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## A NEW METHOD FOR THE INVERSE OF THE SQUARE MATRICES

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**Abstract:** In this paper, we introduce a new method to find the inverse of the square matrices.

**Keywords:** matrix; square matrix; inverse matrix; cross product.

**2010 AMS Subject Classification:** Primary 15A09.

### 1. INTRODUCTION

A matrix is a rectangular arrangement of scalar numbers in the form of rows and columns. The size of the matrices is determined by the number of rows and columns, and a matrix with  $m$  rows and  $n$  columns is named  $m \times n$ . Matrices are generally used to solve systems of equations with  $n$  unknowns and  $m$  equations. In addition, matrices are used for mathematical transformations. Therefore, for computer programmers, the use of matrices in program writing provides a much easier way to solve the problem. In order to solve the systems of equations with  $n$  unknown equations and having  $m$  equations, the reduction system (Gaus Reduction Method or Gaus Jordan Reduction Method) or the inverse of the matrices, if any, are used on the coefficient matrix and the attached matrix of the equation system. For these methods, in general, the matrix can be reduced by using elementary row operations on the attached matrix of the equation

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system or vice versa. In this study, inverse matrix finding method which is developed for the matrix of coefficients matrix in systems with  $n$  equations with  $n$  unknowns and used in the solution of linear equation systems is discussed. For this purpose, the methods developed to find the inverse of a matrix are examined.

We know that there is a unique inverse matrix  $A^{-1}$  for any square matrix  $A$  if the determinant of  $A$  is different from zero. In the literature [1,2,3,4,5,6] there are some algorithms for constructing  $A^{-1}$ . Nowadays, 3 methods are used to find the inverse of a square matrix. The first is the Montante's Method (Bareiss algorithm) [1], the second is the Gauss Jordan Elimination method and the third is the use of the adjoint matrix. It is generally used elementary row operations or the formula  $A^{-1} = \frac{\text{Adj}(A)}{\det(A)}$ .

In this paper we introduced a new method and algorithm to find the inverse of a square matrix  $A$  if  $|A| \neq 0$ . Our method is new and it is more easy than others.

## 2. MAIN RESULTS

### A New Method For $A^{-1}$

**Theorem 2.1:** Let  $A = \begin{bmatrix} a_{11} & a_{21} & a_{31} \\ a_{21} & a_{22} & a_{32} \\ a_{31} & a_{23} & a_{33} \end{bmatrix}$  be a  $3 \times 3$  matrix and  $\det(A) \neq 0$ . We know that there

is only one unique inverse matrix  $A^{-1} = \begin{bmatrix} b_{11} & b_{21} & b_{31} \\ b_{21} & b_{22} & b_{32} \\ b_{31} & b_{23} & b_{33} \end{bmatrix}$ .

If  $i$ -th row of the matrix  $A$  is  $R_i$  and  $i$ -th column of the matrix  $A^{-1}$  is  $C_i$  ( $R_i$  and  $C_i$  are vectors in  $R^3$ ,  $R$ : Real or complex vector spaces). Then

$$C_1 = \frac{1}{\det A} (R_2 \times R_3)$$

$$C_2 = \frac{1}{\det A} (R_3 \times R_1)$$

$$C_3 = \frac{1}{\det A} (R_1 \times R_2)$$

where  $R_i \times R_j$  is the cross product of two vectors in  $R^3$ . That is

$$A^{-1} = [C_1 \ C_2 \ C_3] .$$

**Proof:** Let  $A = \begin{bmatrix} R_1 \\ R_2 \\ R_3 \end{bmatrix}$  and  $A^{-1} = [C_1 \ C_2 \ C_3]$ . We must show that  $A.A^{-1} = A^{-1}.A = I_{3 \times 3}$ .

