

第二次作业

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题目 1: 分别用高斯消去法和高斯-若当消去法求解线性方程组

$$\begin{pmatrix} 1 & 4 & 2 \\ 1 & 5 & 2 \\ 0 & 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 2 \\ 3 \\ 1 \end{pmatrix}$$

解:

(1) 高斯消去法:

$$(A:b) = \begin{pmatrix} 1 & 4 & 2 & : & 2 \\ 1 & 5 & 2 & : & 3 \\ 0 & 1 & 1 & : & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 4 & 2 & : & 2 \\ 0 & 1 & 0 & : & 1 \\ 0 & 1 & 1 & : & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 4 & 2 & : & 2 \\ 0 & 1 & 0 & : & 1 \\ 0 & 0 & 1 & : & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 4 & 2 & : & 2 \\ 0 & 1 & 0 & : & 1 \\ 0 & 0 & 1 & : & 0 \end{pmatrix}$$

把原方程等价约化为:

$$\begin{cases} x_1 + 4x_2 + 2x_3 = 2 \\ x_2 + 0x_3 = 1 \\ x_3 = 0 \end{cases}$$

据之回代解得:

$$\begin{cases} x_1 = -2 \\ x_2 = 1 \\ x_3 = 0 \end{cases}$$

(2) 高斯-若当消去法

$$(A:b) = \begin{pmatrix} 1 & 4 & 2 & : & 2 \\ 1 & 5 & 2 & : & 3 \\ 0 & 1 & 1 & : & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 4 & 2 & : & 2 \\ 0 & 1 & 0 & : & 1 \\ 0 & 1 & 1 & : & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 2 & : & -2 \\ 0 & 1 & 0 & : & 1 \\ 0 & 0 & 1 & : & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & : & -2 \\ 0 & 1 & 0 & : & 1 \\ 0 & 0 & 1 & : & 0 \end{pmatrix}$$

所以, 解为 $x = \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix}$

题目 2: 用选列主元的高斯消去法求解线性方程组

$$\begin{pmatrix} 1 & 2 & 1 & -2 \\ 2 & 5 & 3 & -2 \\ -2 & -2 & 3 & 5 \\ 1 & 3 & 2 & -3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 4 \\ 7 \\ -1 \\ 0 \end{pmatrix}$$

解:

$$\begin{aligned} (A:b) &= \begin{pmatrix} 1 & 2 & 1 & -2 & \vdots & 4 \\ \mathbf{2} & 5 & 3 & -2 & \vdots & 7 \\ -2 & -2 & 3 & 5 & \vdots & -1 \\ 1 & 3 & 2 & -3 & \vdots & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2.5 & 1.5 & -1 & \vdots & 3.5 \\ 0 & -0.5 & -0.5 & -1 & \vdots & 0.5 \\ 0 & \mathbf{3} & 6 & 3 & \vdots & 6 \\ 0 & 0.5 & 0.5 & -2 & \vdots & -3.5 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2.5 & 1.5 & -1 & \vdots & 3.5 \\ 0 & 1 & 2 & 1 & \vdots & 2 \\ 0 & 0 & \mathbf{0.5} & -0.5 & \vdots & 1.5 \\ 0 & 0 & -0.5 & -2.5 & \vdots & -4.5 \end{pmatrix} \\ &\rightarrow \begin{pmatrix} 1 & 2.5 & 1.5 & -1 & \vdots & 3.5 \\ 0 & 1 & 2 & 1 & \vdots & 2 \\ 0 & 0 & 1 & -1 & \vdots & 3 \\ 0 & 0 & 0 & \mathbf{-3} & \vdots & -3 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2.5 & 1.5 & -1 & \vdots & 3.5 \\ 0 & 1 & 2 & 1 & \vdots & 2 \\ 0 & 0 & 1 & -1 & \vdots & 3 \\ 0 & 0 & 0 & 1 & \vdots & 1 \end{pmatrix} \end{aligned}$$

