

4. Determinants and the Inverse of a Square Matrix

In this section, we are going to use our knowledge of determinants and their properties to derive an explicit formula for the inverse of a square matrix A provided that $|A| \neq 0$. Before we do this, we need one additional property of determinants which is a consequence of our Gaussian algorithm for computing the value of a determinant.

Lemma 8: Suppose the N by N matrix A has the following block upper triangular form:

$$A \equiv \begin{bmatrix} a, & b^T \\ 0_{N-1}, & C \end{bmatrix}$$

where a is a scalar, b is an $N-1$ dimensional column vector and C is an $N-1$ by $N-1$ matrix. Then the determinant of A is equal to a times the determinant of C ; i.e.,

$$(4) \quad |A| = |a| |C| = a |C|.$$

Proof: The Gaussian algorithm explained in the previous section can be used to calculate the determinant of A . For Stage 1 of the algorithm, we do not have to do anything: A is already in the form that is required at the end of Stage 1. Thus stages 2 to $N-1$ of the algorithm reduce the matrix C into an upper triangular matrix U say. Thus we have

$$(5) \quad |C| = |U| = \begin{vmatrix} u_{11} & u_{12} & \dots & u_{1,N-1} \\ 0 & u_{22} & \dots & u_{2,N-1} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & u_{N-1,N-1} \end{vmatrix} = u_{11}u_{22} \dots u_{N-1,N-1}.$$

At the same time, we have the following formula for $|A|$:

$$(6) \quad |A| = \begin{vmatrix} a, & b_1, & b_2, & \dots, & b_{N-1} \\ 0, & u_{11}, & u_{12}, & \dots, & u_{1,N-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0, & 0, & 0, & \dots, & u_{N-1,N-1} \end{vmatrix} = a u_{11} \dots u_{N-1,N-1} = a |C|$$

where the last equality in (6) follows from (5).

Q.E.D.

Corollary: Suppose the N by N matrix A has the following block, lower triangular form:

$$(7) \quad A \equiv \begin{bmatrix} a, & 0_{N-1}^T \\ b, & C \end{bmatrix} \quad \text{Then}$$

$$(8) \quad |A| = |a| \quad |C| = a|C|.$$

Proof: Let A be defined by (7). Then A^T has the form that is required to apply Lemma (8). Thus we have, using Lemma (2),

$$\begin{aligned} |A| &= |A^T| = a|C^T| && \text{using Lemma (8)} \\ &= a|C| && \text{using Lemma (2)} \end{aligned}$$

which is the desired result.

Q.E.D.

The next Lemma requires two definitions. Let A be a square N by N matrix.

Definition: $A(i,j)$ denotes the ij th *minor* of the matrix A and it is the determinant of the $N-1$ by $N-1$ submatrix of A which has deleted row i and column j .

Definition: $A_{ij} \equiv (-1)^{i+j} A(i, j)$ denotes the ij th *cofactor* of the matrix A ; it is equal to the ij th minor of A ; it is equal to the ij th minor of A times minus one raised to the power $i+j$.

Examples:
$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$A_{11} = (-1)^{1+1} a_{22} = a_{22}$$

$$A_{22} = (-1)^{2+2} a_{11} = a_{11}$$

$$A_{12} = (-1)^{1+2} a_{21} = -a_{21}$$

$$A_{21} = (-1)^{1+2} a_{12} = -a_{12};$$

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

$$A_{11} = (1)^{1+1} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix}; A_{12} = (1)^{1+2} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}; A_{13} = (1)^{1+3} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}; \text{etc.}$$

Lemma (9): *Expansion by Cofactors along the First Row:*

$$(9) \quad |A| = a_{11} A_{11} + a_{12} A_{12} + \dots + a_{1N} A_{1N}$$

Proof: Let $A = [a_{ij}]$ be an N by N matrix and let the N dimensional unit vectors be denoted by the columns

$$e_1 \equiv \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, e_2 \equiv \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, e_N \equiv \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

