FACULTAD DE CIENCIAS

Tarea 3 Análisis Númerico

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Noviembre 2018

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1. Problemas de Computadora

Una nota importante es que al inicio de CADA script se incluyen los algoritmos, porfavor cambia la primera linea de cada script para que el path sea el correcto, porfavor.

Esta linea:

getd ('/Users/mac/Documents/Projects/Learning/UNAM/NumericalAnalysis/Homework2/Code/Algorithms')

Para ejecutar cada uno basta con hacer algo como:

exec("/Users/mac/Documents/Projects/Learning/UNAM/NumericalAnalysis/Homework2/Code/22a.sce", -1)

1.1. 23

Ejecuta los scripts que esta dentro de Code llamado: 23.sce

En este código muestra justo lo que se nos pide, por el método de ecuaciones normales veremos la solución aproximada para cada una de las $2\ b$ que nos dan y vemos a aunque son muy parecidas los resultados son completamente diferentes, y si, aunque puede parecer que o el algoritmo esta mal o algo raro acaba de pasar basta con dar un vistazo a la matriz A y verla como lo que la estamos interpretando, un conjunto de vectores.

Recuerda que lo que estamos haciendo es encontrar una combinación lineal de estos vectores para aproximar a nuestros vectores b.

Pues si nuestros vectores no son ortogonales lo que pasa es que un pequeño cambio en el resultado nos da una combinación lineal muy diferente. Es justo lo que esta pasando estamos trabajando con vectores muy parecidos.

La mejor forma de explicar este fenomeno y mostrar porque la ortogonalidad es tan importante es ver este video que justo habla de porque $1, x, x^2, \ldots$ es una base horrible para los polinomios. El link es https://youtu.be/pYoGYQOXqTk

Esto es un pdf y no me puedo quejar del espacio que utilizo, así que dejo el código:

```
    @Author: Alarcón Alvarez Aylin Yadira Guadalupe
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 3
                 = [ \\ 0.16 & 0.10; \\ 0.17 & 0.11; \\ 2.02 & 1.29; \\
11
\frac{14}{15}
                  0.28
17
19
20
22
23
                   0.25;
\frac{24}{25}
26
27
28
29
30
\frac{31}{32}
         disp("b:")
disp(b23)
33
         x23 = LeastSquares(A23, b23)
36
37
         disp("x:")
disp(x23)
38
39
         disp("Ax:")
disp(A23 * x23)
40
41
         disp("b_2:")
disp(b2_23)
42
44
         x2\_23 = LeastSquares(A23, b2\_23)
disp("x_2:")
disp(x2_23)
45
47
48
```

1.2. 24

```
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           \begin{array}{l} x24 = [\\ 1.02;\\ 0.95;\\ 0.87;\\ 0.77;\\ 0.67;\\ 0.56;\\ 0.44; \end{array}
 11
 12
 14
 15
 16
                       0.30;
 17
 19
 20
 21
22
23
                       = [0.39;
                       0.32;
0.27;
0.22;
24
25
 26
                       0.22;
0.18;
0.15;
0.13;
0.12;
0.13;
0.15;
27
28
30
31
33
34
           A24 = eye(10, 5)

b24 = x24
36
37
                       \begin{array}{l} i = (1 : 10) \\ A24(i , 1) = y24(i) * y24(i) \\ A24(i , 2) = x24(i) * y24(i) \\ A24(i , 3) = x24(i) \\ A24(i , 4) = y24(i) \\ A24(i , 5) = 1 \end{array}
 38
39
 40
 41
 42
 43
 44
45
\frac{47}{48}
49
50
51
52
53
55
56
 58
           disp("b:")
disp(b24)
59
\frac{61}{62}
           estimedx24 = LeastSquares(A24, b24)
\frac{64}{65}
 66
67
 69
70
71
72
73
74
75
76
77
78
79
            function [x] = solveEquation(a, b, c) 
 x = (-1 * b + sqrt(b*b - 4*a*c)) / (2 * a) endfunction
            function [y] = solve(coefficients, x)
a = coefficients(1)
b = coefficients(2)
c = coefficients(3)
d = coefficients(4)
e = coefficients(5)
 81
 83
 84
                        reala = (a)realb = (b * x + d)
```

```
realc = (c * x + e - x * x)

y = solveEquation(reala, realb, realc)
endfunction

someX = linspace(0, 1.10, 50)
someY = someX

for i = (1 : 50)
someY(i) = solve(estimedx24, someX(i))
end

plot(someX, someY, 'r-')
scatter(x24, y24)
```