$$A.A^{-1} = \begin{bmatrix} R_1 \\ R_2 \\ R_3 \end{bmatrix} [C_1 \ C_2 \ C_3] = \begin{bmatrix} R_1 C_1 & R_1 C_2 & R_1 C_3 \\ R_2 C_1 & R_2 C_2 & R_2 C_3 \\ R_3 C_1 & R_3 C_2 & R_3 C_3 \end{bmatrix}$$

$$R_1 C_1 = R_1 \left[ \frac{1}{|A|} ((R_2 \times R_3)) \right] = \frac{1}{|A|} R_1 (R_2 \times R_3) = \frac{|A|}{|A|} = 1 \text{ since } R_1 (R_2 \times R_3) \text{ is triple product.}$$

$$R_1 C_2 = R_1 \left[ \frac{1}{|A|} ((R_3 \times R_1)) \right] = \frac{1}{|A|} R_1 (R_3 \times R_1) = \frac{0}{|A|} = 0 \text{ since } R_3 \times R_1 \text{ is perpendicular to both } R_3 \text{ and } R_1.$$

$$R_1 C_3 = R_1 \left[ \frac{1}{|A|} ((R_1 \times R_2)) \right] = \frac{1}{|A|} R_1 (R_1 \times R_2) = \frac{0}{|A|} = 0 \text{ since } R_1 \times R_2 \text{ is perpendicular to both } R_1 \text{ and } R_2.$$

$$R_2 C_1 = R_2 \left[ \frac{1}{|A|} ((R_2 \times R_3)) \right] = \frac{1}{|A|} R_2 (R_2 \times R_3) = \frac{0}{|A|} = 0 \text{ since } R_2 \times R_3 \text{ is perpendicular to both } R_2 \text{ and } R_3.$$

$$R_2 C_2 = R_2 \left[ \frac{1}{|A|} ((R_3 \times R_1)) \right] = \frac{1}{|A|} R_2 (R_3 \times R_1) = \frac{|A|}{|A|} = 1 \text{ since } R_2 (R_3 \times R_1) \text{ is triple product.}$$

$$R_2 C_3 = R_2 \left[ \frac{1}{|A|} ((R_1 \times R_2)) \right] = \frac{1}{|A|} R_2 (R_1 \times R_2) = \frac{0}{|A|} = 0 \text{ since } R_1 \times R_2 \text{ is perpendicular to both } R_1 \text{ and } R_2.$$

$$R_3 C_1 = R_3 \left[ \frac{1}{|A|} ((R_2 \times R_3)) \right] = \frac{1}{|A|} R_3 (R_2 \times R_3) = \frac{0}{|A|} = 0 \text{ since } R_2 \times R_3 \text{ is perpendicular to both } R_2 \text{ and } R_3.$$

$$R_3 C_2 = R_3 \left[ \frac{1}{|A|} ((R_3 \times R_1)) \right] = \frac{1}{|A|} R_3 (R_3 \times R_1) = \frac{0}{|A|} = 0 \text{ since } R_3 \times R_1 \text{ is perpendicular to both } R_3 \text{ and } R_1.$$

$$R_3 C_3 = R_3 \left[ \frac{1}{|A|} ((R_1 \times R_2)) \right] = \frac{1}{|A|} R_3 (R_1 \times R_2) = \frac{|A|}{|A|} = 1 \text{ since } R_3 (R_1 \times R_2) \text{ is triple product.}$$

So we obtain  $A.A^{-1} = I_{3 \times 3}$ .

As a result the inverse of a 3x3 matrix  $A = \begin{bmatrix} R_1 \\ R_2 \\ R_3 \end{bmatrix}$  is also  $A^{-1} = \frac{1}{|A|} \begin{bmatrix} R_2 \times R_3 \\ R_3 \times R_1 \\ R_1 \times R_2 \end{bmatrix}$ .

(it is also  $\text{Adj}A = \begin{bmatrix} R_2 \times R_3 \\ R_3 \times R_1 \\ R_1 \times R_2 \end{bmatrix}$ ).