It can be seen that the first row of the matrix A can be written as $\sum_{j=1}^N a_{1j} e_j^T = A_{1\bullet}$. Thus we have

$$(10) \quad |A| = \begin{vmatrix} A_{1\bullet} \\ A_{2\bullet} \\ \vdots \\ A_{N\bullet} \end{vmatrix} = \begin{vmatrix} \sum_{j=1}^N a_{1j} e_j^T \\ A_{2\bullet} \\ \vdots \\ A_{N\bullet} \end{vmatrix} = \sum_{j=1}^N \begin{vmatrix} a_{1j} e_j^T \\ A_{2\bullet} \\ \vdots \\ A_{N\bullet} \end{vmatrix} \quad \text{making repeated use of Lemma (4).}$$

The first of the N determinants on the right hand side of (10) can be written as

$$(11) \quad \begin{vmatrix} a_{11} & 0 & \dots & 0 \\ \tilde{A}_{\bullet 1} & \tilde{A}_{\bullet 2} & \dots & \tilde{A}_{\bullet N} \end{vmatrix} = a_{11} |\tilde{A}_{\bullet 2}, \dots, \tilde{A}_{\bullet N}|$$

using the Corollary to Lemma (8)

$$= a_{11} A(1, 1)$$

$$= a_{11} A_{11}$$

where $\tilde{A}_{\bullet j}$ is the j th column of A after dropping the first row of A , for $j = 1, 2, \dots, N$.

The second of the N determinants on the right hand side of (10) can be written as

$$\begin{vmatrix} 0 & a_{12} & 0 & \dots & 0 \\ \tilde{A}_{\bullet 1} & \tilde{A}_{\bullet 2} & \tilde{A}_{\bullet 3} & \dots & \tilde{A}_{\bullet N} \end{vmatrix} = - \begin{vmatrix} a_{12} & 0 & 0 & \dots & 0 \\ \tilde{A}_{\bullet 2} & \tilde{A}_{\bullet 1} & \tilde{A}_{\bullet 3} & \dots & \tilde{A}_{\bullet N} \end{vmatrix}$$

where we have interchanged the first two columns of the first matrix in the second matrix and hence by Lemma (5) we must multiply by -1 to preserve the equality

$$= (-1) a_{12} |\tilde{A}_{\bullet 1}, \tilde{A}_{\bullet 3}, \dots, \tilde{A}_{\bullet N}| \quad \text{by the Corollary to Lemma (8)}$$

$$= (-1) a_{12} A(1,2) \quad \text{by the definition of the minor } A(1,2)$$

$$= (-1)^{2+1} a_{12} A(1,2) \quad \text{multiplying by } (-1)^2$$

$$= a_{12} A_{12} \quad \text{by the definition of the cofactor } A_{12}.$$

The third of the N determinants on the right hand side of (10) can be written as:

$$\begin{vmatrix} 0 & 0 & a_{13} & 0 & \dots & 0 \\ \tilde{A}_{\bullet 1} & \tilde{A}_{\bullet 2} & \tilde{A}_{\bullet 3} & \tilde{A}_{\bullet 4} & \dots & \tilde{A}_{\bullet N} \end{vmatrix} = (\square 1)^2 \begin{vmatrix} a_{13} & 0 & 0 & 0 & \dots & 0 \\ \tilde{A}_{\bullet 3} & \tilde{A}_{\bullet 1} & \tilde{A}_{\bullet 2} & \tilde{A}_{\bullet 4} & \dots & \tilde{A}_{\bullet N} \end{vmatrix}$$

where we have made 2 column interchanges to move the original third column first to column 2 and then to column 1

$$\begin{aligned} &= (-1)^2 a_{13} \begin{vmatrix} \tilde{A}_{\bullet 1} & \tilde{A}_{\bullet 2} & \tilde{A}_{\bullet 3} & \tilde{A}_{\bullet 4} & \tilde{A}_{\bullet 5} & \dots & \tilde{A}_{\bullet N} \end{vmatrix} && \text{by the Corollary to Lemma (8)} \\ &= a_{13} (-1)^2 A(1,3) && \text{by the definition of the minor } A(1,3) \\ &= a_{13} (-1)^{3+1} A(1,3) && \text{multiplying by } (-1)^2 \\ &= a_{13} A_{13} && \text{by the definition of the cofactor } A_{13}. \end{aligned}$$

Continuing on in the same way for the remaining $N-3$ terms on the right hand side of (10), we see that formula (9) results.

Q.E.D.