1.3. 25

```
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    @Author: Laurrabaquio Rodríguez Miguel Salvador
    @Author: Pahua Castro Jesús Miguel Angel

10
11
13
14
16
17
19
21
22
23
24
25
26
27
28
                 disp("A")
disp(A25)
29
30
                 disp("b")
disp(b25)
32
33
34
35
36
                 38
                 disp("Real Solution: A x")
disp(A25 * realX25)
39
40
41
43
44
                 disp("Estimated Condition")
disp( Condition(A25, 10) )
\frac{46}{47}
49
50
```

1.4. 26

```
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    @Author: Pahua Castro Jesús Miguel Ángel

             \begin{array}{l} digitsAltered = eye\left(12\,,\ 1\right)\\ digitsClassic = eye\left(12\,,\ 1\right)\\ digitsTwo = eye\left(12\,,\ 1\right)\\ data = eye\left(12\,,\ 1\right) \end{array}
10
13
              for i = (2 : 12)

H = eye(i, i)

data(i) = i
16
18
19
\frac{20}{21}
\frac{23}{24}
                             \begin{array}{lll} \left[ \ Q26\, Classic \ , \ R26 \right] &= \ Gram Schmidt (H, \ 0) \\ \left[ \ Q26\, Altered \ , \ R26 \right] &= \ Gram Schmidt (H, \ 1) \\ \left[ \ Q26\, Two \ , \ R26 \right] &= \ Gram Schmidt ( \ Q26\, Classic \ , \ 0) \end{array} 
26
27
29
                            \begin{array}{lll} \mbox{digitsClassic(i)} &= -\log 10 \left( \mbox{norm(eye(i, i)} - \mbox{Q26Classic'} * \mbox{Q26Classic)} \right) \\ \mbox{digitsAltered(i)} &= -\log 10 \left( \mbox{norm(eye(i, i)} - \mbox{Q26Altered'} * \mbox{Q26Altered)} \right) \\ \mbox{digitsTwo(i)} &= -\log 10 \left( \mbox{norm(eye(i, i)} - \mbox{Q26Two'} * \mbox{Q26Two)} \right) \end{array}
30
31
32
33
                            disp("Q of GramSchmidt Classical:") disp(Q26Classic)
34
35
                            disp("Digits:")
disp(digitsClassic)
37
38
39
                            disp("Q of GramSchmidt Altered:")
disp(Q26Altered)
40
41
42
                            disp("Digits:")
disp(digitsAltered)
43
45
46
              plot(data, digitsClassic , '.b-')
plot(data, digitsAltered , '.r-')
plot(data, digitsTwo , '.g-')
48
49
```

```
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                   \begin{array}{l} {\rm digitsAltered} = {\rm eye}\,(12\,,\,\,1) \\ {\rm digitsClassic} = {\rm eye}\,(12\,,\,\,1) \\ {\rm digitsTwo} = {\rm eye}\,(12\,,\,\,1) \\ {\rm digitsHouse} = {\rm eye}\,(12\,,\,\,1) \\ {\rm data} = {\rm eye}\,(12\,,\,\,1) \end{array}
11
13
                    for i = (2 : 12)

H = eye(i, i)

data(i) = i
16
17
18
19
21
23
24
25
                                          \begin{array}{lll} \left[ \ Q26 \ Classic \ , & R26 \right] &= \ Gram \ Schmidt (H, \ 0) \\ \left[ \ Q26 \ Altered \ , & R26 \right] &= \ Gram \ Schmidt (H, \ 1) \\ \left[ \ Q26 \ Two, \ R26 \right] &= \ Gram \ Schmidt (Q26 \ Classic \ , \ 0) \\ \left[ \ Q26 \ H, \ R26 \right] &= \ House \ Holder (H) \end{array} 
27
29
30
```

```