回代解得:

$$\begin{cases} x_1 = 16 \\ x_2 = -7 \\ x_3 = 4 \\ x_4 = 1 \end{cases}$$

题目 3: 用选全主元的高斯-若当消去法求解如下线性方程组

$$\begin{pmatrix} 2 & 1 & -2 \\ 3 & 1 & -4 \\ 1 & -1 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 6 \\ 6 \\ 0 \end{pmatrix}$$

解:

$$\begin{aligned}
 (A:b) &= \begin{pmatrix} 2 & 1 & -2 & \vdots & 6 \\ 3 & 1 & -4 & \vdots & 6 \\ 1 & -1 & 2 & \vdots & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -0.25 & -0.75 & \vdots & -1.5 \\ 0 & 0.5 & 0.5 & \vdots & 3 \\ 0 & -0.5 & 2.5 & \vdots & 3 \end{pmatrix} \\
 &\rightarrow \begin{pmatrix} 1 & 0 & -0.4 & \vdots & -0.6 \\ 0 & 1 & -0.2 & \vdots & 1.2 \\ 0 & 0 & 0.6 & \vdots & 2.4 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & 0 & \vdots & 1 \\ 0 & 1 & 0 & \vdots & 2 \\ 0 & 0 & 1 & \vdots & 4 \end{pmatrix}
 \end{aligned}$$

所以解得

$$\begin{cases} x_1 = \tilde{x}_2 = 2 \\ x_2 = \tilde{x}_3 = 4 \\ x_3 = \tilde{x}_1 = 1 \end{cases}$$

题目 4: 用克洛特分解法分解以下矩阵

$$(1) \mathbf{A} = \begin{pmatrix} 1 & 2 & 1 \\ 0 & 2 & 2 \\ 2 & 4 & 5 \end{pmatrix} \quad (2) \begin{pmatrix} 1 & -1 & 1 \\ 5 & -4 & 3 \\ 2 & 1 & 1 \end{pmatrix}$$

解:

(1)

$$\begin{pmatrix} 1 & 2 & 1 \\ 0 & 2 & 2 \\ 2 & 4 & 5 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 \\ 0 & 2 & 2 \\ 2 & 4 & 5 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 \\ 0 & 2 & 2 \\ 2 & 0 & 5 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 \\ 0 & 2 & 1 \\ 2 & 0 & 5 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 1 \\ 0 & 2 & 1 \\ 2 & 0 & 3 \end{pmatrix}$$

解得:

$$L = \begin{pmatrix} 1 & & \\ 0 & 2 & \\ 2 & 0 & 3 \end{pmatrix}, R = \begin{pmatrix} 1 & 2 & 1 \\ & 1 & 1 \\ & & 1 \end{pmatrix}$$

(2)

$$\begin{pmatrix} 1 & -1 & 1 \\ 5 & -4 & 3 \\ 2 & 1 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 1 \\ 5 & -4 & 3 \\ 2 & 1 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 1 \\ 5 & 1 & 3 \\ 2 & 3 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 1 \\ 5 & 1 & -2 \\ 2 & 3 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & -1 & 1 \\ 5 & 1 & -2 \\ 2 & 3 & 5 \end{pmatrix}$$

解得:

$$L = \begin{pmatrix} 1 & & \\ 5 & 1 & \\ 2 & 3 & 5 \end{pmatrix}, R = \begin{pmatrix} 1 & -1 & 1 \\ & 1 & -2 \\ & & 1 \end{pmatrix}$$

题目 5: 用克洛特分解法求解线性方程组

$$\begin{pmatrix} 1 & 2 & 1 \\ 2 & 2 & 3 \\ -1 & -3 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \\ 2 \end{pmatrix}$$

解:

对 A 进行克洛特分解

$$\begin{pmatrix} 1 & 2 & 1 \\ 2 & 2 & 3 \\ -1 & -3 & 0 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & 1 \\ 2 & 2 & 3 \\ -1 & -3 & 0 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & 1 \\ 2 & -2 & 3 \\ -1 & -1 & 0 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & 1 \\ 2 & -2 & -0.5 \\ -1 & -1 & 0 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & 1 \\ 2 & -2 & -0.5 \\ -1 & -1 & 0.5 \end{pmatrix}$$

