

Lecture 23:Cholesky Factorization

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1 Hermitian Positive Definite Matrix

Definition 1.1. A complex matrix $A \in C^{m \times m}$ is hermitian if $A^* = A$ ($\bar{A}^T = A$ or $a_{ij} = \bar{a}_{ji}$). A is said to be hermitian positive definite if $x^*Ax > 0$ for all $x \neq 0$.

Remark:

- A is hermitian positive definite if and only if it's eigenvalues are all positive.
- If A is hermitian positive definite and $A = LU$ is the LU decomposition of A then $u_{11} > 0, u_{22} > 0, \dots, u_{mm} > 0$.

This can be proved by the following steps.

1. A is hermitian positive definite then $\det(A_k) > 0$ for $k = 1, \dots, n$ where A_k is the $k \times k$ diagonal matrix of A , i.e. $A_k = (a_{ij})_{1 \leq i \leq k, 1 \leq j \leq k}$. If we write $L = \begin{bmatrix} L_k & 0 \\ P & Q \end{bmatrix}$ and $U = \begin{bmatrix} U_k & R \\ 0 & S \end{bmatrix}$ where $L_k = (l_{ij})_{1 \leq i \leq k, 1 \leq j \leq k}$ and $U_k = (u_{ij})_{1 \leq i \leq k, 1 \leq j \leq k}$ then $A_k = L_k U_k$. Note that L_k is unit lower-triangular and $\det(L_k) = 1$. Similarly, U_k is upper-triangular and $\det(U_k) = u_{11}u_{22} \dots u_{kk}$. Therefore $\det(A_k) = \det(L_k)\det(U_k) = u_{11}u_{22} \dots u_{kk} > 0$ for $1 \leq k \leq m$. So $u_{11} > 0, u_{11}u_{22} > 0, \dots, u_{11}u_{22} \dots u_{mm} > 0$. This implies that $u_{11} > 0, u_{22} > 0, \dots, u_{mm} > 0$.

2 Cholesky Factorization

Definition 2.2. A complex matrix $A \in C^{m \times m}$ is has a Cholesky factorization if $A = R^*R$ where R is a upper-triangular matrix

Theorem 2.3. Every hermitian positive definite matrix A has a unique Cholesky factorization.

Proof: From the remark of previous section, we know that $A = LU$ where L is unit lower-triangular and U is upper-triangular with $u_{11} > 0, u_{22} > 0, \dots, u_{mm} > 0$. First, we factor U as

$$U = \begin{bmatrix} u_{11} & u_{12} & \dots & u_{1m-1} & u_{1m} \\ 0 & u_{22} & \dots & u_{2m-1} & u_{2m} \\ 0 & 0 & \dots & \dots & \dots \\ 0 & 0 & \dots & u_{m-1m-1} & u_{m-1m} \\ 0 & 0 & \dots & 0 & u_{mm} \end{bmatrix}$$

$$= \begin{bmatrix} u_{11} & 0 & \cdots & 0 & 0 \\ 0 & u_{22} & \cdots & 0 & 0 \\ 0 & 0 & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & u_{m-1m-1} & 0 \\ 0 & 0 & \cdots & 0 & u_{mm} \end{bmatrix} \begin{bmatrix} 1 & \frac{u_{12}}{u_{11}} & \cdots & \frac{u_{1m-1}}{u_{11}} & \frac{u_{1m}}{u_{11}} \\ 0 & 1 & \cdots & \frac{u_{2m-1}}{u_{22}} & \frac{u_{2m}}{u_{22}} \\ 0 & 0 & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & 1 & \frac{u_{m-1m}}{u_{m-1m-1}} \\ 0 & 0 & \cdots & 0 & 1 \end{bmatrix} = \Lambda W.$$

Since $u_{11} > 0, \dots, u_{mm} > 0$, we can write $\Lambda = D^2$

$$\text{where } D = \begin{bmatrix} \sqrt{u_{11}} & 0 & \cdots & 0 & 0 \\ 0 & \sqrt{u_{22}} & \cdots & 0 & 0 \\ 0 & 0 & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & \sqrt{u_{m-1m-1}} & 0 \\ 0 & 0 & \cdots & 0 & \sqrt{u_{mm}} \end{bmatrix}.$$

So we have $A = LU = L\Lambda W = LD^2W$ where L is unit-triangular and W is unit upper-triangular. Since $A^* = A$, we have $LD^2W = (LD^2W)^* = W^*(D^2)^*L^* = W^*(D^2)L^*$. Note that W^* is unit lower-triangular. By the uniqueness of LU factorization, we have $L = W^*$. So $A = LDDL^* = (LD)(LD)^*$. Let $R = DL^*$. Then R is upper-triangular and $A = R^*R$.