\begin{array}{lll} \mbox{digitsClassic(i)} &= -\log 10 \left( \mbox{norm(eye(i, i)} - \mbox{Q26Classic'} * \mbox{Q26Classic'} \right) \\ \mbox{digitsAltered(i)} &= -\log 10 \left( \mbox{norm(eye(i, i)} - \mbox{Q26Altered'} * \mbox{Q26Altered} \right) \\ \mbox{digitsTwo(i)} &= -\log 10 \left( \mbox{norm(eye(i, i)} - \mbox{Q26Two'} * \mbox{Q26Two)} \right) \\ \mbox{digitsHouse(i)} &= -\log 10 \left( \mbox{norm(eye(i, i)} - \mbox{Q26H'} * \mbox{Q26H} \right) \right) \end{array}
\frac{33}{34}
 36
                            \begin{array}{ll} disp \mbox{ ("Q of GramSchmidt Classical:")} \\ disp \mbox{ (Q26Classic)} \end{array}
 37
 38
 39
                            disp("Digits:")
disp(digitsClassic)
 40
 41
 42
                            44
 45
                            disp("Digits:")
disp(digitsAltered)
 47
 49
             plot (data, digitsClassic, '.b-')
plot (data, digitsAltered, '.r-')
plot (data, digitsTwo, '.g-')
plot (data, digitsHouse, '.black-')
 50
 53
 56
```

```
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                  \begin{array}{l} {\rm digitsAltered} \ = \ {\rm eye} \, (12 \, , \, \, 1) \\ {\rm digitsClassic} \ = \ {\rm eye} \, (12 \, , \, \, 1) \\ {\rm digitsTwo} \ = \ {\rm eye} \, (12 \, , \, \, 1) \\ {\rm digitsNormal} \ = \ {\rm eye} \, (12 \, , \, \, 1) \\ {\rm data} \ = \ {\rm eye} \, (12 \, , \, \, 1) \end{array}
10
15
                   for i = (2 : 12)

H26 = eye(i, i)

data(i) = i
16
18
19
                                      20
21
23
24
26
                                       \begin{array}{lll} \left[ \begin{array}{lll} Q26\, Classic\;,\;\; R26 \end{array} \right] &=\; GramSchmidt\left( H26\;,\;\;0 \right) \\ \left[ \begin{array}{lll} Q26\, Altered\;,\;\; R26 \end{array} \right] &=\; GramSchmidt\left( H26\;,\;\;1 \right) \\ \left[ \begin{array}{lll} Q26\, Two\;,\;\; R26 \right] &=\; GramSchmidt\left( \begin{array}{lll} Q26\, Classic\;,\;\;0 \right) \\ \left[ \begin{array}{lll} L26 \end{array} \right] &=\; CholeskyGaussian\left( H26\; '\; *\; H26 \right) \\ Q26N &=\; H26\; *\; L26\; ' \end{array} 
27
29
30
31
32
                                      \begin{array}{lll} \mbox{digitsClassic(i)} &= -log10 \, (norm(eye\,(i\,,\,i\,)\, -\, Q26 Classic\,'\, *\, Q26 Classic\,)) \\ \mbox{digitsAltered(i)} &= -log10 \, (norm(eye\,(i\,,\,i\,)\, -\, Q26 Altered\,'\, *\, Q26 Altered\,)) \\ \mbox{digitsTwo(i)} &= -log10 \, (norm(eye\,(i\,,\,i\,)\, -\, Q26 Two\,'\, *\, Q26 Two\,)) \\ \mbox{digitsNormal(i)} &= -log10 \, (norm(eye\,(i\,,\,i\,)\, -\, Q26 N\,'\, *\, Q26 N\,)) \end{array}
33
34
35
36
37
38
                                       disp (Q26Classic)
39
40
                                       \begin{array}{ll} disp \, (\, \mbox{\tt "Q} \  \, of \  \, GramSchmidt \  \, Altered:\, \mbox{\tt "}\, ) \\ disp \, (\, Q26Altered\,) \end{array}
41
43
45
                  plot(data, digitsClassic, '.b-')
plot(data, digitsAltered, '.r-')
plot(data, digitsTwo, '.g-')
plot(data, digitsNormal, '.black-')
46
48
49
51
52
```

1.5. 27

Este proceso lo haremos hasta $\epsilon = 2^{-10}$ porque ϵ más pequeñas simplemente ya hacer que se trate a la matriz como singular pues se ve como una matriz con 3 columnas iguales, por lo tanto causa errores al tomarse como una matriz singular.

