h) \approx 1 - \frac{1}{2}(\pi h)^2$$

substitute it to (11) we get:

$$\epsilon(h) = \left(\frac{1}{(\alpha + \pi^2)} - \frac{1}{\left(\left(\alpha + \frac{2}{h^2}\right) - \frac{2}{h^2}(1 - \frac{1}{2}(\pi h)^2)\right)}\right)$$
$$\epsilon(h) = \left(\frac{1}{(\alpha + \pi^2)} - \frac{1}{(\alpha + \pi^2)}\right)$$

which is just a constant, so the leading order is depend on what order of taylor expansion we choose to approximate the value of $\cos(\pi h)$ around h = 0. In this case if we choose taylor expansion orde 2, we get the leading term of ϵ is $\mathbb{O}(F)$ with F is a constant.

Problem 3. We consider system of linear equations:

$$\begin{cases}
-v_{i-1,j} - v_{i+1,j} - v_{i,j-1} - v_{i,j+1} + 4v_{i,j} = b_{i,j} & i, j = 1, \dots, N-1 \\
v_{0,j} = v_{0,N} = v_{i,0} = v_{i,N} = 0 & i, j = 1, \dots, N-1
\end{cases}$$
(12)

for unknowns $v_{i,j}$, $i, j = 1, \dots, N-1$.

(a) We want to find the eigenvalues and eigenvectors of the matrix A for the system (12). Using the idea of discrete separation variables, we set $w_{i,j} = v_i \tilde{v_j}$ s.t.

$$Aw_{i,j} = \lambda w_{i,j}$$

$$4w_{i,j} - w_{i-1,j} - w_{i+1,j} - w_{i,j-1} - w_{i,j+1} = \lambda w_{i,j}$$

$$4(v_i\tilde{v_j}) - (v_{i-1}\tilde{v_j}) - (v_{i+1}\tilde{v_j}) - (v_i\tilde{v}_{j-1}) - (v_i\tilde{v}_{j+1}) = \lambda(v_i\tilde{v_j})$$

$$4(v_i\tilde{v_j}) - \tilde{v_j}(v_{i-1} + v_{i+1}) - v_i(\tilde{v}_{j-1} + \tilde{v}_{j+1}) = \lambda(v_i\tilde{v_j})$$

$$(4 - \lambda)(v_i\tilde{v_j}) - \tilde{v_j}(v_{i-1} + v_{i+1}) - v_i(\tilde{v}_{j-1} + \tilde{v}_{j+1}) = 0$$

then, divide by $v_i \tilde{v_i}$ on both side, we get:

$$(4 - \lambda) - \frac{(v_{i-1} + v_{i+1})}{v_i} - \frac{(\tilde{v}_{j-1} + \tilde{v}_{j+1})}{\tilde{v}_j} = 0$$
$$(4 - \lambda) - \frac{(v_{i-1} + v_{i+1})}{v_i} = \frac{(\tilde{v}_{j-1} + \tilde{v}_{j+1})}{\tilde{v}_j}$$

now, to satisfies equality, both side should be equal to some constant, let say G, hence:

$$(4 - \lambda) - \frac{(v_{i-1} + v_{i+1})}{v_i} = G = \frac{(\tilde{v}_{j-1} + \tilde{v}_{j+1})}{\tilde{v}_j}$$

First, we solve for:

$$\frac{(\tilde{v}_{j-1} + \tilde{v}_{j+1})}{\tilde{v}_j} = G$$
$$(\tilde{v}_{j-1} + \tilde{v}_{j+1}) = \tilde{v}_j G$$

set: $\tilde{v}_i = \varphi^j$

$$\varphi^{j-1} + \varphi^{j+1} = \varphi^j G$$
$$\varphi^{j-1} + \varphi^{j+1} - \varphi^j G = 0$$

divide both side by φ^{j-1}

$$1 + \varphi^2 - \varphi G = 0$$

$$\therefore \varphi_{\pm} = \frac{G \pm \sqrt{(G^2 - 4)}}{2}$$
(13)

here we recall that eigenvector of \tilde{v}_j should be a linear combination of φ s.t.

$$\tilde{v}_j = c_1 \varphi_+^j + c_2 \varphi_-^j$$

and we know the boundary condition on (12), when j = N then $\tilde{v}_N = 0$, to satisfies this condition, then $c_1 = -c_2$, then

$$\tilde{v}_i = c_1(\varphi_+^j - \varphi_-^j)$$

 $\therefore \varphi^j_+$ and φ^j_- should be distinguished.

