

Lecture 5: Determinant, computation of inverse & Cramer's rule

The definition of determinant by permutation is not convenient for computation. Here we will give further properties of determinant which will help in computation.

Let $A = (a_{ij})$ be a square matrix of order $n > 1$. Let A_{ij} be the submatrix of order $n-1$ obtained from A by deleting its i th row and j th column of A . Determinant of A_{ij} is called ij th minor of A .

Theorem: For each $i \in \{1, 2, \dots, n\}$, $n > 1$

$$|A| = \sum_{j=1}^n (-1)^{i+j} a_{ij} |A_{ij}|.$$

Proof: Fix $i \in \{1, 2, \dots, n\}$

For $j \in \{1, 2, \dots, n\}$,

$$\text{let } A_j = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1j} & \dots & a_{1n} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_{ij} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nj} & \dots & a_{nn} \end{pmatrix}$$

i.e. A_j is the matrix (b_{pq}) such that
 $b_{pq} = a_{pq}$ if $p \neq i$, $b_{iq} = 0$ if $q \neq j$
 & $b_{ij} = a_{ij}$...

$$A = \begin{pmatrix} a_{11} & \dots & a_{1j} & \dots & a_{1n} \\ \vdots & & \vdots & & \vdots \\ a_{i1} + 0 + \dots + 0 & \dots & 0 + \dots + a_{ij} + \dots + 0 & \dots & 0 + \dots + 0 + a_{in} \\ \vdots & & \vdots & & \vdots \\ a_{n1} & \dots & a_{nj} & \dots & a_{nn} \end{pmatrix}$$

Using property 4, we have

$$|A| = |A_1| + \dots + |A_j| + \dots + |A_n|$$

Claim: $|A_j| = \sum_{i=1}^n (-1)^{i+j} a_{ij} |A_{ij}|$

$$E_{nn-1} E_{n-1, n-2} \dots E_{i+1, i} A_j E_{j, j+1} \dots E_{n-1, n}$$

$$= \begin{pmatrix} & & & a_{1j} \\ & & & a_{2j} \\ & & & \vdots \\ A_{ij} & & & \\ 0 & 0 & \dots & 0 & a_{ij} \end{pmatrix} = \dots \text{ (say)}$$

C is of the form $\begin{pmatrix} B & * \\ 0 & C_{nn} \end{pmatrix}$,

where B is a square matrix of order $n-1$. Let $c = C(i, j)$

$$|C| = \sum_{\sigma \in S_n} \text{sign}(\sigma) C_{1, \sigma(1)} \cdots C_{n, \sigma(n)}$$

$$= \sum_{\substack{\sigma \in S_n \\ \sigma(n) = n}} \text{sign}(\sigma) C_{1, \sigma(1)} \cdots C_{n-1, \sigma(n-1)} C_{nn}$$

(since $C_{n, \sigma(n)} = 0$ if $\sigma(n) \neq n$)

$$= C_{nn} \sum_{\sigma \in S_{n-1}} \text{sign}(\sigma) C_{1, \sigma(1)} \cdots C_{n-1, \sigma(n-1)}$$

$$= C_{nn} |B|$$

Using it we have,

$$|E_{nn-1} \cdots E_{i+1, i} A_j E_{j, j+1} \cdots E_{n-1, n}| = a_{ij} |A_{ij}|$$

$$(-1)^{n-i} |A_j| (-1)^{n-j} = a_{ij} |A_{ij}|$$

$$\Rightarrow |A_j| = (-1)^{i+j} a_{ij} |A_{ij}|$$

$$\therefore |A| = \sum_{j=1}^n (-1)^{i+j} a_{ij} |A_{ij}| \quad \square$$

Corollary: Let $A = (a_{ij})$ be a matrix of order n . Fix $i \in \{1, 2, \dots, n\}$, then for all $k \neq i$, we have

$$\sum_{j=1}^n (-1)^{i+j} a_{kj} |A_{i,j}| = 0$$

Proof: Let B be the matrix obtained from A by replacing the i th row of A with k th row of A & remaining k th row as it is.

$$B = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \bar{a}_{k1} & \bar{a}_{k2} & \dots & \bar{a}_{kn} \\ a_{ki} & a_{k2} & \dots & a_{kn} \\ \bar{a}_{n1} & \bar{a}_{n2} & \dots & \bar{a}_{nn} \end{pmatrix} \begin{matrix} \leftarrow i\text{th row} \\ \leftarrow k\text{th row} \end{matrix}$$

By previous theorem

$$|B| = \sum_{j=1}^n (-1)^{i+j} a_{kj} |A_{i,j}|$$

B has two identical rows, hence $|B| = 0$

□

Let $B_{ij} = (-1)^{i+j} |A_{ij}|$, then

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} \begin{pmatrix} B_{11} & B_{21} & \dots & B_{n1} \\ B_{12} & B_{22} & \dots & B_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ B_{1n} & B_{2n} & \dots & B_{nn} \end{pmatrix}$$
$$= \begin{pmatrix} |A| & 0 & \dots & 0 \\ 0 & |A| & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & |A| \end{pmatrix} = |A| I_n$$

* $A B^T = |A| I_n$

The element B_{ij} is called ij th cofactor of A & the matrix B^T is called **(classical) adjoint (or adjugate)** of the matrix A . It is denoted by $\text{Adj}(A)$. Thus,

$$\boxed{A (\text{Adj}(A)) = |A| I}$$

As $A = A^T$, we also have

$$\text{Adj}(A) A = |A| I$$

$$A (\text{Adj}(A)) = (\text{Adj}(A)) A = |A| I$$

Thus if $|A| \neq 0$, $A^{-1} = \frac{1}{|A|} \text{Adj}(A)$.

Cramer's rule for solving linear equations

Consider the system of linear equations:

$$AX = D, \text{ where } A = (a_{ij})_{n \times n}$$

$$X = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \quad D = \begin{pmatrix} d_1 \\ \vdots \\ d_n \end{pmatrix}$$

This system has a unique solution if $|A| \neq 0$ & the solution is given by $X = A^{-1} B$

$$\Rightarrow x_i = \frac{1}{|A|} \sum_{j=1}^n B_{ji} d_j$$

$$= \frac{1}{|A|} (B_{1i} d_1 + B_{2i} d_2 + \dots + B_{ni} d_n)$$

Let B_i be the matrix obtained from A by replacing i th column by $(d_1, d_2, \dots, d_n)^T$.

Then $|B_i| = \sum_{j=1}^n B_{ji} d_j$ (check)

Hence, $x_i = \frac{|B_i|}{|A|}$, $i = 1, 2, \dots, n$.

If $|A| \neq 0$, then $AX = B$ either has infinitely many solutions or no solution.

Example: $A = \begin{pmatrix} 1 & 2 & 1 \\ 3 & 1 & 2 \\ 4 & 2 & 3 \end{pmatrix}$. $|A| = 1(3-4) - 2(9-8) + 1(6-4) = -1$

Cofactor matrix $\begin{pmatrix} -1 & -1 & 2 \\ -2 & -1 & 6 \\ 3 & 1 & -5 \end{pmatrix} = C$ (say)

$$\text{Adj}(A) = C^T = \begin{pmatrix} -1 & -2 & 3 \\ -1 & -1 & 1 \\ 2 & 6 & -5 \end{pmatrix}$$

$$A^{-1} = \frac{1}{|A|} \text{Adj}(A)$$

$$= \begin{pmatrix} 1 & 2 & -3 \\ 1 & 1 & -1 \\ -2 & -6 & 5 \end{pmatrix}$$