Lo que podemos ver que mientras epsilon más y más pequeña nos acercamos al claro resultado donde $x_1=x_2=x_3=0.333$

Algo curioso y diferente de los métodos anteriores, del parcial pasado es que practicamente todos los métodos dan el mismo resultado, curiosamente GramSchmidt es el único diferente en ciertos ϵ .

```
    @Author: Alarcón Alvarez Aylin Yadira Guadalupe
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             values = [
11
12
15
17
18
                       2e - 05;
2e - 08;
19
20
22
23
24
25
26
28
29
30
31
32
                                = [
1 1 1;
e 0 0;
0 e 0
33
34
36
37
39
40
                        ь27
42
43
44
45
46
                        disp("Ax = b")
47
48
50
51
52
                                   x27LeastSquares = LeastSquares(A27, b27)
53
54
55
56
57
58
59
60
                                    \begin{array}{lll} \hbox{$\left[\textrm{Q27HH, R27HH}\right]$} &=& \hbox{$H$ouseHolder} \left(\textrm{A27}\right) \\ \hbox{$x27$HouseHolder} &=& \hbox{$QRDecomposition} \left(\textrm{Q27HH, R27HH, b27}\right) \end{array} 
                                    \begin{array}{lll} \hbox{\tt [Q27GS, R27GS]} &=& \hbox{\tt GramSchmidt}(A27,\ 0) \\ \hbox{\tt x27GramSchmidt} &=& \hbox{\tt QRDecomposition}(Q27GS,\ R27GS,\ b27) \\ \end{array} 
                                    \begin{array}{ll} \left[ \text{Q27G, R27G} \right] &=& \text{Givens} \left( \text{A27} \right) \\ \text{x27Givens} &=& \text{QRDecomposition} \left( \text{Q27G, R27G, b27} \right) \end{array} 
61
                                   disp("x of Least Squares:")
disp(x27LeastSquares)
```

```
disp("Ax of Least Squares:")
disp(A27 * x27LeastSquares)

disp("x of HouseHolder:")
disp(x27HouseHolder)

disp("Ax of HouseHolder:")
disp(A27 * x27HouseHolder)

disp(A27 * x27HouseHolder)

disp(X27 * x27HouseHolder)

disp(x27GramSchmidt:")
disp(x27GramSchmidt:")

disp(x27GramSchmidt)

disp(A27 * x27GramSchmidt)

disp(A27 * x27GramSchmidt)

disp(X27GramSchmidt)

disp(A27 * x27GramSchmidt)

disp(x27Givens)

disp(x27Givens)

disp(x27Givens)

disp(x27Givens)

disp(A27 * x27Givens)

end

end
```

2. Anexo

2.1. Givens

```
Get the estimated solution to Ax = b using the HouseHolder transformation @param: A a matriz in M \{m \times n\} where m > n @return: Q a matriz in \overline{M} \{m \times m\} that is ortogonal @return: R a matriz in M \{m \times n\} that is triangular superior
  2 3
   5
6
7
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@Author: Laurrabaquio Rodríguez Miguel Salvador
@Author: Pahua Castro Jesús Miguel Angel
   9
               \begin{array}{l} {\rm function} \ [Q,\ R] = {\rm Givens}\,(A) \\ [m,\ n] = {\rm size}\,(A)\,; \\ Q = {\rm eye}\,(m,\ m)\,; \\ R = A; \end{array}
 11
 12
\frac{14}{15}
                              \begin{array}{lll} for & j = (1:n) \\ & for & i = (m:-1:j+1) \\ & & GivenMatrix = eye(m,m); \\ & [c,s] = GivensRotations(R(i-1,j),R(i,j)); \\ & & GivenMatrix([i-1,i],[i-1,i]) = [c-s;sc]; \end{array}
17
18
20
21
                                                           \begin{array}{ll} R \, = \, Given Matrix \, , & R \, ; \\ Q \, = \, Q \, * \, Given Matrix \, ; \end{array}
22
23
24
25
26
27
28
29
30
31
32
33
34
35
36
37
38
39
40
42
43
```

