

## 신호처리를 위한 행렬 계산 Programming Assignment 3

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### 1. Code implementation

#### a) $LDL^T$ Factorization.

LDL decomposition에서 L과 D의 각 요소의 공식은 아래와 같다.

##### LDL decomposition [edit]

An alternative form, eliminating the need to take square roots when  $\mathbf{A}$  is symmetric, is the symmetric indefinite factorization<sup>[17]</sup>

$$\mathbf{A} = \mathbf{LDL}^T = \begin{pmatrix} 1 & 0 & 0 \\ L_{21} & 1 & 0 \\ L_{31} & L_{32} & 1 \end{pmatrix} \begin{pmatrix} D_1 & 0 & 0 \\ 0 & D_2 & 0 \\ 0 & 0 & D_3 \end{pmatrix} \begin{pmatrix} 1 & L_{21} & L_{31} \\ 0 & 1 & L_{32} \\ 0 & 0 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} D_1 & & \\ L_{21}D_1 & L_{21}^2D_1 + D_2 & \\ L_{31}D_1 & L_{31}L_{21}D_1 + L_{32}D_2 & L_{31}^2D_1 + L_{32}^2D_2 + D_3 \end{pmatrix} \quad (\text{symmetric}).$$

The following recursive relations apply for the entries of  $\mathbf{D}$  and  $\mathbf{L}$ :

$$D_j = A_{jj} - \sum_{k=1}^{j-1} L_{jk}^2 D_k,$$

$$L_{ij} = \frac{1}{D_j} \left( A_{ij} - \sum_{k=1}^{j-1} L_{ik} L_{jk} D_k \right) \quad \text{for } i > j.$$

This works as long as the generated diagonal elements in  $\mathbf{D}$  stay non-zero. The decomposition is then **unique**.  $\mathbf{D}$  and  $\mathbf{L}$  are real if  $\mathbf{A}$  is real.

이를 for문을 활용해 구현해주면 다음과 같은 코드로 완성할 수 있다.

```
n = size(x,1);
L = eye(n); % L is unit lower triangular
D = eye(n);
D(1,1) = x(1,1);
% get L(i,j) for i>j
for i=2:n
    for j=1:i-1
        ① L(i,j) = x(i,j);
        ② ( for k=1:j-1
            L(i,j) = L(i,j) - L(i,k)*L(j,k)*D(k,k);
        end
        ③ L(i,j) = L(i,j)/D(j,j);
    end
    % get Diagonal entry for D
    ④ D(i,i) = x(i,i);
    ⑤ ( for j=1:i-1
        D(i,i) = D(i,i) - L(i,j)^2*D(j,j);
    end
end
```

공식에 해당하는 코드의 부분을 숫자로 표기하였다.

b) *Cholesky Factorization*( $LL^T$  Factorization).

Cholesky decomposition에서 L의 각 요소의 공식은 아래와 같다.

### The Cholesky–Banachiewicz and Cholesky–Crout algorithms

If we write out the equation

$$\begin{aligned} \mathbf{A} = \mathbf{L}\mathbf{L}^T &= \begin{pmatrix} L_{11} & 0 & 0 \\ L_{21} & L_{22} & 0 \\ L_{31} & L_{32} & L_{33} \end{pmatrix} \begin{pmatrix} L_{11} & L_{21} & L_{31} \\ 0 & L_{22} & L_{32} \\ 0 & 0 & L_{33} \end{pmatrix} \\ &= \begin{pmatrix} L_{11}^2 & & \\ L_{21}L_{11} & L_{22}^2 + L_{21}^2 & \\ L_{31}L_{11} & L_{31}L_{21} + L_{32}L_{22} & L_{31}^2 + L_{32}^2 + L_{33}^2 \end{pmatrix}, \end{aligned}$$

we obtain the following:

$$\mathbf{L} = \begin{pmatrix} \sqrt{A_{11}} & 0 & 0 \\ A_{21}/L_{11} & \sqrt{A_{22} - L_{21}^2} & 0 \\ A_{31}/L_{11} & (A_{32} - L_{31}L_{21})/L_{22} & \sqrt{A_{33} - L_{31}^2 - L_{32}^2} \end{pmatrix}$$

and therefore the following formulas for the entries of L:

$$\begin{aligned} L_{j,j} &= (\pm) \sqrt{A_{j,j} - \sum_{k=1}^{j-1} L_{j,k}^2}, \\ L_{i,j} &= \frac{1}{L_{j,j}} \left( A_{i,j} - \sum_{k=1}^{j-1} L_{i,k} L_{j,k} \right) \quad \text{for } i > j. \end{aligned}$$

이를 코드로 구현하면 아래와 같다.

```
n = size(x,1);
L = zeros(n);
L(1,1) = sqrt(x(1,1)); % initialize
%get L(i,j) for i>j
for i=2:n
    for j=1:i-1
        ① L(i,j) = x(i,j);
        ② ( for k=1:j-1
            L(i,j) = L(i,j) - L(i,k)*L(j,k);
        end
        ③ L(i,j) = L(i,j)/L(j,j);
    end
    % get the diagonal entry L(i,i)
    ④ L(i,i) = x(i,i);
    ⑤ ( for j=1:i-1
        L(i,i) = L(i,i) - L(i,j)^2;
    end
    ⑥ L(i,i) = sqrt(L(i,i));
end
```

마찬가지로 공식에 해당하는 코드의 부분을 숫자로 표시하였다.

c) *LU decomposition with partial pivoting.*

```

n = size(A,1);
Ak = A;
L = eye(n); % L is unit lower triangular
U = zeros(n);
P = eye(n);
for k = 1:n-1
    [~,r] = max(abs(Ak(k:end,k))); % r = the index of the maximum of a part of column
    r = n-(n-k+1)+r; % consider the relative index
    Ak([k r],:) = Ak([r k],:); % exchange the row
    P([k r],:) = P([r k],:);
    L([k r],:) = L([r k],:);
    for i = k+1:n
        L(i,k) = Ak(i,k) / Ak(k,k); % get tau
        for j = 1:n
            U(k,j) = Ak(k,j);
            Ak(i,j) = Ak(i,j) - L(i,k)*Ak(k,j); % subtraction of the rows below the pivot
        end
    end
end
U(:,end) = Ak(:,end);

```

K를 증가시켜가며 pivot의 column의 값을 decreasing order로 정렬하기 위해 행들의 위치를 서로 바꿔준다. 이 행의 교환을 Ak, P, L에 모두 적용해주어야 된다. 그 후 pivot의 값을 이용해 Gaussian transformation을 하듯이  $\tau_k$ 의 값들을 구해 이를 pivot의 row에 곱해 각 row에서 빼주면 그 행이 0으로 맞춰지고,  $\tau_k$ 는 L의 요소가 된다.

## 2. Test.m Screenshot.

1. LDL factorization

```

L:
    1.0000    0    0    0
    0.4749    1.0000    0    0
    0.7865    0.5143    1.0000    0
    0.6894    0.8821    0.9700    1.0000

D:
    1.7177    0    0    0
    0    0.7643    0    0
    0    0    0.4969    0
    0    0    0    0.0057

L * D * L':
    1.7177    0.8158    1.3509    1.1841
    0.8158    1.1518    1.0347    1.2366
    1.3509    1.0347    1.7615    1.7600
    1.1841    1.2366    1.7600    1.8842

x:
    1.7177    0.8159    1.3509    1.1841
    0.8158    1.1518    1.0347    1.2366
    1.3509    1.0347    1.7615    1.7600
    1.1841    1.2366    1.7600    1.8842

```

2.