**Example 2.2:** Let  $A = \begin{bmatrix} 1 & -1 & 2 \\ 3 & 1 & 4 \\ 1 & 5 & 7 \end{bmatrix}$  be 3x3 matrix. We can find the inverse of A by applying the new algorithm.

$$|A| = \begin{vmatrix} 1 & -1 & 2 \\ 3 & 1 & 4 \\ 1 & 5 & 7 \end{vmatrix} = 1 \begin{vmatrix} 1 & 4 \\ 5 & 7 \end{vmatrix} + 1 \begin{vmatrix} 3 & 4 \\ 1 & 7 \end{vmatrix} + 2 \begin{vmatrix} 3 & 1 \\ 1 & 5 \end{vmatrix} = (7-20) + (21-4) + 2(15-1) = -13+17+28 = 32.$$

Let  $A^{-1} = [C_1 \ C_2 \ C_3]$ .  
where

$$C_1 = \frac{1}{|A|} (R_2 \times R_3) = \frac{1}{32} \begin{vmatrix} i & j & k \\ 3 & 1 & 4 \\ 1 & 5 & 7 \end{vmatrix} = \frac{1}{32} (-13i - 17j + 14k)$$

$$C_2 = \frac{1}{|A|} (R_3 \times R_1) = \frac{1}{32} \begin{vmatrix} i & j & k \\ 1 & 5 & 7 \\ 1 & -1 & 2 \end{vmatrix} = \frac{1}{32} (17i + 5j - 6k)$$

$$C_3 = \frac{1}{|A|} (R_1 \times R_2) = \frac{1}{32} \begin{vmatrix} i & j & k \\ 1 & -1 & 2 \\ 3 & 1 & 4 \end{vmatrix} = \frac{1}{32} (-6i + 2j + 4k)$$

Then  $A^{-1} = \frac{1}{32} \begin{bmatrix} -13 & 17 & -6 \\ -17 & 5 & 2 \\ 14 & -6 & 4 \end{bmatrix}$ .

$$\begin{aligned} A \cdot A^{-1} &= \frac{1}{32} \begin{bmatrix} 1 & -1 & 2 \\ 3 & 1 & 4 \\ 1 & 5 & 7 \end{bmatrix} \begin{bmatrix} -13 & 17 & -6 \\ -17 & 5 & 2 \\ 14 & -6 & 4 \end{bmatrix} \\ &= \frac{1}{32} \begin{bmatrix} -13 + 17 + 28 & 17 - 5 - 12 & -6 - 2 + 8 \\ -39 - 17 + 56 & 51 + 5 - 24 & -18 + 2 + 16 \\ -13 - 65 + 98 & 17 + 25 - 42 & -6 + 10 + 28 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \end{aligned}$$

We cannot use the above method to generalize the new method since the cross product of two vectors is not defined for  $n > 3$ . We need the following lemma.

**Lemma 2.3:** Let  $R^n$  be vector spaces with dimension  $n \geq 3$ . Let  $v_1, v_2, v_3, \dots, v_n \in R^n$ .

Define  $\Lambda: R^n \times R^n \times \dots \times R^n \rightarrow R^n$

$$v_1 \Lambda v_2 \Lambda v_3 \Lambda \dots \Lambda \tilde{v}_1 \Lambda \dots \Lambda v_n = \begin{vmatrix} k_1 & k_2 & \dots & k_n \\ & & v_1 & \\ & & v_2 & \\ & & & v_n \end{vmatrix}$$

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where  $k_1=(1,0,\dots,0)$ ,  $k_2=(0,1,0,\dots,0)$ ,  $k_3=(0,0,1,0,\dots,0),\dots, k_n=(0,0,\dots,0,1)$  are standart basis vectors in  $R^n$ . Then

$$v_j \cdot (v_1 \wedge v_2 \wedge v_3 \wedge \dots \wedge \tilde{v}_i \wedge \dots \wedge v_n) = \begin{cases} \begin{vmatrix} v_j \\ v_1 \\ \vdots \\ v_n \end{vmatrix} & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

That is  $v_j$  is orthogonal to  $v_1 \wedge v_2 \wedge v_3 \wedge \dots \wedge \tilde{v}_i \wedge \dots \wedge v_n$  if  $i \neq j$ .