2 ANEXO 2.2 GramSchmidt

2.2. GramSchmidt

```
Get the estimated solution to Ax = b using the Gramm-Schmmidt transformation @param: A a matriz in M_{m \times n} where m > n @param: option if 1 then is the alterate else is the classic algorithm @return: Q a matriz in M_{m \times n} that is ortogonal @return: R a matriz in M_{m \times n} that is triangular superior
   4
   5

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 10
 11
                   \begin{array}{lll} \text{function} & \left[ \, \mathbf{Q}, \; \, \mathbf{R} \, \right] \; = \; \mathbf{GramSchmidt}(\mathbf{A}, \; \; \mathbf{option} \, ) \\ & \left[ \, \mathbf{m}, \; \, \mathbf{n} \, \right] \; = \; \mathbf{size} \left( \mathbf{A} \right) \, ; \end{array} 
 13
 14
                                \begin{array}{ll} Q \,=\, z\, e\, r\, o\, s\, \left(\, m\,,\  \  \, n\,\right)\,; \\ R \,=\, A \end{array}
16
 17
 19
21
22
23
24
25
                                                   \begin{array}{cccc} for & i = (1 : j - 1) \\ & if & (option == 1) \\ & & R(i , j) = Q(:, i) \, `* \, Q(:, j); \end{array}
26
27
28
29
30
31
32
33
35
36
```

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2 Anexo 2.3 HouseHolder

2.3. HouseHolder

```
Get the estimated solution to Ax = b using the HouseHolder transformation @param: A a matriz in M \{m \times n\} where m > n @return: Q a matriz in M \{m \times m\} that is ortogonal @return: R a matriz in M \{m \times n\} that is triangular superior
  4
 5

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10
             \begin{array}{ll} function \ [Q,\ R] = HouseHolder(A,\ b) \\ [m,\ n] = size(A)\,; \\ Q = eye(m,\ m)\,; \end{array} 
11
13
14
16
17
                                    \begin{array}{lll} alpha & = -sign\left( A(\,i\,\,,\,i\,\,) \,\right) & * & norm\left( A(\,i\,\,:\,\,m,\,\,\,i\,\,) \,\right); \\ aei\,(\,1) & = & alpha \end{array}
19
                                    \begin{array}{l} v = A(i : m, \ i) - aei; \\ House Holder = eye(m - (i-1), \ m-(i-1)) \ -2 * ((v * v') \ / \ (v' * v)); \end{array}
21
23
24
                                     \begin{array}{lll} Real House Holder & = & \mathrm{eye}\left(m, & m\right); \\ Real House Holder\left(i \; : \; m, \; \; i \; : \; m\right) & = \; House Holder; \\ \end{array} 
26
27
                                    egin{array}{ll} A &=& RealHouseHolder * A; \ Q &=& Q * RealHouseHolder; \end{array}
29
30
```

2.4. LeastSquares

```
// Get the estimated solution to Ax = b using least squares
// @param: A a matriz in M_{m x n} where m > n
// @param: b a vector of m rows
// @return: x a estimated solution
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// @Author: Laurrabaquio Rodríguez Miguel Salvador
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function [x] = LeastSquares(A, b)
[L] = CholeskyBanachiewicz(A' * A, 1);

y = FowardSubstitution(L, A' * b);
x = BackwardSubstitution(L', y);
endfunction
```