Cholesky factorization

L:

1.3106	0	0	0
0.6225	0.8743	0	0
1.0307	0.4496	0.7049	0
0.9035	0.7712	0.6838	0.0753

L \* L':

1.7177	0.8158	1.3509	1.1841
0.8158	1.1518	1.0347	1.2366
1.3509	1.0347	1.7615	1.7600
1.1841	1.2366	1.7600	1.8842

x:

1.7177	0.8159	1.3509	1.1841
0.8158	1.1518	1.0347	1.2366
1.3509	1.0347	1.7615	1.7600
1.1841	1.2366	1.7600	1.8842

3.

LU factorization

L:

1.0000	0	0	0
0.4749	1.0000	0	0
0.7865	0.5142	1.0000	0
0.6894	0.8821	0.9700	1.0000

U:

1.7177	0.8159	1.3509	1.1841
0	0.7643	0.3931	0.6742
0	-0.0000	0.4969	0.4820
0	0	0	0.0057

L \* U:

1.7177	0.8159	1.3509	1.1841
0.8158	1.1518	1.0347	1.2366
1.3509	1.0347	1.7615	1.7600
1.1841	1.2366	1.7600	1.8842

P \* x:

1.7177	0.8159	1.3509	1.1841
0.8158	1.1518	1.0347	1.2366
1.3509	1.0347	1.7615	1.7600
1.1841	1.2366	1.7600	1.8842

### 3. Uniqueness of $LDL^T, LL^T$

#### 1) $LL^T$

Cholesky Decomposition is unique up to the sign of its diagonal entries.

Pf) Assume  $A = LL^T = MM^T$  are two Cholesky decompositions of A.

$$I = L^{-1}MM^TL^{-T} = (L^{-1}M)(L^{-1}M)^T$$

$L^{-1}M$  is a lower triangular matrix and thus  $(L^{-1}M)^T$  is an upper triangular matrix.

$$D := L^{-1}M = (L^{-1}M)^T$$

$D$  is both lower triangular and upper triangular. Therefore,  $D$  is diagonal.

$$D^2 = I \rightarrow D = \text{diag}(\pm 1, \pm 1, \dots, \pm 1)$$

$$M = LD$$

$L$  and  $D$  differ by the sign of their columns.

Therefore, if we restrict the diagonal entries of  $L$  to be positive,  $L$  is unique.

#### 2) $LDL^T$

From the lecture, we know the following.

**$LDL^T$  Factorization**

If  $A \in \mathbb{R}^{n \times n}$  is **symmetric**,  
 and  $\det(A(1:k, 1:k)) \neq 0$ , for  $k = 1:n-1$   
 then there exist a **unique unit lower triangular** matrix  
 $L$ , and a **unique diagonal** matrix  $D = \text{diag}\{d_1, \dots, d_n\}$   
 such that  $A = LDL^T$ .

We can prove that  $D$  is unique.

suppose

$$A = L_1 D_1 L_1^* = L_2 D_2 L_2^*$$

where  $D_1, D_2$  are diagonal matrices and  $L_1, L_2$  are lower triangular with 1's on the diagonal.

$$D_2 = L_2^{-1} L_1 D_1 L_1^* (L_2^*)^{-1} = (L_2^{-1} L_1) D_1 (L_2^{-1} L_1)^*$$

$$= R^* D_1 R$$

with  $R := (L_2^{-1} L_1)^*$  which is upper triangular with ones on the diagonal

$$\implies D_2 R^{-1} = R^* D_1$$

the LHS is upper triangular while the RHS is lower triangular, which in fact implies each side is diagonal. And since  $R$  has ones on the diagonal:

$$\implies D_2 = D_2 R^{-1} = R^* D_1 = D_1$$

alternatively, looking at  $D_2 R^{-1} = R^* D_1$  and examining component (k,k) of each side

$$\implies d_{k,k}^{(2)} \cdot 1 = d_{k,k}^{(1)} \cdot 1 \implies D_2 = D_1$$