

به نام خدا

پاسخ تمرین سری سوم

-۱

(b) i.

$$\begin{aligned} A &= \begin{pmatrix} 2 & 4 & 6 \\ 4 & 11 & 15 \\ 6 & 15 & 23 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \begin{pmatrix} 2 & 4 & 6 \end{pmatrix} + \begin{pmatrix} 0 & 0 & 0 \\ 0 & 3 & 3 \\ 0 & 3 & 5 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \begin{pmatrix} 2 & 4 & 6 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 3 & 3 \end{pmatrix} + \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 2 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \begin{pmatrix} 2 & 4 & 6 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 3 & 3 \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 0 & 2 \end{pmatrix} \\ &= \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 1 & 1 \end{pmatrix}}_L \underbrace{\begin{pmatrix} 2 & 4 & 6 \\ 0 & 3 & 3 \\ 0 & 0 & 2 \end{pmatrix}}_U \end{aligned}$$

ii. To find $A = LDL^T$, we will try to write $U = DL^T$ by pulling out a row scaling matrix D

from U to leave something unit upper triangular:

$$\begin{aligned}
 A &= \begin{pmatrix} 2 & 4 & 6 \\ 4 & 11 & 15 \\ 6 & 15 & 23 \end{pmatrix} \\
 &= \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 1 & 1 \end{pmatrix}}_L \underbrace{\begin{pmatrix} 2 & 4 & 6 \\ 0 & 3 & 3 \\ 0 & 0 & 2 \end{pmatrix}}_U \\
 &= \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 1 & 1 \end{pmatrix}}_L \underbrace{\begin{pmatrix} 2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 2 \end{pmatrix}}_D \underbrace{\begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}}_{L^T}
 \end{aligned}$$

iii. To get the Cholesky factorization of A from the factorization $A = LDL^T$, we do the following:

$$\begin{aligned}
 A &= LDL^T \\
 &= LD^{\frac{1}{2}}D^{\frac{1}{2}}L^T \\
 &= (LD^{\frac{1}{2}})(D^{\frac{1}{2}}L^T) \\
 &= R^T R.
 \end{aligned}$$

For the matrix above, this process gives

$$\begin{aligned}
 A &= \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 1 & 1 \end{pmatrix} \begin{pmatrix} 2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 2 \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \\
 &= \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ 3 & 1 & 1 \end{pmatrix} \begin{pmatrix} \sqrt{2} & 0 & 0 \\ 0 & \sqrt{3} & 0 \\ 0 & 0 & \sqrt{2} \end{pmatrix} \begin{pmatrix} \sqrt{2} & 0 & 0 \\ 0 & \sqrt{3} & 0 \\ 0 & 0 & \sqrt{2} \end{pmatrix} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \\
 &= \underbrace{\begin{pmatrix} \sqrt{2} & 0 & 0 \\ 2\sqrt{2} & \sqrt{3} & 0 \\ 3\sqrt{2} & \sqrt{3} & \sqrt{2} \end{pmatrix}}_{R^T} \underbrace{\begin{pmatrix} \sqrt{2} & 2\sqrt{2} & 3\sqrt{2} \\ 0 & \sqrt{3} & \sqrt{3} \\ 0 & 0 & \sqrt{2} \end{pmatrix}}_R
 \end{aligned}$$

Answer

Using material from the worked example in the notes we set

$$\begin{bmatrix} 3 & 1 & 6 \\ -6 & 0 & -16 \\ 0 & 8 & -17 \end{bmatrix} = \begin{bmatrix} U_{11} & U_{12} & U_{13} \\ L_{21}U_{11} & L_{21}U_{12} + U_{22} & L_{21}U_{13} + U_{23} \\ L_{31}U_{11} & L_{31}U_{12} + L_{32}U_{22} & L_{31}U_{13} + L_{32}U_{23} + U_{33} \end{bmatrix}$$

and comparing elements row by row we see that

$$\begin{array}{lll} U_{11} = 3, & U_{12} = 1, & U_{13} = 6, \\ L_{21} = -2, & U_{22} = 2, & U_{23} = -4 \\ L_{31} = 0 & L_{32} = 4 & U_{33} = -1 \end{array}$$

and it follows that

$$\begin{bmatrix} 3 & 1 & 6 \\ -6 & 0 & -16 \\ 0 & 8 & -17 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 4 & 1 \end{bmatrix} \begin{bmatrix} 3 & 1 & 6 \\ 0 & 2 & -4 \\ 0 & 0 & -1 \end{bmatrix}$$

is an LU decomposition of the given matrix.