2.5. BackwardSubstitution

```
// Solve a system Ux = y where U is an upper triangular
// using the famous algorithm backward substitution
// @param: U triangular superior matrix
// @param: b the b in Ux = b
// @return: x the solution vector

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function [x] = BackwardSubstitution(U, b)
[m, n] = size(U);
x = zeros(n, 1);

for i = (n: -1: 1)
    if (U(i, i) == 0)
        error('Error: Singular matrix');
    return;
    end

x(i) = b(i) / U(i,i);

for j = (1: i - 1)
        b(j) = b(j) - U(j, i) * x(i);
end
end
endfunction
```

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2.6. CholeskyBanachiewicz

```
Factor A as A = L * L^T using the famous algorithm called Cholesky using this really awesome propierty First A = L U then we make U a unit upper triangular matrix so we have L D L' and then we do L D2 D2 L' were D2(i, j) = \operatorname{sqrt}(D(i, j)) finally we associate and we have A = L2 * L2' where L2 = L * D2 @param: A a positive defined matrix (so A is symetric) @param: option if 1 then A = L * L else A = L * D * L @return: L lower triangule matrix
   4

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 10
 13
 14
               \begin{array}{ll} function\left[L,\ D\right] = CholeskyBanachiewicz(A,\ option)\\ \left[m,\ n\right] = size\left(A\right);\\ D = eye\left(n,\ n\right);\\ L = eye\left(m,\ n\right);\\ U = A; \end{array}
16
 17
 19
21
 23
 24
                                          \begin{array}{l} for \ row = (step \ + \ 1 \ : \ n) \\ L(row, \ step) = U(row, \ step) \ / \ U(step, \ step); \\ for \ column = (1 \ : \ n) \\ U(row, \ column) = U(row, \ column) \ - \ L(row, \ step) \ * \ U(step, \ column); \end{array}
 27
29
30
32
33
                             \begin{array}{l} \text{if option} == 1 \\ \text{for step} = (1 : n) \\ \text{for row} = (\text{step} : n) \\ \text{L(row, step)} = \text{L(row, step)} * \text{sqrt(U(step, step))}; \end{array}
35
36
38
39
 40
41
                                            \begin{array}{c} \texttt{for} \\ \texttt{for} \\ \texttt{D}(\texttt{step} \,,\,\, \texttt{step}) = \texttt{U}(\texttt{step} \,,\,\, \texttt{step}); \end{array} 
 43
 44
 46
 49
```

2.7. CholeskyGaussian

```
Factor A as A=L*L^T using the famous algorithm called Cholesky using a modification of Gaussian Elimination @param: A a positive defined matrix (so A is symetric) @return: L lower triangule matrix
  4

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 10
 11
             function [L] = CholeskyGaussian (A)
                         \begin{array}{l} [\, m, \;\; n \,] \; = \; \text{size} \, (A) \; ; \\ L \; = \; \text{zeros} \, (m, \;\; n) \; ; \end{array} \label{eq:Laplace}
 13
 14
                        \begin{array}{ll} for & step \, = \, (1 \, : \, n) \\ & A(\,step \, , \, \, step \, ) \, = \, sqrt \, ( \, \, A(\,step \, , \, \, \, step \, ) \, \, ) \, ; \end{array} \label{eq:formula}
 16
 17
                                      \begin{array}{ll} for \ column \, = \, (\, step \, + \, 1 \, : \, n \,) \\ A \ (\, column \, , \ step \,) \, = \, A(\, column \, , \ step \,) \, \, / \, \, A(\, step \, , \ step \,) \, ; \end{array} 
 19
21
22
23
\frac{24}{25}
26
27
28
                       for row = (1 : n)

for column = (1 : n)

if (row >= column)

L(row, column) = A(row, column);
29
30
32
33
35
36
```