所以

$$L = \begin{pmatrix} 1 & & \\ 2 & -2 & \\ -1 & -1 & 0.5 \end{pmatrix}, R = \begin{pmatrix} 1 & 2 & 1 \\ & 1 & -0.5 \\ & & 1 \end{pmatrix}$$

由:

$$\begin{pmatrix} 1 & & \\ 2 & -2 & \\ -1 & -1 & 0.5 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \\ 2 \end{pmatrix}$$

解得:

$$\begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} = \begin{pmatrix} 0 \\ -1.5 \\ 1 \end{pmatrix}$$

由:

$$\begin{pmatrix} 1 & 2 & 1 \\ & 1 & -0.5 \\ & & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ -1.5 \\ 1 \end{pmatrix}$$

解得:

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$

题目 6: 取初值 $x_1^{(0)} = x_2^{(0)} = x_3^{(0)} = 0.0$, 分别用雅可比迭代法与高斯-塞德尔迭代法解线性方程组 (精度要求为 $\epsilon = 10^{-3}$)

$$\begin{pmatrix} 3.0 & 0.15 & -0.09 \\ 0.08 & 4.0 & -0.16 \\ 0.05 & -0.3 & 5.0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 6.09 \\ 11.52 \\ 19.20 \end{pmatrix}$$

解:

相应的迭代公式为

1. 雅可比迭代:

$$\begin{cases} x_1^{k+1} = 2.03 - 0.05x_2^k + 0.03x_3^k \\ x_2^{k+1} = 2.88 - 0.02x_1^k + 0.04x_3^k \\ x_3^{k+1} = 3.84 - 0.01x_1^k + 0.06x_2^k \end{cases}$$

迭代过程为

$$\begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{pmatrix} 2.03 \\ 2.88 \\ 3.84 \end{pmatrix} \rightarrow \begin{pmatrix} 2.0012 \\ 2.9930 \\ 3.9925 \end{pmatrix} \rightarrow \begin{pmatrix} 2.0001 \\ 2.9997 \\ 3.9996 \end{pmatrix} \rightarrow \begin{pmatrix} 2.0000 \\ 3.0000 \\ 4.0000 \end{pmatrix} \rightarrow \begin{pmatrix} 2.0000 \\ 3.0000 \\ 4.0000 \end{pmatrix}$$

所以

$$\begin{pmatrix} x_1^4 \\ x_2^4 \\ x_3^4 \end{pmatrix} = \begin{pmatrix} 2.000 \\ 3.000 \\ 4.000 \end{pmatrix}$$

2. 高斯-塞德尔迭代:

$$\begin{cases} x_1^{k+1} = 2.03 - 0.05x_2^k + 0.03x_3^k \\ x_2^{k+1} = 2.88 - 0.02x_1^{k+1} + 0.04x_3^k \\ x_3^{k+1} = 3.84 - 0.01x_1^{k+1} + 0.06x_2^{k+1} \end{cases}$$

迭代过程为

$$\begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \rightarrow \begin{pmatrix} 2.0300 \\ 2.8394 \\ 2.9901 \end{pmatrix} \rightarrow \begin{pmatrix} 2.0077 \\ 2.9994 \\ 3.9999 \end{pmatrix} \rightarrow \begin{pmatrix} 2.0000 \\ 3.0000 \\ 4.0000 \end{pmatrix} \rightarrow \begin{pmatrix} 2.0000 \\ 3.0000 \\ 4.0000 \end{pmatrix}$$

所以

$$\begin{pmatrix} x_1^3 \\ x_2^3 \\ x_3^3 \end{pmatrix} = \begin{pmatrix} 2.000 \\ 3.000 \\ 4.000 \end{pmatrix}$$

题目 7: 利用定理 3.8 对以下线性方程组讨论雅可比迭代法与高斯-塞德尔迭代法的收敛性。

$$\begin{pmatrix} 1 & 2 & -2 \\ 1 & 1 & 1 \\ 2 & 2 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \\ 1 \end{pmatrix}$$