We found earlier that the coefficient matrix is equal to $LU = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 4 & 1 \end{bmatrix} \begin{bmatrix} 3 & 1 & 6 \\ 0 & 2 & -4 \\ 0 & 0 & -1 \end{bmatrix}$.

First we solve $LY = B$ for Y , we have

$$\begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 4 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 4 \\ 17 \end{bmatrix}.$$

The top line implies that $y_1 = 0$. The middle line states that $-2y_1 + y_2 = 4$ and therefore $y_2 = 4$. The last line tells us that $4y_2 + y_3 = 17$ and therefore $y_3 = 1$.

Finally we solve $UX = Y$ for X , we have

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

The bottom line shows that $x_3 = -1$. The middle line then shows that $x_2 = 0$, and then the top line gives us that $x_1 = 2$. The required solution is $X = \begin{bmatrix} 2 \\ 0 \\ -1 \end{bmatrix}$.

$$\begin{aligned}
 \begin{bmatrix} a_1 & a_2 & a_3 \end{bmatrix} &= \begin{bmatrix} -1 & -1 & 1 \\ 1 & 3 & 3 \\ -1 & -1 & 5 \\ 1 & 3 & 7 \end{bmatrix} \\
 &= \begin{bmatrix} q_1 & q_2 & q_3 \end{bmatrix} \begin{bmatrix} R_{11} & R_{12} & R_{13} \\ 0 & R_{22} & R_{23} \\ 0 & 0 & R_{33} \end{bmatrix}
 \end{aligned}$$

First column of Q and R

$$\tilde{q}_1 = a_1 = \begin{bmatrix} -1 \\ 1 \\ -1 \\ 1 \end{bmatrix}, \quad R_{11} = \|\tilde{q}_1\| = 2, \quad q_1 = \frac{1}{R_{11}}\tilde{q}_1 = \begin{bmatrix} -1/2 \\ 1/2 \\ -1/2 \\ 1/2 \end{bmatrix}$$

Second column of Q and R

- compute $R_{12} = q_1^T a_2 = 4$
- compute

$$\tilde{q}_2 = a_2 - R_{12}q_1 = \begin{bmatrix} -1 \\ 3 \\ -1 \\ 3 \end{bmatrix} - 4 \begin{bmatrix} -1/2 \\ 1/2 \\ -1/2 \\ 1/2 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

- normalize to get

$$R_{22} = \|\tilde{q}_2\| = 2, \quad q_2 = \frac{1}{R_{22}}\tilde{q}_2 = \begin{bmatrix} 1/2 \\ 1/2 \\ 1/2 \\ 1/2 \end{bmatrix}$$

Third column of Q and R

- compute $R_{13} = q_1^T a_3 = 2$ and $R_{23} = q_2^T a_3 = 8$
- compute

$$\tilde{q}_3 = a_3 - R_{13}q_1 - R_{23}q_2 = \begin{bmatrix} 1 \\ 3 \\ 5 \\ 7 \end{bmatrix} - 2 \begin{bmatrix} -1/2 \\ 1/2 \\ -1/2 \\ 1/2 \end{bmatrix} - 8 \begin{bmatrix} 1/2 \\ 1/2 \\ 1/2 \\ 1/2 \end{bmatrix} = \begin{bmatrix} -2 \\ -2 \\ 2 \\ 2 \end{bmatrix}$$

- normalize to get

$$R_{33} = \|\tilde{q}_3\| = 4, \quad q_3 = \frac{1}{R_{33}}\tilde{q}_3 = \begin{bmatrix} -1/2 \\ -1/2 \\ 1/2 \\ 1/2 \end{bmatrix}$$

Final result

$$\begin{aligned} \begin{bmatrix} -1 & -1 & 1 \\ 1 & 3 & 3 \\ -1 & -1 & 5 \\ 1 & 3 & 7 \end{bmatrix} &= \begin{bmatrix} q_1 & q_2 & q_3 \end{bmatrix} \begin{bmatrix} R_{11} & R_{12} & R_{13} \\ 0 & R_{22} & R_{23} \\ 0 & 0 & R_{33} \end{bmatrix} \\ &= \begin{bmatrix} -1/2 & 1/2 & -1/2 \\ 1/2 & 1/2 & -1/2 \\ -1/2 & 1/2 & 1/2 \\ 1/2 & 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 2 & 4 & 2 \\ 0 & 2 & 8 \\ 0 & 0 & 4 \end{bmatrix} \end{aligned}$$

Application to linear regression

The QR method is often used to estimate [linear regressions](#).