Hence, we choose G < 2 for (13) we then rewrite (13) as:

$$\varphi_{\pm} = \frac{G \pm \sqrt{-(4 - G^2)}}{2}$$
$$\varphi_{\pm} = \frac{G \pm i\sqrt{(4 - G^2)}}{2}$$

then, the possible solution is

$$\varphi_{\pm}^{j} = \cos(\theta j) \pm i \sin(\theta j)$$

$$\therefore \tilde{v}_{j} = c_{1}(\cos(\theta j) + i \sin(\theta j) - (\cos(\theta j) - i \sin(\theta j)))$$

$$\tilde{v}_{j} = c_{1}(2i \sin(\theta j))$$

we choose $c_1 = -\frac{i}{2}$ so that

$$\tilde{v}_j = \sin(\theta j)$$

inserting boundary condition, when j = N then $\tilde{v}_N = 0$, we get:

$$0 = \sin(\theta N)$$

$$\theta = m\pi, \ \forall m : 2, 4, 6 \cdots N, \ even$$

$$\theta = \frac{m\pi}{N}$$

then, we know that from imaginer triangle

$$\cos \theta = \frac{G}{2}$$

$$\therefore G = 2\cos(\theta) = 2\cos(\frac{m\pi}{N})$$

Second, we solve for:

$$(4 - \lambda) - \frac{(v_{i-1} + v_{i+1})}{v_i} = G$$

multiplied both side by v_i , we get:

$$v_i(4 - \lambda) - (v_{i-1} + v_{i+1}) = v_i G$$

$$v_i(4 - \lambda) - (v_{i-1} + v_{i+1}) - v_i G = 0$$

$$v_i(4 - \lambda - G) - (v_{i-1} + v_{i+1}) = 0$$

 $set v_i = \xi^i$

$$\xi^{i}(4 - \lambda - G) - (\xi^{i-1} + \xi^{i+1}) = 0$$

divide both side by ξ^{i-1}

$$\xi(4 - \lambda - G) - (1 + \xi^2) = 0$$

$$\xi^2 - \xi(4 - \lambda - G) + 1 = 0$$

$$\therefore \xi_{\pm} = \frac{(4 - \lambda - G) \pm \sqrt{(4 - \lambda - G)^2 - 4}}{2}$$

 $take \ 2H = (4 - \lambda - G)$

$$\xi_{\pm} = \frac{2H \pm \sqrt{(2H)^2 - 4}}{2}$$

$$\xi_{\pm} = H \pm \sqrt{H^2 - 1}$$
(14)

here we recall that eigenvector of v_i should be a linear combination of ξ s.t.

$$v_i = d_1 \xi_+^i + d_2 \xi_-^i$$

and we know the boundary condition on (12), when i = N then $v_N = 0$, to satisfies this condition, then $d_1 = -d_2$, then

$$v_i = d_1(\xi + ^j - \xi_-^j)$$

 $\therefore \xi_{+}^{j}$ and ξ_{-}^{j} should be distinguished.

Hence, we choose H < 1 for (14) we then rewrite (14) as:

$$\xi_{\pm} = H \pm \sqrt{-(1 - H^2)}$$

 $\xi_{\pm} = H \pm i\sqrt{(1 - H^2)}$

then, the possible solution is

$$\xi_{+}^{i} = \cos(\theta i) \pm I \sin(\theta i)$$

here i use "I" as imaginer number, so it is not confusing between "i" index and "I" imaginer number.

$$\therefore v_i = d_1(\cos(\theta i) + I\sin(\theta i) - (\cos(\theta i) - I\sin(\theta i)))$$
$$v_i = d_1(2I\sin(\theta i))$$

we choose $d_1 = -\frac{I}{2}$ so that

$$v_i = \sin(\theta i)$$

inserting boundary condition, when i = N then $v_N = 0$, we get:

$$0 = \sin(\theta N)$$

$$\theta = m\pi, \ \forall m : 2, 4, 6 \cdots N, \ even$$

$$\theta = \frac{m\pi}{N}$$

then, we know that from imaginer triangle

$$\cos \theta = H$$

$$\therefore H = \cos(\theta) = \cos(\frac{m\pi}{N})$$
substitute H and G to $2H = (4 - \lambda - G)$, we get

$$2\cos\left(\frac{m\pi}{N}\right) = 4 - \lambda - 2\cos\left(\frac{m\pi}{N}\right)$$
$$\lambda = 4 - 2\cos\left(\frac{m\pi}{N}\right) - 2\cos\left(\frac{m\pi}{N}\right)$$
$$\therefore \lambda = 4 - 2\left(\cos\left(\frac{m\pi}{N}\right) + \cos\left(\frac{m\pi}{N}\right)\right)$$

for the eigenvectors, we already know that

$$w_{i,j} = v_i \tilde{v}_j$$

$$v_i = \sin(\theta i)$$

$$\tilde{v}_j = \sin(\theta j)$$

$$\therefore w_{i,j} = \sin\left(\frac{m\pi i}{N}\right) \sin\left(\frac{m\pi j}{N}\right)$$