**Proof:** It is clear that for  $n=3$  by Theorem 1.1. We obtain

$$\begin{aligned} v_1 \wedge v_2 \wedge v_3 \wedge \dots \wedge \tilde{v}_i \wedge \dots \wedge v_n &= \begin{vmatrix} k_1 & k_2 & \dots & k_n \\ & v_1 & & \\ & v_2 & & \\ & & & v_n \end{vmatrix} \\ &= k_1 M_{k_1} - k_2 M_{k_2} + \dots \widetilde{k_i M_{k_i}} \dots + (-1)^n k_n M_{k_n} \end{aligned}$$

from the definition of the mapping.

So the following statement

$$\begin{aligned} v_j \cdot (v_1 \wedge v_2 \wedge v_3 \wedge \dots \wedge \tilde{v}_i \wedge \dots \wedge v_n) &= v_j \cdot \begin{vmatrix} k_1 & k_2 & \dots & k_n \\ & v_1 & & \\ & v_2 & & \\ & & & \dots \\ & & & v_n \end{vmatrix} \\ &= v_j \cdot (k_1 M_{k_1} - k_2 M_{k_2} + \dots \widetilde{k_i M_{k_i}} \dots + (-1)^n k_n M_{k_n}) \\ &= v_{j_1} M_{k_1} - v_{j_2} M_{k_2} + \dots \widetilde{v_{j_i} M_{k_i}} \dots + (-1)^n v_{j_n} M_{k_n} \\ &= \begin{vmatrix} v_j \\ v_1 \\ v_2 \\ \dots \\ v_n \end{vmatrix} \end{aligned}$$

is true for  $n$ .

Finally we obtain

$$v_j \cdot (v_1 \wedge v_2 \wedge v_3 \wedge \dots \wedge \tilde{v}_i \wedge \dots \wedge v_n) = \begin{cases} \begin{vmatrix} v_j \\ v_1 \\ \vdots \\ v_n \\ 0 \end{vmatrix} & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

by the definition of the determinant function.

**Corollary 2.4:**  $v_1 \cdot (v_2 \wedge v_3 \wedge \dots \wedge \tilde{v}_1 \wedge \dots \wedge v_n) = \begin{vmatrix} v_1 \\ v_2 \\ v_3 \\ \vdots \\ v_n \end{vmatrix} = |A|$

**Theorem 2.5:** Let  $A$  be  $n \times n$  matrix and  $|A| \neq 0$  with  $n \geq 4$ . If  $i$ -th row of the matrix  $A$  is  $R_i$  and  $j$ -th column of the matrix  $A^{-1}$  is  $C_j$  ( $R_i$  and  $C_j$  are vectors in  $R^n$ ) then

$$A^{-1} = [C_1 \ C_2 \ \dots \ C_n]$$

where  $C_j = \frac{(-1)^{j+1}}{|A|} (R_1 \wedge R_2 \wedge \dots \wedge \tilde{R}_j \wedge \dots \wedge R_n)$ .

**Proof:** Let  $A = \begin{bmatrix} R_1 \\ R_2 \\ \vdots \\ R_n \end{bmatrix}$  and  $A^{-1} = [C_1 \ C_2 \ \dots \ C_n]$  be matrices.

$$A \cdot A^{-1} = \begin{bmatrix} R_1 \\ R_2 \\ \vdots \\ R_n \end{bmatrix} [C_1 \ C_2 \ \dots \ C_n] = \begin{bmatrix} R_1 C_1 & R_1 C_2 & \dots & R_1 C_n \\ \vdots & \vdots & \ddots & \vdots \\ R_n C_1 & R_n C_2 & \dots & R_n C_n \end{bmatrix}$$

$$\begin{aligned} R_i \cdot C_j &= R_i \cdot \left[ \frac{(-1)^{j+1}}{|A|} (R_1 \wedge R_2 \wedge \dots \wedge \tilde{R}_j \wedge \dots \wedge R_n) \right] \\ &= \frac{(-1)^{j+1}}{|A|} R_i \cdot (R_1 \wedge R_2 \wedge \dots \wedge \tilde{R}_j \wedge \dots \wedge R_n) \quad \text{if } i = j. \\ &= \frac{(-1)^{j+1}}{|A|} (-1)^{j+1} |A| = 1 \end{aligned}$$

$$R_i \cdot (R_1 \wedge R_2 \wedge \dots \wedge \tilde{R}_j \wedge \dots \wedge R_n) = 0 \quad (R_i \text{ is orthogonal to } R_1 \wedge R_2 \wedge \dots \wedge \tilde{R}_j \wedge \dots \wedge R_n)$$

if  $i \neq j$  by Lemma 1.2.