In a linear regression we have an $N \times 1$ vector y of outputs and an $N \times K$ matrix of inputs whose columns are assumed to be linearly independent. We need to find the $K \times 1$ coefficient vector β that minimizes the mean squared errors made by using the fitted values

$$\hat{y} = X\beta$$

to predict the actual values y .

The well-known solution to this problem is the so-called [ordinary least squares](#) (OLS) estimator

$$\beta = (X^T X)^{-1} X^T y$$

We can simplify the formula for the OLS estimator, avoid to invert a matrix and thus reduce the computational burden (and the possible numerical instabilities) by computing the QR decomposition of X :

$$X = QR$$

where Q is $N \times K$ and R is $K \times K$.

Then, the OLS estimator becomes

$$\begin{aligned} \beta &= (X^T X)^{-1} X^T y \\ &= (R^T Q^T Q R)^{-1} R^T Q^T y \\ &= (R^T R)^{-1} R^T Q^T y \\ &= R^{-1} (R^T)^{-1} R^T Q^T y \\ &= R^{-1} Q^T y \end{aligned}$$

or

$$R\beta = Q^T y$$

The latter way of writing the solution is more convenient: since R is upper triangular, we do not need to invert it, but we can use the [back-substitution algorithm](#) to find the solution β .

$$\begin{aligned}
 A &= A_P \dots A_2 A_1 \quad \text{حيث } A_i \text{ مصفوفة قابلة للعكس} \\
 Q_0 &= I \\
 \tilde{A}_1 &\triangleq A_1 Q_0 \xrightarrow{QR} \tilde{A}_1 = Q_1 R_1 \rightarrow Q_1^T \tilde{A}_1 = R_1 \rightarrow Q_1^T A_1 Q_0 = R_1 \quad \text{المصفوفة الأولى} \\
 \tilde{A}_2 &\triangleq A_2 Q_1 \xrightarrow{QR} \tilde{A}_2 = Q_2 R_2 \rightarrow Q_2^T (A_2 Q_1) = R_2 \quad \text{المصفوفة الثانية} \\
 &\vdots \\
 \tilde{A}_P &\triangleq A_P Q_{P-1} \xrightarrow{QR} \tilde{A}_P = Q_P R_P \rightarrow Q_P^T (A_P Q_{P-1}) = R_P \quad \text{المصفوفة الأخيرة} \\
 A &= A_P A_{P-1} \dots A_2 A_1 = A_P \underbrace{Q_{P-1}^T Q_{P-1}}_I A_{P-1} \underbrace{Q_{P-2}^T Q_{P-2}}_I \dots \underbrace{Q_2^T Q_2}_I A_1 \underbrace{Q_1^T Q_1}_I Q_0 \xrightarrow{I} \\
 &= A_P Q_{P-1} R_{P-1} \dots R_2 R_1 \\
 \Rightarrow Q_P^T A &= Q_P^T A_P Q_{P-1} R_{P-1} \dots R_1 = R_P \dots R_1 \Rightarrow A = Q_P (R_P \dots R_1) \Rightarrow \begin{cases} A = QR \\ Q = Q_P \\ R = R_P \dots R_1 \end{cases}
 \end{aligned}$$

$$\begin{aligned}
 QR \text{ decomposition} \rightarrow A = QR &\Rightarrow |\det(A)| = |\det(Q)| |\det(R)| = |\tau_1| \dots |\tau_n| \leq \|a_1\|_2 \dots \|a_n\|_2 \quad (*) \\
 R &= [\tau_1, \dots, \tau_n]
 \end{aligned}$$

$$(*) \text{ حيث } \tau_i(i) \leq \|a_i\|_2 \text{ حيث } \|a_i\|_2 = \sqrt{a_{i1}^2 + \dots + a_{in}^2}$$

$$Q^T A = R \Rightarrow Q^T a_i = \tau_i \Rightarrow \|Q^T a_i\|_2 = \|\tau_i\|_2 \Rightarrow \|a_i\|_2 = \|\tau_i\|_2 \geq |\tau_i(i)| \quad \checkmark$$

اثبات این تعبیر به راحتی از رابطه ضرب ماتریس بلوکی روشن است.