Lemma 2.4. Suppose A^*A is invertible. Then $A^*A = R^*R$ where R is upper-triangular.

Proof: One can check easily that A^*A is hermitian (b/c $(A^*A)^* = A^*(A^*)^* = A^*A$). Since $x^*(A^*A)x = (Ax)^*(Ax) = \|Ax\|^2 \geq 0$ and $x^*(A^*A)x = 0$ if $Ax = 0$. Note that A^*A is invertible. So $Ax = 0$ implies $A^*Ax = 0$ and $x = 0$.

Algorithm for Cholesky Factorization for a Hermitian positive definite matrix

Step1. Find a LU decomposition of $A = LU$.

Step2. Factor $U = D^2W$ where W is a unit upper-triangular matrix and D is a diagonal matrix.

Step3. $A = R^*R$ where $R = DW$.

Example 2.5. Determine if the following matrix is hermitian positive definite. Also find its Cholesky factorization if possible.

$$A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 3 & 3 \\ 1 & 3 & 2 \end{bmatrix}, B = \begin{bmatrix} 1 & 2 & 2 \\ 2 & 8 & 0 \\ 2 & 0 & 24 \end{bmatrix}.$$

Solution:

(1) From the row reduction, we have the following.

$$A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 3 & 3 \\ 1 & 3 & 2 \end{bmatrix} \xrightarrow{-2r_1 + r_2, -r_1 + r_3} \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & 1 \\ 0 & 1 & 1 \end{bmatrix} \xrightarrow{1 \cdot r_2 + r_3} \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & 1 \\ 0 & 0 & 2 \end{bmatrix}.$$

$$\text{So } A = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & 1 \\ 0 & 0 & 2 \end{bmatrix}$$

Now $u_{22} = -1 < 0$. So A is not positive definite.

(2) From the row reduction, we have the following.

$$B = \begin{bmatrix} 1 & 2 & 2 \\ 2 & 8 & 0 \\ 2 & 0 & 24 \end{bmatrix} \xrightarrow{-2r_1 + r_2, -2r_1 + r_3} \begin{bmatrix} 1 & 2 & 2 \\ 0 & 4 & -4 \\ 0 & -4 & 20 \end{bmatrix} \xrightarrow{1 \cdot r_2 + r_3} \begin{bmatrix} 1 & 2 & 2 \\ 0 & 4 & -4 \\ 0 & 0 & 16 \end{bmatrix}.$$

$$\begin{aligned} \text{So } B &= \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 2 \\ 0 & 4 & -4 \\ 0 & 0 & 16 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 16 \end{bmatrix} \begin{bmatrix} 1 & 2 & 2 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{bmatrix}^2 \begin{bmatrix} 1 & 2 & 2 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix}. \text{ Since } u_{11} = 1 > 0, u_{22} = 4 > 0 \text{ and } \\ u_{33} &= 16 > 0, \text{ we know that } B \text{ is positive definite.} \end{aligned}$$

$$\text{So } R = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} 1 & 2 & 2 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 2 \\ 0 & 2 & -2 \\ 0 & 0 & 4 \end{bmatrix} \text{ and } B = R^*R$$

3 Least square via Cholesky factorization

Recall that the solution of the least square problem $Ax = b$ is the solution to $A^*Ax = A^*b$. Assume that A has full rank. Then A^*A is hermitian and positive definite. Then $A^*A = R^*R$ (the Cholesky factorization of A^*A) where R is upper-triangular. Then $A^*Ax = A^*b \iff R^* \underbrace{Rx}_y = A^*b$

$\iff R^*y = A^*b$ and $Rx = y$.

Algorithm: Least Squares via Cholesky factorization

1. Compute A^*A
2. Find the Cholesky factorization of $A^*A = R^*R$.
2. Solve the lower-triangular system $R^*y = A^*b$
3. Solve the upper-triangular system $Rx = y$ for x .