解:

$$M_J = D^{-1}(L + U) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & -2 & 2 \\ -1 & 0 & -1 \\ -2 & -2 & 0 \end{pmatrix} = \begin{pmatrix} 0 & -2 & 2 \\ -1 & 0 & -1 \\ -2 & -2 & 0 \end{pmatrix}$$

特征值 $\lambda_1 = \lambda_2 = \lambda_3 = 0$, 所以 $\rho(M) = \max_{1 \leq i \leq n} |\lambda_i| = 0 < 1$, 所以雅可比迭代法收敛

$$M_{G-S} = (D - L)^{-1}U = \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -2 & 1 \end{pmatrix} \begin{pmatrix} 0 & -2 & 2 \\ 0 & 0 & -1 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & -2 & 2 \\ 0 & 2 & -3 \\ 0 & 0 & 2 \end{pmatrix}$$

特征值 $\lambda_1 = 0, \lambda_2 = \lambda_3 = 2$, 所以 $\rho(M) = \max_{1 \leq i \leq n} |\lambda_i| = 2 > 1$, 所以高斯-塞德尔迭代法无法收敛

具体代码实现见下一页

此外，在作业的过程中，我将上述方法封装成了一个 `Matrix` 类，方便计算，以下是对高斯消元法，高斯-若当消元法（包括主元的选择方法），克洛特分解法，雅可比迭代法和高斯-塞德尔迭代法的 C++ 实现，复杂度均为课件中所提到的复杂度

```
1 #include<bits/stdc++.h>
2 using namespace std;
3
4 const int N = 5e2 + 5;
5 double mat[N][N];
6
7 struct Matrix {
8     double mat[N][N];
9     int n;
10
11     void read(void) {
12         scanf("%d", &n);
13         for (int i = 0; i < n; i++) {
14             for (int j = 0; j < n + 1; j++) {
15                 scanf("%lf", &mat[i][j]);
16             }
17         }
18     }
19     void printMat(double mat[N][N], int n, int m) {
20         for (int i = 0; i < n; i++) {
21             for (int j = 0; j < m; j++) {
22                 printf("%f ", mat[i][j]);
23             }
24             printf("\n");
25         }
26     }
27     void printVec(double vec[], int n) {
28         for (int i = 0; i < n; i++) {
29             printf("%f ", vec[i]);
30         }
31         printf("\n");
32     }
33
34     // pivot_choice_method must be in ["none", "column", "all"]
35     void GaussianElimination(bool GaussianJordan=true, string
        pivot_choice_method="none") {
```

```

36     double ans[N];
37     int row, col, pos[N];
38     for (int i = 0; i < n; i++) {
39         pos[i] = i;
40     }
41     for (row = 0, col = 0; row < n && col < n; row++, col++) {
42         printf("Stage %d\n", row);
43         if (pivot_choice_method == "column") {
44             int maxrow = row;
45             for (int i = row; i < n; i++) {
46                 if (fabs(mat[i][col]) > fabs(mat[maxrow][col])) {
47                     maxrow = i;
48                 }
49             }
50             for (int j = 0; j < n + 1; j++){
51                 swap(mat[row][j], mat[maxrow][j]);
52             }
53         }
54         else if (pivot_choice_method == "all") {
55             int maxrow = row, maxcol = col;
56             for (int i = row; i < n; i++) {
57                 for (int j = col; j < n; j++) {
58                     if (fabs(mat[i][j]) > fabs(mat[maxrow][maxcol])) {
59                         maxrow = i; maxcol = j;
60                     }
61                 }
62             }
63             for (int j = 0; j < n + 1; j++) {
64                 swap(mat[row][j], mat[maxrow][j]);
65             }
66             for (int i = 0; i < n; i++) {
67                 swap(mat[i][col], mat[i][maxcol]);
68             }
69             swap(pos[col], pos[maxcol]);
70         }
71         if (mat[row][col] == 0) {
72             printf("Fail!\n");
73             return;
74         }