2.8. CompleteLUDecomposition

```
Factor A as PAQ = LU @param: A a not singular matrix @return: L (not sure) lower triangule matrix @return: U upper triangule matrix @return: P permutation matrix @return: Q permutation matrix
    4

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 10
 13
                  \begin{array}{lll} function & [L,\ U,\ P,\ Q] = CompleteLUDecomposition(A) \\ & [m,\ n] = \operatorname{size}(A); \\ & \mathrm{if} & (m\ ^- n) = 0 \text{ then} \\ & & \mathrm{error}(\, '\operatorname{Error}\colon \operatorname{Not} \ \operatorname{square} \ \operatorname{matrix}\, ')\,; \\ & end \\ & P = \operatorname{eye}(n,\ n)\,; \\ & Q = \operatorname{eye}(n,\ n)\,; \\ & L = \operatorname{eye}(n,\ n)\,; \\ & U = A\,; \end{array} 
 14
 16
 17
 19
21
 24
                                 \begin{array}{l} \text{for step} \, = (1 \, : \, n - \, 1) \\ \text{Qi} \, = \, \text{eye} \, (n \, , \, \, n) \, ; \\ \text{Pi} \, = \, \text{eye} \, (n \, , \, \, n) \, ; \end{array}
27
29
30
                                                 32
33
                                                 if (maxIndex == 0)
    error('Error: Singular matrix');
end
35
36
                                                 \begin{array}{l} temporal \, = \, Pi\,(\,step\,\,,\,\,\,:)\,\,; \\ Pi\,(\,step\,\,,\,\,\,:) \, = \, Pi\,(\,index\,(\,1\,)\,\,,\,\,:)\,\,; \\ Pi\,(\,index\,(\,1\,)\,\,,\,\,:) \, = \, temporal\,; \end{array}
38
39
 40
                                                 \begin{array}{l} temporal \, = \, Qi\,(:\,,\,\,step\,)\,\,; \\ Qi\,(:\,,\,\,step\,) \, = \, Qi\,(:\,,\,\,index\,(2)\,)\,\,; \\ Qi\,(:\,,\,\,index\,(2)\,) \, = \, temporal\,; \end{array}
41
43
 44
                                                 \begin{array}{l} temporal \ = \ U(\,step\,\,,\ :)\,\,;\\ U(\,step\,\,,\ :) \ = \ U(\,index\,(1)\,\,,\ :)\,\,;\\ U(\,index\,(1)\,\,,\ :) \ = \ temporal\,; \end{array}
 46
 47
                                                 \begin{array}{l} temporal \, = \, U(:\,,\ step\,)\,; \\ U(:\,,\ step\,) \, = \, U(:\,,\ index\,(2)\,)\,; \\ U(:\,,\ index\,(2)\,) \, = \, temporal\,; \end{array}
 49
51
52
                                                 \begin{array}{l} for \ row = (step \, + \, 1 \, : \, n) \\ L(row, \ step) = U(row, \ step) \, / \, U(step \, , \ step); \\ for \ column = \, (1 \, : \, n) \\ U(row, \ column) = \, U(row, \ column) \, - \, L(row, \ step) \, * \, U(step \, , \ column); \end{array}
54
55
57
58
 60
                                                Q = Q * Qi;

P = Pi * P;
 61
63
64
```

2.9. FowardSubstitution

```
// Solve a system Ly = b where L is triangular inferior
// using the famous algorithm foward substitution
// @param: L triangular inferior matrix
// @param: b the b in Ly = b
// @return: x the solution vector

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function [y] = FowardSubstitution(L, b)
[m, n] = size(L);
y = zeros(n, 1);

for i = (1 : n)
    if (L(i, i) == 0)
        error('Error: Singular matrix');
    return;
end

y(i) = b(i) / L(i, i);

for j = (i + 1 : n)
        b(j) = b(j) - L(j, i) * y(i);
end
end
endfunction
```

2.10. LUDecomposition

```
Factor A as A=L*U @param: A a not singular matrix @return: L lower triangule matrix @return: U upper triangule matrix
  4

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10
           \begin{array}{ll} function \ [L,\ U] = LUDecomposition (A) \\ [m,\ n] = size (A); \\ L = eye (m,\ n); \\ U = A; \end{array} 
11
13
14
16
17
19
21
22
23
24
25
                             \begin{array}{ll} for \ row = (step \, + \, 1 \, : \, n) \\ L(row \, , \ step \, ) \, = \, U(row \, , \ step \, ) \, \, / \, \, U(step \, , \ step \, ) \, ; \end{array}
26
27
28
                                        29
30
```