

#Exercise 6.2

#15.2.2

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import numpy as np

import pdb

#exercise6.2.2

def jacobi(A, b, x, tol):

d = np.diag(A)

d = [x ** -1 for x in d]

D = np.diag(d)

L = np.tril(A, -1)

U = np.triu(A, 1)

B = np.dot(D, L + U)

g = np.dot(D, b)

count = 0

while True:

new_x = np.dot(B, x) + g

count += 1

c = np.linalg.norm((new_x - x), np.inf)

x = new_x

if c < tol:

print(count)

print(c)

return new_x

def sor(A, b, x, omega, tol):

d = np.diag(A)

D = np.diag(d)

L = np.tril(A, -1)

U = np.triu(A, 1)

T = np.linalg.inv(D - omega * L)

B = np.dot(T, (1 - omega) * D + omega * U)

g = omega * T

count = 0

while True:

new_x = np.dot(B, x) + g

count += 1

c = np.linalg.norm((new_x - x), np.inf)

x = new_x

if c < tol:

print(omega)

print(count)

```

    print(c)
    return new_x

```

```

def cg(A, b, x, tol):
    r = b - np.dot(A, x)
    count = 0
    x = [x]
    r = [r]
    while True:
        count += 1
        if count == 1:
            p = r[0]
        else:
            beta = np.dot(np.transpose(r[-1]), r[-1]) / np.dot(np.transpose(r[-2]), r[-2])
            p = r[-1] + beta * p
        alpha = np.dot(np.transpose(r[-1]), r[-1]) / np.dot(np.dot(np.transpose(p), A), p)
        x_new = x + alpha * p
        r.append(r[-1] - alpha * np.dot(A, p))
        c = np.linalg.norm((x_new - x), np.inf)
        x = x_new
        if c < tol:
            print(count)
            print(c)
            return x_new

```

```

def main():
    N = 10
    h = 1/(N+1)
    f = lambda x, y: (x ** 2 + y ** 2) * np.exp(x * y)

```

#exercise 6.2.1

```

x = np.linspace(0, 1, N + 2)
y = np.linspace(0, 1, N + 2)

```

```

rank = N * N
A = np.diag([4 for i in range(rank)]) + \
    np.diag([-1 for i in range(rank - 1)], -1) + \
    np.diag([-1 for i in range(rank - 1)], 1) + \
    np.diag([-1 for i in range(rank - N)], -N) + \
    np.diag([-1 for i in range(rank - N)], N)
b = np.zeros(N * N)
b = np.transpose(b)
for i in range(N):

```

```

        for j in range(N):
            t = j + i * N
            b[t] = f(x[j + 1], y[i + 1]) * (h**2)
            if i == 0:
                b[t] += 1
            elif i == N - 1:
                b[t] += np.exp(x[j + 1])
            else:
                continue
            if j == 0:
                b[t] += 1
            elif j == N - 1:
                b[t] += np.exp(y[i + 1])
            else:
                continue
    print('exercise6.2.1')
    print(A)
    print(b)
    print('exercise6.2.2')
    print('jacobi')
    tol = 1e-5
    init = np.zeros(N*N).transpose() + 1
    result = jacobi(A, b, init, tol)
    print(result)
    print('SOR')
    omega_list = [1, 1.25, 1.5, 1.75]
    for omega in omega_list:
        result = sor(A, b, init, omega, tol)
        print(result)
    print('exercise6.2.3')
    print('CG')
    result = cg(A, b, init, tol)
    print(result)

if __name__ == '__main__':
    main()、

```

运行结果如下：

exercise6.2.1

```

[[ 4 -1  0 ...,  0  0  0]
 [-1  4 -1 ...,  0  0  0]
 [ 0 -1  4 ...,  0  0  0]
 ...,
 [ 0  0  0 ...,  4 -1  0]

```

[0 0 0 ..., -1 4 -1]
 [0 0 0 ..., 0 -1 4]]