Lemma(10): *Expansion by Cofactors along the i th Row:*

$$(12) \quad |A| = a_{i1} A_{i1} + a_{i2} A_{i2} + \dots + a_{iN} A_{iN} \quad \text{for } i = 1, 2, \dots, N.$$

Proof: If $i = 1$, Lemma (10) reduces to Lemma (9). For $i > 1$, we make $i - 1$ row interchanges to move the i th row up to the first row and then we apply Lemma (9). Thus we have

$$|A| = \begin{vmatrix} A_{1\bullet} \\ \vdots \\ A_{i\bullet} \\ \vdots \\ A_{N\bullet} \end{vmatrix} = (\square 1)^{i\square 1} \begin{vmatrix} A_{i\bullet} \\ A_{1\bullet} \\ A_{2\bullet} \\ \vdots \\ A_{i\square 1\bullet} \\ A_{i+1\bullet} \\ \vdots \\ A_{N\bullet} \end{vmatrix} \quad \text{because there are } i - 1 \text{ row interchanges}$$

$$= (\square 1)^{i\square 1} \prod_{j=1}^{\square N} a_{ij} (\square 1)^{j+1} A(i, j) \quad \text{applying Lemma (9)}$$

$$= \prod_{j=1}^N a_{ij} A_{ij} \quad \text{using the definition of the cofactor } A_{ij}.$$

Q.E.D.

Corollary to Lemma (10): $a_{j1} A_{i1} + a_{j2} A_{i2} + \dots + a_{jN} A_{iN} = 0$ if $i \neq j$.

$$\text{Proof: Let } A = \begin{bmatrix} A_{1\bullet} \\ \vdots \\ A_{N\bullet} \end{bmatrix}$$

Suppose we replace the i th row of the matrix A with the j th row of A where $i \neq j$. Then by lemma (1), we have

$$0 = \begin{vmatrix} A_{1\bullet} \\ \vdots \\ A_{j\bullet} \\ \vdots \\ A_{j\bullet} \\ \vdots \\ A_{N\bullet} \end{vmatrix} \quad \text{ith row has been replaced by jth row.}$$

$= a_{j1} A_{i1} + a_{j2} A_{i2} + \dots + a_{jN} A_{iN}$ using lemma (10) except that the numbers $a_{i1}, a_{i2}, \dots, a_{iN}$ have been replaced by $a_{j1}, a_{j2}, \dots, a_{jN}$.

Q.E.D.

Lemma (10) and its corollary enable us to calculate a *right inverse* of an N by N matrix A provided that $|A| \neq 0$.

Lemma (11): If A is an N by N matrix and $|A| \neq 0$, then a right inverse for A , say A_R^{-1} , given by

$$A_R^{-1} = \begin{bmatrix} \frac{A_{11}}{|A|} & \dots & \frac{A_{N1}}{|A|} \\ \vdots & & \vdots \\ \frac{A_{1N}}{|A|} & \dots & \frac{A_{NN}}{|A|} \end{bmatrix} = \frac{1}{|A|} \begin{bmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \vdots & \vdots & & \vdots \\ A_{N1} & A_{N2} & \dots & A_{NN} \end{bmatrix}^T$$

where A_{ij} is the ij th cofactor of the matrix A .

Proof: Take the i th row of A , $A_{i\bullet}$, time the j th column of A_R^{-1} . We get

$$a_{i1} \frac{A_{j1}}{|A|} + a_{i2} \frac{A_{j2}}{|A|} + \dots + a_{iN} \frac{A_{jN}}{|A|} = \begin{cases} \frac{|A|}{|A|} & \text{if } i = j \text{ by lemma (10)} \\ 0 & \text{if } i \neq j \text{ by corollary to (10)} \end{cases}$$

Thus we have $A A_R^{-1} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix} = I_N$ and thus A_R^{-1} is a right inverse for A

by definition.

Q.E.D.

Problem 9: Calculate a right inverse for the following matrices:

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad \text{assuming } ad - bc \neq 0 \quad \text{and} \quad \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 3 & 4 \end{bmatrix}$$

Problem 10: Suppose that the N by N matrix A has a right inverse B and a left inverse C . Show that $B = C$. *Hint:* You will need the results of problem 2.