$$\text{We obtain } R_i \cdot C_j = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{if } i \neq j \end{cases}$$

That is

$$\begin{aligned} A^{-1} &= [C_1 \ C_2 \ \dots \ C_n] \\ &= \frac{(-1)^{j+1}}{|A|} [R_2 \wedge R_3 \wedge \dots \wedge R_n \quad R_1 \wedge R_3 \wedge \dots \wedge R_n \quad \dots \quad R_1 \wedge R_2 \wedge \dots \wedge R_{n-1}] \end{aligned}$$

Now we can use the new algorithm for constructing  $A^{-1}$  for the given some examples.

**Example 2.6:** Let  $A = \begin{bmatrix} -1 & 3 & 2 & 0 \\ 2 & 1 & 1 & 0 \\ 1 & 4 & 0 & 1 \\ 0 & 1 & 1 & 1 \end{bmatrix}$  be a 4x4 matrix.

$$|A| = -1 \begin{vmatrix} -1 & 3 & 2 \\ 2 & 1 & 1 \\ 0 & 1 & 1 \end{vmatrix} + 1 \begin{vmatrix} -1 & 3 & 2 \\ 2 & 1 & 1 \\ 1 & 4 & 0 \end{vmatrix} = 23$$

Let  $A^{-1} = [C_1 \ C_2 \ C_3 \ C_4]$

$$C_1 = \frac{1}{|A|} (R_2 \wedge R_3 \wedge R_4) = \frac{1}{23} \begin{vmatrix} i & j & k & t \\ 2 & 1 & 1 & 0 \\ 1 & 4 & 0 & 1 \\ 0 & 1 & 1 & 1 \end{vmatrix} = \frac{1}{23} (-4i + 3j + 5k - 8t)$$

$$C_2 = \frac{-1}{|A|} (R_1 \wedge R_3 \wedge R_4) = -\frac{1}{23} \begin{vmatrix} i & j & k & t \\ -1 & 3 & 2 & 0 \\ 1 & 4 & 0 & 1 \\ 0 & 1 & 1 & 1 \end{vmatrix} = -\frac{1}{23} (-9i + j - 6k + 5t)$$

$$C_3 = \frac{1}{|A|} (R_1 \wedge R_2 \wedge R_4) = \frac{1}{23} \begin{vmatrix} i & j & k & t \\ -1 & 3 & 2 & 0 \\ 2 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 \end{vmatrix} = \frac{1}{23} (i + 5j - 7k + 2t)$$

$$C_4 = \frac{-1}{|A|} (R_1 \wedge R_2 \wedge R_3) = -\frac{1}{23} \begin{vmatrix} i & j & k & t \\ -1 & 3 & 2 & 0 \\ 2 & 1 & 1 & 0 \\ 1 & 4 & 0 & 1 \end{vmatrix} = -\frac{1}{23} (i + 15j - 7k - 21t)$$

We find  $A^{-1} = \frac{1}{23} \begin{bmatrix} -4 & 9 & 1 & -1 \\ 3 & -1 & 5 & -5 \\ 5 & 6 & -7 & 7 \\ -8 & -5 & 2 & 21 \end{bmatrix}$  if the new method is applied.

The result satisfies the following equality

$$A.A^{-1} = \frac{1}{23} \begin{bmatrix} 4+9+10+0 & -9-3+12+0 & -1+15-14+0 & 0 \\ -8+3+5 & 18-1+6 & 2+5-7 & 0 \\ -4+12+8 & 9-4-5 & 1+20+2 & -1-20+21 \\ 3+5-8 & -1+6-5 & 5-7+2 & -5+7+21 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

**Example 2.6:** Let  $A = \begin{bmatrix} 2 & 4 & -3 & 2 & 1 & 2 & -1 \\ 1 & 3 & -1 & -3 & 0 & 3 & 2 \\ -4 & 2 & 0 & 2 & -3 & 4 & 3 \\ -2 & -1 & 2 & 1 & 3 & 2 & 2 \\ 0 & 2 & -1 & 3 & 5 & 2 & 1 \\ -3 & 1 & 2 & -2 & -1 & 1 & 0 \\ 2 & 3 & 6 & 5 & 4 & 2 & -1 \end{bmatrix}$  be a 7x7 matrix. Then  $|A|= 22740$ .