```



```

75     double div = mat[row][col];
76     for (int j = col; j < n + 1; j++) {
77         mat[row][j] /= div;
78     }
79     for (int i = GaussionJordan? 0: row + 1; i < n; i++) {
80         if (i == row) continue;
81         double temp = mat[i][col];
82         for (int j = col; j < n + 1; j++) {
83             mat[i][j] -= mat[row][j] * temp;
84         }
85         mat[i][col] = 0;
86     }
87     printMat(mat, n, n + 1);
88 }
89 for (int i = n - 1; i >= 0; i--) {
90     ans[i] = mat[i][n];
91     for (int j = i + 1; j < n; j++) {
92         ans[i] -= ans[j] * mat[i][j];
93     }
94 }
95 printf("Result: \n");
96 for (int i = 0; i < n; i++) {
97     printf("ans%d = ans'%d = %f\n", pos[i], i, ans[i]);
98 }
99 }
100 void CroutSplit(void) {
101     double l[N][N], u[N][N], x[N], y[N];
102     for (int i = 0; i < n; i++) {
103         double sum;
104         for (int j = 0; j <= i; j++) {
105             sum = 0;
106             for (int k = 0; k < j; k++) {
107                 sum += l[i][k] * u[k][j];
108             }
109             l[i][j] = mat[i][j] - sum;
110         }
111         for (int j = i + 1; j < n; j++) {
112             sum = 0;
113             for (int k = 0; k < i; k++) {

```

```

114         sum += l[i][k] * u[k][j];
115     }
116     u[i][j] = (mat[i][j] - sum) / l[i][i];
117 }
118 u[i][i] = 1;
119 }
120 printf("L = \n"); printMat(l, n, n);
121 printf("U = \n"); printMat(u, n, n);
122 for (int k = 0; k < n; k++) {
123     double sum = 0;
124     for (int i = 0; i < k; i++) {
125         sum += l[k][i] * y[i];
126     }
127     y[k] = (mat[k][n] - sum) / l[k][k];
128 }
129 printf("y = \n"); printVec(y, n);
130 for (int k = n - 1; k >= 0; k--) {
131     double sum = 0;
132     for (int i = k + 1; i < n; i++) {
133         sum += u[k][i] * x[i];
134     }
135     x[k] = (y[k] - sum) / u[k][k];
136 }
137 printf("x = \n"); printVec(x, n);
138 }
139
140 void JacobiMethod(double x[], int iter) {
141     double newx[N];
142     for (int iter_num = 1; iter_num <= iter; iter_num++) {
143         for (int i = 0; i < n; i++) {
144             double sum = 0;
145             for (int j = 0; j < n; j++) {
146                 if (j == i) continue;
147                 sum += mat[i][j] * x[j];
148             }
149             newx[i] = (mat[i][n] - sum) / mat[i][i];
150         }
151         printf("Iter %d: \n", iter_num);
152         for (int i = 0; i < n; i++) {

```

```

153         x[i] = newx[i];
154     }
155     printVec(x, n);
156 }
157 }
158
159 void GaussSeidelMethod(double x[], int iter) {
160     double newx[N];
161     for (int iter_num = 1; iter_num <= iter; iter_num++) {
162         for (int i = 0; i < n; i++) {
163             double sum = 0;
164             for (int j = 0; j < i; j++) {
165                 sum += mat[i][j] * newx[j];
166             }
167             for (int j = i + 1; j < n; j++) {
168                 sum += mat[i][j] * x[j];
169             }
170             newx[i] = (mat[i][n] - sum) / mat[i][i];
171         }
172         printf("Iter %d: \n", iter_num);
173         for (int i = 0; i < n; i++) {
174             x[i] = newx[i];
175         }
176         printVec(x, n);
177     }
178 }
179 }now;
180
181 int main(void) {
182     now.read();
183     now.GaussianElimination(true, "none");
184     // now.CroutSplit();
185     // double x[N] = {0, 0, 0};
186     // now.JacobiMethod(x, 20)
187     // now.GaussSeidelMethod(x, 20);
188     return 0;
189 }

```