الف - (۲) با تعریف دترمینان نشان دهید:

$$\det \begin{bmatrix} I_{m_1} & & \\ & \ddots & \\ & & A_j & \\ & & & \ddots \\ & & & & I_{m_k} \end{bmatrix} = \det(A_j)$$

S_j

اثبات به کمک رابطه بازگشت دترمینال روشن است. (با اندازش S_j اندیشه خنثی می شود)

$$\begin{aligned} \det(S_i) &= (-1)^{1+1} \times \det(S_i(2:\text{end}, 2:\text{end})) = \det(S_i(3:\text{end}, 3:\text{end})) = \dots = \det(S_i(m_1 + \dots + m_{i-1} + 1:\text{end}, m_1 + \dots + m_{i-1} + 1:\text{end})) \\ &= \det(S_i(m_1 + \dots + m_{i-1} + 1:\text{end} - 1, m_1 + \dots + m_{i-1} + 1:\text{end} - 1)) = \dots = \\ &= \det(S_i(m_1 + \dots + m_{i-1} + 1 + \text{end} - m_n - \dots - m_{i+1}, m_1 + \dots + m_{i-1} + 1 + \text{end} - m_n - \dots - m_{i+1})) \\ &= \det(S_i(\underbrace{\quad}_{A_i}, \underbrace{\quad}_{A_i})) = \det(A_i) \end{aligned}$$

الف - (۳) نشان دهید:

$$\det(A) = \det(A_1) \det(A_2) \dots \det(A_k)$$

طبیقت پس $\rightarrow \det(A) = \det(S_1) \dots \det(S_n) = \det(A_1) \dots \det(A_n)$

ب) فرض کنید $M = \begin{bmatrix} A & B \\ C & D \end{bmatrix}$ که $A \in \mathbb{R}^{p \times p}, B \in \mathbb{R}^{p \times q}, C \in \mathbb{R}^{q \times p}, D \in \mathbb{R}^{q \times q}$

ب - (۱) تجزیه های شبه LDU زیر را برای ماتریس بلوکی M ثابت کنید. یعنی اگر A معکوس پذیر باشد، نشان دهید:

$$M = \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I_p & 0 \\ CA^{-1} & I_q \end{bmatrix} \begin{bmatrix} A & 0 \\ 0 & D - CA^{-1}B \end{bmatrix} \begin{bmatrix} I_p & A^{-1}B \\ 0 & I_q \end{bmatrix}$$

و اگر D معکوس پذیر باشد، نشان دهید:

$$M = \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I_p & BD^{-1} \\ 0 & I_q \end{bmatrix} \begin{bmatrix} A - BD^{-1}C & 0 \\ 0 & D \end{bmatrix} \begin{bmatrix} I_p & 0 \\ D^{-1}C & I_q \end{bmatrix}$$

ب - (۲) نشان دهید اگر A و D معکوس پذیر باشند داریم:

$$\det \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \det(A) \det(D - CA^{-1}B) = \det(D) \det(A - BD^{-1}C)$$

ب - (۱) به راحتی هویت از ضرب بلوکی ماتریس روشن است.

ب-2) از طریق روابط نت (ب-1) دترمینان محاسب می‌کنیم. از آنجایی که دترمینان ماتریس $\begin{pmatrix} I_m & -A \\ B & I_n \end{pmatrix}$ معکوس ضرب در برابر دترمینان ماتریس $\begin{pmatrix} I_m & A \\ B & I_n \end{pmatrix}$ است، هر دو رابطه مستقیم بر دست می‌آید.

ج) فرض کنید $A \in \mathbb{R}^{m \times n}, B \in \mathbb{R}^{n \times m}, X \in \mathbb{R}^{m \times m}$ معکوس پذیر است.

ج-1) قضیه دترمینان Sylvester را اثبات کنید:

$$\det(I_m + AB) = \det(I_n + BA)$$

ج-2) فرم کلی تر قضیه ی بالا را اثبات کنید:

$$\det(X + AB) = \det(X) \det(I_n + BX^{-1}A)$$

ج-3) برای دو بردار $u, v \in \mathbb{R}^m$ با استفاده از قضیه ی بالا نشان دهید:

$$\det(X + uv^T) = (1 + v^T X^{-1}u) \det(X)$$

(1-2)

$$M = \begin{pmatrix} I_m & -A \\ B & I_n \end{pmatrix} \Rightarrow \det(M) \stackrel{\text{تغییر سطر}}{=} \det(I_m) \det(I_n + B I_m^{-1} A) = \det(I_n) \det(I_m + A I_n^{-1} B) \Rightarrow \square$$

(2-2)

$$\det(X + AB) = \det(X(I_m + X^{-1}AB)) = \det(X) \det(I_m + X^{-1}AB) \stackrel{(1-2)}{=} \det(X) \det(I_n + B X^{-1}A) \quad \square$$

(3-2)

$$(A \triangleq u \quad B \triangleq v^T) \rightarrow \text{چگونه دترمینان 2-2} \rightarrow$$