Let  $A^{-1} = [C_1 \ C_2 \ C_3 \ C_4 \ C_5 \ C_6 \ C_7]$

$$C_1 = \frac{1}{|A|} (R_2 \wedge R_3 \wedge R_4 \wedge R_5 \wedge R_6 \wedge R_7)$$

$$= \begin{vmatrix} k_1 & k_2 & k_3 & k_4 & k_5 & k_6 & k_7 \\ 1 & 3 & -1 & -3 & 0 & 3 & 2 \\ -4 & 2 & 0 & 2 & -3 & 4 & 3 \\ -2 & -1 & 2 & 1 & 3 & 2 & 2 \\ 0 & 2 & -1 & 3 & 5 & 2 & 1 \\ -3 & 1 & 2 & -2 & -1 & 1 & 0 \\ 2 & 3 & 6 & 5 & 4 & 2 & -1 \end{vmatrix}$$

$$= \frac{1}{22740} (924k_1 - 7572k_2 - 3036k_3 - 508k_4 - 488k_5 + 14912k_6 - 13752k_7)$$



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$$C_2 = \frac{-1}{|A|} (R_1 \wedge R_3 \wedge R_4 \wedge R_5 \wedge R_6 \wedge R_7)$$

$$= \begin{vmatrix} k_1 & k_2 & k_3 & k_4 & k_5 & k_6 & k_7 \\ 2 & 4 & -3 & 2 & 1 & 2 & -1 \\ -4 & 2 & 0 & 2 & -3 & 4 & 3 \\ -2 & -1 & 2 & 1 & 3 & 2 & 2 \\ 0 & 2 & -1 & 3 & 5 & 2 & 1 \\ -3 & 1 & 2 & -2 & -1 & 1 & 0 \\ 2 & 3 & 6 & 5 & 4 & 2 & -1 \end{vmatrix}$$

$$= \frac{-1}{22740} (-2652 k_1 - 2484 k_2 - 1032 k_3 + 2344 k_4 - 76 k_5 + 1204 k_6 - 5124 k_7)$$

$$C_3 = \frac{1}{|A|} (R_1 \wedge R_2 \wedge R_4 \wedge R_5 \wedge R_6 \wedge R_7)$$

$$= \begin{vmatrix} k_1 & k_2 & k_3 & k_4 & k_5 & k_6 & k_7 \\ 2 & 4 & -3 & 2 & 1 & 2 & -1 \\ 1 & 3 & -1 & -3 & 0 & 3 & 2 \\ -2 & -1 & 2 & 1 & 3 & 2 & 2 \\ 0 & 2 & -1 & 3 & 5 & 2 & 1 \\ -3 & 1 & 2 & -2 & -1 & 1 & 0 \\ 2 & 3 & 6 & 5 & 4 & 2 & -1 \end{vmatrix}$$

$$= \frac{1}{22740} (-789 k_1 + 1962 k_2 + 156 k_3 + 2378 k_4 - 2192 k_5 - 2077 k_6 + 4212 k_7)$$

$$C_4 = \frac{-1}{|A|} (R_1 \wedge R_2 \wedge R_3 \wedge R_5 \wedge R_6 \wedge R_7)$$

$$= \begin{vmatrix} k_1 & k_2 & k_3 & k_4 & k_5 & k_6 & k_7 \\ 2 & 4 & -3 & 2 & 1 & 2 & -1 \\ 1 & 3 & -1 & -3 & 0 & 3 & 2 \\ -4 & 2 & 0 & 2 & -3 & 4 & 3 \\ 0 & 2 & -1 & 3 & 5 & 2 & 1 \\ -3 & 1 & 2 & -2 & -1 & 1 & 0 \\ 2 & 3 & 6 & 5 & 4 & 2 & -1 \end{vmatrix}$$