Example 3.6. Use Cholesky factorization to find the solution to the least square

$$\text{problem } \begin{bmatrix} 1 & -1 & 1 \\ -1 & 1 & 1 \\ 0 & 1 & -1 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \\ 1 \\ -2 \end{bmatrix}$$

Solution:

$$1. \text{ Compute } A^*A = \begin{bmatrix} 1 & -1 & 0 & 0 \\ -1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \end{bmatrix} \begin{bmatrix} 1 & -1 & 1 \\ -1 & 1 & 1 \\ 0 & 1 & -1 \\ 0 & 1 & -1 \end{bmatrix} = \begin{bmatrix} 2 & -2 & 0 \\ -2 & 4 & -2 \\ 0 & -2 & 4 \end{bmatrix}.$$

2. From the row reduction, we have the following.

$$A^*A = \begin{bmatrix} 2 & -2 & 0 \\ -2 & 4 & -2 \\ 0 & -2 & 4 \end{bmatrix} \xrightarrow{r_1 + r_2} \begin{bmatrix} 2 & -2 & 0 \\ 0 & 2 & -2 \\ 0 & -2 & 4 \end{bmatrix} \xrightarrow{1 \cdot r_2 + r_3} \begin{bmatrix} 2 & -2 & 0 \\ 0 & 2 & -2 \\ 0 & 0 & 2 \end{bmatrix}.$$

$$\text{So } A^*A = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} 2 & -2 & 0 \\ 0 & 2 & -2 \\ 0 & 0 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 0 & 0 \\ 0 & \sqrt{2} & 0 \\ 0 & 0 & \sqrt{2} \end{bmatrix}^2 \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\text{So } R = \begin{bmatrix} \sqrt{2} & 0 & 0 \\ 0 & \sqrt{2} & 0 \\ 0 & 0 & \sqrt{2} \end{bmatrix} \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} \sqrt{2} & -\sqrt{2} & 0 \\ 0 & \sqrt{2} & -\sqrt{2} \\ 0 & 0 & \sqrt{2} \end{bmatrix} \text{ and}$$

$$A^*A = R^*R.$$

$$\text{Compute } A^* \begin{bmatrix} 2 \\ 1 \\ 1 \\ -2 \end{bmatrix} = \begin{bmatrix} 1 & -1 & 0 & 0 \\ -1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ 1 \\ -2 \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \\ 4 \end{bmatrix}.$$

$$\text{Now solve } R^*y = A^* \begin{bmatrix} 2 \\ 1 \\ 1 \\ -2 \end{bmatrix}$$

$$\Leftrightarrow \begin{bmatrix} \sqrt{2} & 0 & 0 \\ -\sqrt{2} & \sqrt{2} & 0 \\ 0 & -\sqrt{2} & \sqrt{2} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \\ 4 \end{bmatrix}$$

$$\Leftrightarrow y_1 = \frac{1}{\sqrt{2}}, y_2 = \frac{-2 + \sqrt{2}y_1}{\sqrt{2}} = -\frac{1}{\sqrt{2}} \text{ and } y_3 = \frac{4 + \sqrt{2}y_2}{\sqrt{2}} = \frac{3}{\sqrt{2}}$$

Last, we solve $Rx = y$

$$\Leftrightarrow \begin{bmatrix} \sqrt{2} & -\sqrt{2} & 0 \\ 0 & \sqrt{2} & -\sqrt{2} \\ 0 & 0 & \sqrt{2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \\ \frac{3}{\sqrt{2}} \end{bmatrix}$$

$$\Leftrightarrow x_3 = \frac{3}{2}, x_2 = \frac{-\frac{1}{\sqrt{2}} + \sqrt{2}x_3}{\sqrt{2}} = 1 \text{ and } x_1 = \frac{\frac{1}{\sqrt{2}} + \sqrt{2}x_2}{\sqrt{2}} = \frac{3}{2}.$$

$$\text{Hence the solution is } x = \begin{bmatrix} \frac{3}{2} \\ 1 \\ \frac{3}{2} \end{bmatrix}.$$

Homework 11: Due April 4

1. Determine if the following matrix is hermitian positive definite. Also find its Cholesky factorization if possible.

$$A = \begin{bmatrix} 1 & -2 & 1 \\ -2 & 8 & -14 \\ 1 & -14 & 28 \end{bmatrix}, B = \begin{bmatrix} 1 & -2 & 1 \\ -2 & 8 & -14 \\ 1 & -14 & 46 \end{bmatrix}.$$

b. Use Cholesky factorization to find the solution to the least square problem

$$\begin{bmatrix} -1 & 1 & 1 \\ 1 & -1 & 1 \\ 0 & 1 & -1 \\ 0 & 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ -1 \end{bmatrix}.$$