

# 作业五

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本文代码均用 Python 实现

## 1

### LU 分解代码

```
1 import numpy as np
2
3 def lu_decomposition(A):
4     """ 对方阵 A 进行 LU 分解, 返回 L 和 U """
5     n = len(A)
6     L = np.eye(n)
7     U = A.astype(float).copy()
8
9     for i in range(n):
10         for j in range(i+1, n):
11             if U[i, i] == 0:
12                 raise ValueError("不存在 LU 分解!")
13             factor = U[j, i] / U[i, i]
14             L[j, i] = factor
15             U[j, i:] -= factor * U[i, i:]
16
17     return L, U
18
19 # 测试
20 A = np.array([[2, -1, 3],
21               [1, 2, 1],
22               [2, 4, -2]], dtype=float)
23 L, U = lu_decomposition(A)
24
25 print("L:")
26 print(L)
27 print("U:")
28 print(U)
```

测试输出结果为

输出结果

```
1  L:
2      [[1.  0.  0. ]
3       [0.5 1.  0. ]
4       [1.  2.  1. ]]
5  R:
6      [[ 2.  -1.  3. ]
7       [ 0.   2.5 -0.5]
8       [ 0.   0.  -4. ]]
```

因此  $\det A = \det R = -20$ .

## 2

### Cholesky 分解代码

```
1  import numpy as np
2
3  def cholesky_decomposition(A):
4      """ 对称正定矩阵 A 进行 Cholesky 分解, 返回 L """
5      n = A.shape[0]
6      L = np.zeros_like(A)
7
8      for i in range(n):
9          for j in range(i + 1):
10             sum_k = sum(L[i, k] * L[j, k] for k in range(j))
11
12             if i == j:
13                 L[i, j] = np.sqrt(A[i, i] - sum_k)
14             else:
15                 L[i, j] = (A[i, j] - sum_k) / L[j, j]
16
17         return L
18
19  # 测试
20  A = np.array([[5, -2, 0],
21               [-2, 3, -1],
22               [0, -1, 1]], dtype=float)
23
24  L = cholesky_decomposition(A)
25  print("L:\n", L)
26  print("L.T:\n", L.T)
```

测试输出结果为

### 输出结果

```
1  L:
2      [[ 2.23606798  0.          0.          ]
3       [-0.89442719  1.4832397  0.          ]
4       [ 0.          -0.67419986  0.73854895]]
5  L.T:
```

```

6      [[ 2.23606798 -0.89442719  0.          ]
7      [ 0.          1.4832397  -0.67419986]
8      [ 0.          0.          0.73854895]]

```

### 3

仅需  $A$  是正定阵即可，即其各阶顺序主子式大于 0. 计算得结果为  $a \in (-\sqrt{3}, \sqrt{3})$ .

### 4

$$u_1 = \begin{pmatrix} 2 \\ -1 \\ 2 \end{pmatrix}$$

$$u_2 = \alpha_2 - \frac{\alpha_2 \cdot u_1}{u_1 \cdot u_1} u_1 = \frac{2}{9} \begin{pmatrix} -11 \\ 10 \\ 16 \end{pmatrix}$$

$$\text{单位化即有 } Q = \begin{pmatrix} \frac{2}{3} & \frac{-11\sqrt{53}}{159} \\ \frac{1}{3} & \frac{10\sqrt{53}}{159} \\ -\frac{1}{3} & \frac{16\sqrt{53}}{159} \end{pmatrix}$$

$$R = \begin{pmatrix} 3 & -\frac{3}{7} \\ 0 & \frac{2\sqrt{53}}{3} \end{pmatrix}$$

Householder 变换计算 QR 分解代码

```

1  import numpy as np
2
3  def householder_qr(A):
4      """ 使用 Householder 变换计算矩阵 A 的 QR 分解 """
5      m, n = A.shape
6      Q = np.eye(m)
7      R = A.copy().astype(float)
8
9      for k in range(n):
10         # 选取列向量
11         x = R[k:, k]
12
13         # 计算法向量 v
14         e1 = np.zeros_like(x)
15         e1[0] = np.linalg.norm(x) * (1 if x[0] >= 0 else -1)
16         v = x + e1
17         v = v / np.linalg.norm(v)
18
19         # 构造 Householder 变换矩阵 H_k = E - 2uu.T
20         H_k = np.eye(m)