$$= \frac{-1}{22740} (-1623 k_1 + 13374 k_2 + 1272 k_3 + 1606 k_4 - 964 k_5 - 18539 k_6 + 11604 k_7)$$

$$C_5 = \frac{1}{|A|} (R_1 \wedge R_2 \wedge R_3 \wedge R_4 \wedge R_6 \wedge R_7)$$

$$= \begin{vmatrix} k_1 & k_2 & k_3 & k_4 & k_5 & k_6 & k_7 \\ 2 & 4 & -3 & 2 & 1 & 2 & -1 \\ 1 & 3 & -1 & -3 & 0 & 3 & 2 \\ -4 & 2 & 0 & 2 & -3 & 4 & 3 \\ -2 & -1 & 2 & 1 & 3 & 2 & 2 \\ -3 & 1 & 2 & -2 & -1 & 1 & 0 \\ 2 & 3 & 6 & 5 & 4 & 2 & -1 \end{vmatrix}$$

$$= \frac{1}{22740} (-3816 k_1 + 10008 k_2 - 456 k_3 + 1212 k_4 + 3492 k_5 - 14628 k_6 + 10428 k_7)$$

$$C_6 = \frac{-1}{|A|} (R_1 \wedge R_2 \wedge R_3 \wedge R_4 \wedge R_5 \wedge R_7)$$

$$= \begin{vmatrix} k_1 & k_2 & k_3 & k_4 & k_5 & k_6 & k_7 \\ 2 & 4 & -3 & 2 & 1 & 2 & -1 \\ 1 & 3 & -1 & -3 & 0 & 3 & 2 \\ -4 & 2 & 0 & 2 & -3 & 4 & 3 \\ -2 & -1 & 2 & 1 & 3 & 2 & 2 \\ 0 & 2 & -1 & 3 & 5 & 2 & 1 \\ 2 & 3 & 6 & 5 & 4 & 2 & -1 \end{vmatrix}$$

$$= \frac{1}{22740} (5148 k_1 - 3204 k_2 - 672 k_3 + 2584 k_4 - 1636 k_5 + 784 k_6 + 4596 k_7)$$

$$C_7 = \frac{1}{|A|} (R_1 \wedge R_2 \wedge R_3 \wedge R_4 \wedge R_5 \wedge R_6)$$

$$= \begin{vmatrix} k_1 & k_2 & k_3 & k_4 & k_5 & k_6 & k_7 \\ 2 & 4 & -3 & 2 & 1 & 2 & -1 \\ 1 & 3 & -1 & -3 & 0 & 3 & 2 \\ -4 & 2 & 0 & 2 & -3 & 4 & 3 \\ -2 & -1 & 2 & 1 & 3 & 2 & 2 \\ 0 & 2 & -1 & 3 & 5 & 2 & 1 \\ -3 & 1 & 2 & -2 & -1 & 1 & 0 \end{vmatrix}$$

$$= \frac{1}{22740} (1443 k_1 + 1686 k_2 + 2568 k_3 + 954 k_4 - 516 k_5 - 1101 k_6 + 1116 k_7)$$

We find

$$A^{-1} = \frac{1}{22740} \begin{bmatrix} 924 & 2652 & -789 & 1623 & -3816 & -5148 & 1443 \\ -7572 & 2484 & 1962 & -13374 & 10008 & 3204 & 1686 \\ -3036 & 1032 & 156 & -1272 & -456 & 672 & 2568 \\ -508 & -2344 & 2378 & -1606 & 1212 & -2584 & 954 \\ -488 & 76 & -2192 & 964 & 3492 & 1636 & -516 \\ 14912 & -1204 & -2077 & 18539 & -14628 & -784 & -1101 \\ -13752 & 5124 & 4212 & -11604 & 10428 & -4596 & 1116 \end{bmatrix}$$

### CONFLICT OF INTERESTS

The authors declare that there is no conflict of interests.

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