Problem 11: If A is an N by N matrix and $|A| \neq 0$, show that a *left* inverse exists. (Note: Problem *assumed* the existence of a left inverse; here we have to show that one exists. *Hint:* If $|A| \neq 0$, then $|A^T| \neq 0$. Thus A^T will have a right inverse by lemma (11)).

Problem 12: If A is an N by N matrix and $|A| = 0$, show that there does not exist an N by N matrix B such that $AB = I_N$. *Hint:* You may find lemma (6) useful.

The above results give us an easily checked condition for the existence of an inverse matrix for an N by N matrix A : namely if $|A| \neq 0$, then a common right and left inverse matrix exists which we will denote by A^{-1} ; If $|A| = 0$, then A^{-1} does not exist. There is another convenient condition on an N by N matrix A which will ensure that A^{-1} exists and we will develop it below after we discuss Cramer's Rule.

5. Expansion by Cofactors along a column and Cramer's Rule.

Applying lemma (10) to A^T yields the following equation:

$$(13) \quad |A^T| = a_{1i} A_{1i} + a_{2i} A_{2i} + \dots + a_{Ni} A_{Ni} \text{ for any } \textit{column} \text{ index } i = 1, \dots, N$$

(i.e., we have simply interchanged row and column indices)

$$= |A| \text{ since } |A| = |A^T|.$$

Now let us consider the following system of N simultaneous equations in N

$$\text{unknowns } x = \begin{bmatrix} x_1 \\ \vdots \\ x_N \end{bmatrix}$$

$$(14) \quad Ax = b \text{ where}$$

$A = N$ by matrix of coefficients

$$b = \begin{bmatrix} b_1 \\ \vdots \\ b_N \end{bmatrix} \text{ vector of constants.}$$

If $|A| \neq 0$, the solution to (14) is given by

$$(15) \quad A^{-1}Ax = A^{-1}b$$

$$\text{or} \quad Ix = A^{-1}b$$

$$\text{or} \quad x^* = A^{-1}b$$

$$= \frac{\begin{bmatrix} A_{11} & \dots & A_{1N} \\ A_{21} & \dots & A_{2N} \\ \vdots & & \\ A_{N1} & \dots & A_{NN} \end{bmatrix} \begin{bmatrix} b \\ \vdots \\ b \end{bmatrix}}{|A|}$$

using lemma (11) (and problems 10 and 11; i.e. that left and right inverses exist and coincide.)

$$(16) \quad \text{Therefore, } x_i^* = \text{ith component of } x$$

$$= \frac{b_1 A_{1i} + b_2 A_{2i} + \dots + b_N A_{Ni}}{|A|}$$

(perform the relevant matrix multiplication)

$$= \frac{|A_{\bullet 1}, A_{\bullet 2}, \dots, A_{\bullet i-1}, b, A_{\bullet i+1}, \dots, A_{\bullet N}|}{|A|}$$

Using (13) with the vector b replacing the i th column of A for $i = 1, 2, \dots, N$.

That is x_i may be found by replacing the i th column of the matrix A by the column vector b , take the determinant of the resulting matrix and divide by the determinant of the original matrix A . Result (16) is known as *Cramer's Rule*.

Problem 13: Given $Ax = b$ where $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 3 & 4 \end{bmatrix}$ and $b = \begin{bmatrix} 5 \\ 1 \\ 0 \end{bmatrix}$ calculate the solution x^* .

6. The Gauss Elimination Method for Constructing A^{-1}

Define the following two *elementary row operations*:

- (i) add a scalar times a row of a matrix to another row of a matrix and
- (ii) multiply a row of a matrix by a *nonzero* scalar.

The above two types of elementary row operations can be applied to a square matrix A in order to determine whether A^{-1} exists and if A^{-1} exists, these operations can be used to construct an effective method for actually constructing A^{-1} .

Recall the Gaussian algorithm in section 3 above which used elementary row operations of type (i) above to reduce A down to an upper triangular matrix U . If any of the diagonal elements of U are equal to zero; i.e., we have $u_{ii} = 0$ for some i , then

$$|A| = \prod_{i=1}^N u_{ii} = 0$$

and the results in the previous section tell us that A^{-1} cannot exist.