[2.00013774e+00	1.00034720e+00	1.00070016e+00	1.00120015e+00
1.00185075e+00	1.00265562e+00	1.00361846e+00	1.00474303e+00
1.00603318e+00	2.10266222e+00	3.47198376e-04	5.64775812e-04
9.33056355e-04	1.45939529e-03	2.15139108e-03	3.01689210e-03
4.06400354e-03	5.30109444e-03	6.73680499e-03	8.38005387e-03
7.00159347e-04	9.33056355e-04	1.32435574e-03	1.88555756e-03
2.62873214e-03	3.56654382e-03	4.71227552e-03	6.07985433e-03
7.68387790e-03	9.53964196e-03	1.20014860e-03	1.45939529e-03
1.88555756e-03	2.49463268e-03	3.30367508e-03	4.33085514e-03
5.59552078e-03	7.11826227e-03	8.92098019e-03	1.10269568e-02
1.85075384e-03	2.15139108e-03	2.62873214e-03	3.30367508e-03
4.19884196e-03	5.33869871e-03	6.74968259e-03	8.46033753e-03
1.05014580e-02	1.29062419e-02	2.65562193e-03	3.01689210e-03
3.56654382e-03	4.33085514e-03	5.33869871e-03	6.62175849e-03
8.21476284e-03	1.01557352e-02	1.24862634e-02	1.52517895e-02
3.61845968e-03	4.06400354e-03	4.71227552e-03	5.59552078e-03
6.74968259e-03	8.21476284e-03	1.00352144e-02	1.22603677e-02
1.49448954e-02	1.81493163e-02	4.74303469e-03	5.30109444e-03
6.07985433e-03	7.11826227e-03	8.46033753e-03	1.01557352e-02
1.22603677e-02	1.48370892e-02	1.79564477e-02	2.16975129e-02
6.03317617e-03	6.73680499e-03	7.68387790e-03	8.92098019e-03
1.05014580e-02	1.24862634e-02	1.49448954e-02	1.79564477e-02
2.16107745e-02	2.60097865e-02	2.10266222e+00	1.20777616e+00
1.32308160e+00	1.44957797e+00	1.58836335e+00	1.74064426e+00
1.90774643e+00	2.09112652e+00	2.29238519e+00	4.99534639e+00]

exercise6.2.2

jacobi

587

9.89753129249e-06

[6.19844679e-01	-2.18419336e-02	2.63443918e-01	1.26359626e-01
1.91628934e-01	1.72204549e-01	1.45857628e-01	2.48835853e-01
-2.74786408e-02	7.27640645e-01	-4.57405426e-01	2.04417522e-01
-1.57603117e-01	4.06783027e-02	-6.32399975e-02	-2.36602269e-02
-8.63781667e-04	-1.08991003e-01	1.39459066e-01	-3.23029609e-01
2.78049886e-01	-1.80273059e-01	1.22786268e-01	-6.67906547e-02
4.64435041e-02	-1.04642367e-02	-5.70890061e-03	5.38118957e-02
-9.16229353e-02	1.55325104e-01	-1.50816696e-01	1.16744755e-01
-8.51818432e-02	5.91113664e-02	-4.26796774e-02	2.83194230e-02
-1.49660479e-02	-2.87533235e-03	2.55485832e-02	-4.63232476e-02
5.43145177e-02	-4.92816327e-02	4.39361628e-02	-3.93336185e-02
4.01127128e-02	-4.08725507e-02	4.56886167e-02	-4.58109650e-02
4.75117672e-02	-3.89050186e-02	3.09770410e-02	-1.57552045e-02

6.43072588e-04	1.74394887e-02	-3.34047894e-02	5.46694141e-02
-7.43939802e-02	1.01340419e-01	-1.20416883e-01	1.36322125e-01
-1.20945468e-01	8.36605568e-02	-4.46649210e-02	6.62947596e-03
2.66975059e-02	-6.34232047e-02	1.04053816e-01	-1.54622214e-01
2.08941794e-01	-2.49806559e-01	2.36403847e-01	-1.49249026e-01
9.24027286e-02	-2.04303623e-02	-9.87749888e-03	7.64514673e-02
-1.13775610e-01	2.16378818e-01	-2.96010228e-01	4.35675116e-01
-4.20903015e-01	1.89799096e-01	-1.49217194e-01	-3.45042823e-04
-3.47779690e-02	-1.08603102e-01	7.04499729e-02	-2.86298088e-01
3.40974773e-01	-7.54308554e-01	8.27743387e-01	-3.31130747e-02
3.22673796e-01	2.14705043e-01	2.68417711e-01	3.34754335e-01
2.41801476e-01	5.35325722e-01	-5.68966823e-03	1.43883456e+00]

SOR

1

129

9.39455328365e-06

[[3.09862101e-01	-1.10874412e-01	4.58707574e-02 ...,	-3.71368626e-04
	8.00734743e-04	-8.55954460e-04]		
[-1.10874431e-01	3.49532678e-01	-1.27285595e-01 ...,	-1.57876921e-04
	-5.31173749e-04	8.00733881e-04]		
[4.58708275e-02	-1.27285674e-01	3.56321121e-01 ...,	1.01442613e-03
	-1.57883535e-04	-3.71363701e-04]		
...,				
[-3.71170627e-04	-1.58171801e-04	1.01473381e-03 ...,	3.56321131e-01
	-1.27285606e-01	4.58707640e-02]		
[8.00571322e-04	-5.30921835e-04	-1.58163439e-04 ...,	-1.27285663e-01
	3.49532671e-01	-1.10874408e-01]		
[-8.55849644e-04	8.00566037e-04	-3.71165203e-04 ...,	4.58708412e-02
	-1.10874442e-01	3.09862108e-01]]		

