

## 실행 방법

- make clean && make && ./runHw03 <데이터 파일 경로>
- 특이값 행렬의 경우, main.c 최상단에서 #define SKIP\_GAUSSJ 추가
  - 특이값 행렬에서 gaussj 오류로 인해 프로그램 종지를 막기 위한 macro

```
C main.c M X
hw03 > C main.c > print_mat(float **, int, int, int, int)
You, 50 seconds ago | 1 author (You)
1  #include <stdio.h>
2  #include <stdlib.h>
3  #include <string.h>
4
5  #include "nr.h"
6  #include "nrutil.h"
7
8  // #define SKIP_GAUSSJ
9
```

## 실행 결과

- 각 파일별로 Gauss-Jordan, LU Decomposition, 특이값분해, Iterative Improvement로 해를 구함
- 각 파일별로 계수행렬의 역행렬과 특이값을 구함

## 파일 1 (lineq1.dat)

```
File: lineq_dat/lineq1.dat
Numerical Recipes run-time error...
gaussj: Singular Matrix
...now exiting to system...
```

- 행렬값이 0인 특이행렬이므로 실행되지 않음.
- Mac M2 Pro 환경에서 NR 폴더의 gaussj.c를 그대로 사용 시, singular matrix가 감지되지 않아서 프로그램이 중단되지 않고 아래와 같은 무의미한 해가 도출됨

```
File: lineq_dat/lineq1.dat
Method: Gauss-Jordan, Solution: [2.000000, -1.000000, -1.000000, -1.000000]
Method: LU Decomposition, Solution: [3.000000, 1.000001, -4.000001, -2.000000]
Method: SVD, Solution: [1.733333, -1.533333, -0.200000, -0.733333]
Method: Improved LU Decomposition, Solution: [3.000000, 1.000001, -4.000001, -2.000000]
Method: Improved LU Decomposition (noisy), Solution: [3.000008, 1.131539, -3.244396, -1.541350]
Method: Improved LU Decomposition (improve), Solution: [24.541349, 44.082710, -68.624054, -23.541349]
Inverse Matrix:
-33554432.000000    -12582912.000000    12582911.000000    4194304.500000
-67108888.000000    -25165830.000000    25165832.000000    8388610.000000
100663312.000000    37748740.000000    -37748740.000000    -12582915.000000
33554432.000000    12582912.000000    -12582911.000000    -4194304.500000
Determinant: -0.000000
```

- 아래와 같이 특이값 판정 조건문 수정하여 Mac M2 Pro 에서도 특이값 감지가 가능함

```
C main.c C gaussj.c M X lineq1.dat
NRs > ansi > recipes > C gaussj.c > gaussj(float **, int, float **, int)
12 void gaussj(float **a, int n, float **b, int m) {
33     | | | | }
34     | | | | ++(ipiv[icol]);
35     | | | | if (irow != icol) {
36     | | | | | for (l = 1; l <= n; l++) SWAP(a[irow][l], a[icol][l])
37     | | | | | for (l = 1; l <= m; l++) SWAP(b[irow][l], b[icol][l])
38     | | | | }
39     | | | | indxr[i] = irow;
40     | | | | indxc[i] = icol;
41     | | | | // 수정 부분
42     | | | | // if (a[icol][icol] == 0.0) nrerror("gaussj: Singular Matrix");
43     | | | | if (fabs(a[icol][icol]) < 1e-7) nrerror("gaussj: Singular Matrix");
44     | | | | pivinv = 1.0 / a[icol][icol];
45     | | | | a[icol][icol] = 1.0; You, 3 minutes ago * code done
46     | | | | for (l = 1; l <= n; l++) a[icol][l] *= pivinv;
47     | | | | for (l = 1; l <= m; l++) b[icol][l] *= pivinv;
48     | | | | for (ll = 1; ll <= n; ll++)
```

## 파일 2 (lineq2.dat)

```
File: lineq_dat/lineq3.dat
Method: Gauss-Jordan, Solution: [-0.326608, 1.532292, -1.044825, -1.587448, 2.928480, -2.218931]
Method: LU Decomposition, Solution: [-0.326608, 1.532292, -1.044826, -1.587447, 2.928480, -2.218930]
Method: SVD, Solution: [-0.326608, 1.532290, -1.044823, -1.587447, 2.928478, -2.218929]
Method: Improved LU Decomposition, Solution: [-0.326608, 1.532292, -1.044826, -1.587447, 2.928480, -2.218930]
Method: Improved LU Decomposition (noisy), Solution: [-0.326600, 1.663830, -0.289220, -1.128797, 3.461247, -1.999971]
Method: Improved LU Decomposition (mprove), Solution: [-0.326608, 1.532292, -1.044826, -1.587447, 2.928480, -2.218930]
Inverse Matrix:
-0.162205      0.122801      0.024068      -0.016431      -0.022840      0.046132
0.169407      -0.041117      0.228313      -0.087624      0.180306      -0.395655
-0.011636      0.122745      -0.117407      -0.180981      0.015910      0.186766
0.105669      -0.051726      -0.108916      0.299774      0.000859      -0.190541
-0.053026      -0.042361      0.160508      -0.224034      0.161811      0.015024
-0.062341      -0.064694      -0.234216      0.351126      -0.364828      0.434633
Determinant: 16178.401367
```

## 파일 3 (lineq3.dat)

```
File: lineq_dat/lineq2.dat
Method: Gauss-Jordan, Solution: [-2.873565, -0.612356, 0.976277, 0.635818, -0.553441]
Method: LU Decomposition, Solution: [-2.873566, -0.612357, 0.976277, 0.635819, -0.553441]
Method: SVD, Solution: [-2.873564, -0.612357, 0.976277, 0.635818, -0.553440]
Method: Improved LU Decomposition, Solution: [-2.873566, -0.612357, 0.976277, 0.635819, -0.553441]
Method: Improved LU Decomposition (noisy), Solution: [-2.873558, -0.480819, 1.731883, 1.094469, -0.020674]
Method: Improved LU Decomposition (mprove), Solution: [-2.873566, -0.612357, 0.976277, 0.635818, -0.553441]
Inverse Matrix:
0.354536      0.766945      0.207769      -0.595412      0.253128
0.035454      0.126695      0.195777      -0.159541      0.050313
-0.138686      -0.098540      -0.096715      0.124088      0.016423
-0.052138      -0.303962      -0.023201      0.234619      -0.044578
0.149114      0.459333      0.051356      -0.171011      0.042492
Determinant: 3835.999512
```