However, if all of the $u_{ii} \neq 0$, then we can continue to use elementary row operations of the first type to further reduce U into a diagonal matrix, say

$$(17) \quad D = \begin{bmatrix} u_{11} & 0 & \dots & 0 \\ 0 & u_{22} & \dots & 0 \\ \vdots & & & \\ 0 & 0 & \dots & u_{NN} \end{bmatrix}$$

Finally, once A has been reduced to the form (17), we can apply type (ii) elementary row operations and reduce the diagonal matrix to the N by N identity matrix I_N .

Each of the two types of elementary row operations can be represented by premultiplying A by an N by N matrix. The operation of adding k times row i of A to row j can be accomplished by premultiplying A by the following matrix:

$$(18) \quad E = I_N + k e_j e_i^T; \quad i \neq j.$$

Note that E is lower triangular if $i < j$ and is upper triangular if $i > j$. In either case, $|E| = 1$.

The operation of multiplying the i th row of A by $k \neq 0$ can be accomplished by premultiplying A by the diagonal matrix

$$(19) \quad D = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0, k & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix} \text{ ith row; i.e., } d_{jj} = 1 \text{ if } i \neq j \text{ and } d_{ii} = k.$$

Note that the determinant of D equals $k \neq 0$.

In the case where $|A| \neq 0$, we can premultiply A by a series of elementary row matrices of the form (18) and (19), say E_n, E_{n-1}, \dots, E_1 such that the transformed A is reduced to the identity matrix; i.e., we have

$$(20) \quad E_n E_{n-1} \dots E_1 A = I_N$$

Thus $B = (E_n E_{n-1} \dots E_1)$ is a left inverse for A by the definition of a left inverse.

To construct B , we need only apply the elementary row operation matrices E_n, E_{n-1}, \dots, E_1 to I_N : i.e.,

$$(21) \quad B = A^{-1} = E_n E_{n-1} \dots E_1 I_N.$$

Thus as we reduce A to I_N by means of elementary row operations, apply the same elementary row operations to I_N and in the end, I_N will be transformed into A^{-1} .

Example:

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}, \quad I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Take -3 times row 1 and add to row 2; get:

$$\begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & 0 \\ 3 & 1 \end{bmatrix}$$

Multiply the second row by $-1/2$; get:

$$\begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & 0 \\ 3/2 & 1/2 \end{bmatrix}$$

Now add -2 times row 2 to row 1; get:

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \begin{bmatrix} -1/2 & 1 \\ 3/2 & 1/2 \end{bmatrix} = A^{-1}.$$

$$\text{Check: } A \cdot A^{-1} = \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} -1/2 & 1 \\ 3/2 & 1/2 \end{bmatrix} = \begin{bmatrix} -1/2 + 3 & 1 + 1 \\ 3/2 + 1/2 & 1/2 \end{bmatrix} = \begin{bmatrix} 5/2 & 2 \\ 2 & 1/2 \end{bmatrix} = I_2.$$

Now each elementary operation can be represented by means of a matrix; i.e., the first elementary row operation can be represented by the matrix E_1 :

$$E_1 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ and } E_1 A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

The second elementary row operation matrix is E_2 :

$$E_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1/2 \end{bmatrix} \text{ and } E_2 (E_1 A) = \begin{bmatrix} 1 & 0 \\ 0 & 1/2 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 3/2 & 2 \end{bmatrix}$$

The final elementary row operation matrix is E_3 :

$$E_3 = \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} \text{ and } E_3 (E_2 E_1 A) = \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3/2 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 3/2 & 1 \end{bmatrix}$$

Thus we have $E_3(E_2(E_1 A)) = (E_3 E_2 E_1)A = I_2$. Thus $E_3 E_2 E_1$ is a left inverse for A and by the results in the previous section is also the unique right inverse.

Problem 13: Let A be a 2 by N matrix. Find a sequence of elementary row operations of the form defined by (i) and (ii) above that will interchange the rows of A ; i.e., transform $A = \begin{bmatrix} A_{1 \cdot} \\ A_{2 \cdot} \end{bmatrix}$ into $\begin{bmatrix} A_{2 \cdot} \\ A_{1 \cdot} \end{bmatrix}$ using the two elementary row operations that we have defined. *Hint:* four elementary row operations will be required.