1.25

82

9.38581286303e-06

[[3.09862111e-01	-1.10874481e-01	4.58708237e-02 ...,	-3.71287778e-04
	8.00679602e-04	-8.55942653e-04]		
[-1.10874433e-01	3.49532753e-01	-1.27285675e-01 ...,	-1.58027259e-04
	-5.31059821e-04	8.00687755e-04]		
[4.58708166e-02	-1.27285727e-01	3.56321190e-01 ...,	1.01463145e-03
	-1.58049571e-04	-3.71278722e-04]		
...,				
[-3.71206910e-04	-1.58098888e-04	1.01468371e-03 ...,	3.56321218e-01
	-1.27285698e-01	4.58708433e-02]		
[8.00602059e-04	-5.30991534e-04	-1.58109930e-04 ...,	-1.27285708e-01
	3.49532728e-01	-1.10874464e-01]		
[-8.55869435e-04	8.00616078e-04	-3.71206962e-04 ...,	4.58708515e-02

```

-1.10874463e-01  3.09862136e-01]]
1.5
54
9.01359846291e-06
[[ 3.09862132e-01 -1.10874369e-01  4.58708090e-02 ..., -3.71146486e-04
   8.00556882e-04 -8.55810182e-04]
 [ -1.10874463e-01  3.49532600e-01 -1.27285644e-01 ..., -1.58157280e-04
  -5.30941209e-04  8.00537631e-04]
 [ 4.58708473e-02 -1.27285567e-01  3.56321147e-01 ..., 1.01469131e-03
  -1.58113969e-04 -3.71159223e-04]
 ...,
 [ -3.71215423e-04 -1.58114477e-04  1.01466338e-03 ..., 3.56321095e-01
  -1.27285602e-01  4.58707736e-02]
 [ 8.00610365e-04 -5.30965503e-04 -1.58097472e-04 ..., -1.27285602e-01
  3.49532642e-01 -1.10874396e-01]
 [ -8.55875438e-04  8.00591089e-04 -3.71211828e-04 ..., 4.58707816e-02
  -1.10874406e-01  3.09862087e-01]]
1.75
90
7.95773223646e-06
[[ 3.09862087e-01 -1.10874428e-01  4.58707890e-02 ..., -3.71214972e-04
   8.00597224e-04 -8.55869846e-04]
 [ -1.10874406e-01  3.49532674e-01 -1.27285623e-01 ..., -1.58095916e-04
  -5.30976133e-04  8.00603351e-04]
 [ 4.58707846e-02 -1.27285642e-01  3.56321123e-01 ..., 1.01466110e-03
  -1.58106202e-04 -3.71211078e-04]
 ...,
 [ -3.71236368e-04 -1.58115019e-04  1.01464583e-03 ..., 3.56321126e-01
  -1.27285642e-01  4.58707978e-02]
 [ 8.00579086e-04 -5.30994149e-04 -1.58117985e-04 ..., -1.27285642e-01
  3.49532658e-01 -1.10874422e-01]
 [ -8.55885388e-04  8.00593631e-04 -3.71222146e-04 ..., 4.58707943e-02
  -1.10874425e-01  3.09862100e-01]]

```

exercise6.2.3

CG

35

```

5.56182387346e-06
[[ 1.11261174  1.1236006  1.12003575  1.11825228  1.12096855  1.12931851
   1.14657453  1.18327922  1.27367188  1.53687082  1.32670848  1.26140779
   1.23758981  1.23080479  1.23445266  1.24707546  1.27008184  1.3081276
   1.36850414  1.44444053  1.39559631  1.35716723  1.337178  1.33146508
   1.33681057  1.35143192  1.37448591  1.40534416  1.44103984  1.4684107
   1.45336859  1.43355375  1.42116537  1.41918165  1.4272639  1.4437892
   1.46637361  1.49164338  1.51421642  1.52525419  1.51471331  1.50105435

```

1.49286259	1.4943377	1.50597051	1.52575672	1.54998034	1.57352125
1.5900075	1.59264921	1.57732536	1.56093612	1.55226422	1.55603235
1.57232493	1.59794796	1.62752033	1.65399348	1.66914163	1.66510367
1.63834701	1.61008361	1.59565931	1.6008717	1.6240102	1.65956829
1.6999446	1.73563504	1.75497571	1.74502485	1.69725699	1.641328
1.61470533	1.62218951	1.65652625	1.70815571	1.76701967	1.82136601
1.85515648	1.84461366	1.7595849	1.6379648	1.59356458	1.60953621
1.66328927	1.73935285	1.82635214	1.91281586	1.98171408	1.99699081
1.85247071	1.55064494	1.50436789	1.55018056	1.6372403	1.74712814
1.8712754	2.00387504	2.14028228	2.28315489]]		

[Finished in 0.3s]

结果分析：

Jacobi 迭代的收敛速度最慢，其次为 C—S 迭代，再为 $\omega = 1.75$ 的 SOR 迭代，再为 $\omega = 1.25$ 的 SOR 迭代，再为 $\omega = 1.5$ 的 SOR 迭代，CG 方法的收敛速度最快。此外，由 SOR 迭代的结果可以得知，最佳松弛因子位于 1.25 和 1.75 之间。