- (b) I attached the Python Code that solves (12) in 1
- (c) Minimal Number of Iterations K such the:

$$\max_{1 \le i, j \le N-1} \left| v_{i,j}^{(K+1)} - v_{i,j}^{(K)} \right| \le 10^{-4}$$

- (a) N = 10
 - i. Jacobi, K = 53
 - ii. Gauss-Seidel, K = 33
 - iii. SOR (omega=1.5), K = 12
- (b) N = 20
 - i. Jacobi, K = 137
 - ii. Gauss-Seidel, K = 94
 - iii. SOR (omega=1.5), K=45
- (c) N = 50
 - $i. \ Jacobi, \ K=1$
 - ii. Gauss-Seidel, K = 173
 - iii. SOR (omega=1.5), K = 148
- (d) Optimal ω_b for SOR:
 - (a) N = 10, $\omega_b = 1.16527511683289$
 - (b) N = 20, $\omega_b = 1.2933223938735654$
 - (c) N = 50, $\omega_b = 1.3746381523099196$

1 Attachment

```
import numpy as np
 \#Function\ Definition
 #Function to create Right Hand Side and Left Hand Side Matrix def const_mat(N):
               h = 1/N
A = np.zeros((N,N))
b = np.zeros((N,N))
                for i in range(N):
                            A[i][j] = 0
f i==j-1:
                                             elif i==j
                                             elif i=-j·i.
    A[i][j] = -1
elif i==j+1:
    A[i][j] = -1
                                             elif i==j-4:
A[i][j] = -1
elif i==j+4:
              A[i][j] = -1 return A,b
 #Jacobi
 def jacobi(b):
               v = np.zeros_like(b)
N = len(b)
               for k in range(1000):
    v_old = np.copy(v)
                             v_old = np.copy(v)
for i in range(N-2):
    for j in range(N-2):
        v[i+1][j+1] = 0.25 * (v_old[i][j+1] + v_old[i+2][j+1] + v_old[i+1][j] + v_old[i+1][j+2] + b[i+1][j+1])
#print(np.max(v))
if np.max(np.abs(v - v_old)) < 1e-4:
    hreat</pre>
                return (k)
 #Gauss-Seidel
 def gauss_seidel(b):
               v = np.zeros_like(b)
N = len(b)
              N = len(b)
for k in range(1000):
    v_old = np.copy(v)
    for i in range(N-2):
        for j in range(N-2):
            v[i+1][j+1] = 0.25 * (v[i][j+1] + v[i+2][j+1] + v[i+1][j] + v[i+1][j+2] + b[i+1][j+1])
    #print(np.max(v))
    if np.max(np.abs(v - v_old)) < 1e-4:
        break</pre>
                return (k)
  #SOR
 def sor(omega,b):
               v = np.zeros_like(b)
N = len(b)
for k in range(1000):
                              v_old = np.copy(v)
for i in range(N-2):
                                           To provide the control of the contro
                              #print(np.max(v))
if np.max(np.abs(v - v_old)) < 1e-4:
                                             break
               return (k)
 #Calculate Optimum Omega for SOR
#Calculate Optimum Omega for SOR
def calc_op_omega(A):
   D = np.diag(np.diag(A))
   L = np.tril(A) - D
   U = np.tril(A) - D
   T = 1/np.diag(A)*(L+U)
   rho = np.max(np.linalg.eigvals(T))
   opt_omega = 2/(1+np.sqrt(1-rho*rho))
               return opt_omega
 #main
Ns = [10,20,30,40,50]
om = 1.5
  for n in Ns:
              matA, matB = const_mat(n)
iterj = jacobi(matB)
iterg = gauss_seidel(matB)
iters = sor(om, matB)
op_om = calc_op_omega(matA)
               op_om = calc_op_omega(mata/)
print("Nu=",n)
print("Minimum_Iteration_Jacobi_u=",iterj)
print("Minimum_Iteration_Gauss-Seidel_u=",iterg)
print("Minimum_Iteration_SOR_using_omega(1.5)_u=",iters)
print("Optimal_omega_for_SOR_u=",op_om)
```