```

```

21     H_k[k:, k:] -= 2.0 * np.outer(v, v)
22
23     # 更新 R 和 Q
24     R = H_k @ R
25     Q = Q @ H_k.T # 注意 Q 需要不断右乘 H_k 的转置
26
27     return Q, R
28
29 # 测试
30 A = np.array([[3, 14, 9],
31               [6, 43, 3],
32               [6, 22, 15]], dtype=float)
33
34 Q_A, R_A = householder_qr(A)
35
36 print("Q_A:\n", Q_A)
37 print("R_A:\n", R_A)
38
39 B = np.array([[1, 1, 1],
40               [2, -1, -1],
41               [2, -4, 10]], dtype=float)
42
43 Q_B, R_B = householder_qr(B)
44
45 print("Q_B\n", Q_B)
46 print("R_B:\n", R_B)

```

输出结果为

输出结果

```

1  Q_A:
2  [[-0.33333333  0.13333333  0.93333333]
3   [-0.66666667 -0.73333333 -0.13333333]
4   [-0.66666667  0.66666667 -0.33333333]]
5  R_A:
6  [[-9.00000000e+00 -4.80000000e+01 -1.50000000e+01]
7   [ 1.77635684e-16 -1.50000000e+01  9.00000000e+00]
8   [-4.21884749e-16 -6.21724894e-16  3.00000000e+00]]
9  Q_B
10 [[-0.33333333 -0.66666667  0.66666667]
11  [-0.66666667 -0.33333333 -0.66666667]
12  [-0.66666667  0.66666667  0.33333333]]
13 R_B:
14 [[-3.00000000e+00  3.00000000e+00 -6.33333333e+00]
15  [-6.66133815e-16 -3.00000000e+00  6.33333333e+00]
16  [ 6.66133815e-16  4.44089210e-16  4.66666667e+00]]

```

极小数值视为 0

因此  $B^{-1} = R^{-1}Q^T$ , 计算结果为

## B 的逆

```
1 [[ 3.33333333e-01  3.33333333e-01  3.81822733e-18]
2 [ 5.23809524e-01 -1.90476190e-01 -7.14285714e-02]
3 [ 1.42857143e-01 -1.42857143e-01  7.14285714e-02]]
```

## 5

## Givens 变换计算 QR 分解代码

```
1 import numpy as np
2
3 def givens_rotation(A):
4     """ 使用 Givens 变换计算矩阵 A 的 QR 分解 """
5     m, n = A.shape
6     Q = np.eye(m)
7     R = A.copy().astype(float)
8
9     for j in range(n):
10         for i in range(m - 1, j, -1): # 从底部向上遍历行
11             if R[i, j] != 0: # 只在非零元素时进行旋转
12                 a, b = R[j, j], R[i, j]
13                 r = np.hypot(a, b) # 计算 sqrt(a^2 + b^2) 避免溢出
14                 c, s = a / r, -b / r
15
16                 # 构造 Givens 旋转矩阵
17                 G = np.eye(m)
18                 G[[j, i], [j, i]] = c
19                 G[j, i], G[i, j] = -s, s
20
21                 R = G @ R
22                 Q = Q @ G.T
23
24     return Q, R
25
26 # 测试
27 A = np.array([[2, 2, 1],
28               [0, 2, 2],
29               [2, 1, 2]], dtype=float)
30
31 Q, R = givens_rotation(A)
32
33 print("Q:\n", Q)
34 print("R:\n", R)
```

## 输出结果

```
1 Q:
2 [[ 0.70710678  0.23570226 -0.66666667]
3 [ 0.          0.94280904  0.33333333]
```

```
4      [ 0.70710678 -0.23570226  0.66666667]]
5      R:
6      [[ 2.82842712e+00  2.12132034e+00  2.12132034e+00]
7      [ 0.00000000e+00  2.12132034e+00  1.64991582e+00]
8      [ 0.00000000e+00 -4.96469267e-17  1.33333333e+00]]
```