

Vv417 Lecture 6

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September 23, 2019

- Expansions by cofactors are very useful when computing the determinant of matrices that have a row or column with several zero entries.
- However, it is also useful to establish properties of determinants.

Theorem

Suppose \mathbf{A} is a square matrix.

1. If \mathbf{A} has a row or column of zeros, then

$$\det(\mathbf{A}) = 0$$

2. If \mathbf{A} has two identical rows or two identical columns, then

$$\det(\mathbf{A}) = 0$$

Q: How to show the second statement is true?

Q: Is the converse of each of those two statements true?

Q: What is the effect of elementary row operations on $\det(\mathbf{A})$?

Interchanging two rows of \mathbf{A}

Suppose \mathbf{A} is an $n \times n$ matrix and

$$\mathbf{E}_{i,j} = \begin{bmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & 0 & & 1 & \\ & & & \ddots & & \\ & & 1 & & 0 & \\ & & & & & \ddots \\ & & & & & & 1 \end{bmatrix}$$

is the elementary matrix corresponding to **interchanging** row i with row j , then

$$\det(\mathbf{E}_{i,j}\mathbf{A}) = -\det(\mathbf{A})$$

Furthermore,

$$\det(\mathbf{E}_{i,j}\mathbf{A}) = \det(\mathbf{E}_{i,j}) \det(\mathbf{A})$$

Multiplying a row of \mathbf{A} by a nonzero constant

Suppose \mathbf{A} is an $n \times n$ matrix and

$$\mathbf{E}_{(\alpha)i} = \begin{bmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & 1 & & & \\ & & & \alpha & & \\ & & & & 1 & \\ & & & & & \ddots \\ & & & & & & 1 \end{bmatrix}$$

is the elementary matrix of **multiplying** the i th row by the nonzero scalar α , then

$$\det(\mathbf{E}_{(\alpha)i}\mathbf{A}) = \alpha \det(\mathbf{A})$$

Furthermore,

$$\det(\mathbf{E}_{(\alpha)i}\mathbf{A}) = \det(\mathbf{E}_{(\alpha)i}) \det(\mathbf{A})$$

Proof

- Notice the i th row of $\mathbf{E}_{(\alpha)i}\mathbf{A}$ is

$$[\alpha a_{i1} \quad \alpha a_{i2} \quad \cdots \quad \alpha a_{in}]$$

- Expand $\det(\mathbf{E}_{(\alpha)i}\mathbf{A})$ along the i th row, we have

$$\begin{aligned}\det(\mathbf{E}_{(\alpha)i}\mathbf{A}) &= (\alpha a_{i1}C_{i1}) + (\alpha a_{i2}C_{i2}) + \cdots + (\alpha a_{in}C_{in}) \\ &= \alpha (a_{i1}C_{i1} + a_{i2}C_{i2} + \cdots + a_{in}C_{in}) \\ &= \alpha \det(\mathbf{A})\end{aligned}$$

- Using the fact $\det(\mathbf{I}) = 1$ and the result above, we have

$$\begin{aligned}\det(\mathbf{E}_{(\alpha)i}\mathbf{I}) &= \alpha \det(\mathbf{I}) = \alpha \\ \implies \det(\mathbf{E}_{(\alpha)i}) &= \alpha \\ \implies \det(\mathbf{E}_{(\alpha)i}\mathbf{A}) &= \alpha \det(\mathbf{A}) = \det(\mathbf{E}_{(\alpha)i}) \det(\mathbf{A}) \quad \square\end{aligned}$$

Adding a multiple of one row to another row

Suppose \mathbf{A} is an $n \times n$ matrix and

$$\mathbf{E}_{(\alpha)i,j} = \begin{bmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & 1 & & & \\ & & & \ddots & & \\ & & \alpha & & 1 & \\ & & & & & \ddots \\ & & & & & & 1 \end{bmatrix}$$

is the elementary matrix of **adding** α times the i th row to the j th row, then

$$\det(\mathbf{E}_{(\alpha)i,j} \mathbf{A}) = \det(\mathbf{A})$$

Furthermore,

$$\det(\mathbf{E}_{(\alpha)i,j} \mathbf{A}) = \det(\mathbf{E}_{(\alpha)i,j}) \det(\mathbf{A})$$

Proof

- Note this last type of row operations only changes the j th row,

$$\begin{bmatrix} \vdots & \vdots & \vdots \\ (a_{j1} + \alpha a_{i1}) & (a_{j2} + \alpha a_{i2}) & \cdots & (a_{jn} + \alpha a_{in}) \\ \vdots & \vdots & \vdots \end{bmatrix}$$

- Using the cofactor expansion along the j th row, we have

$$\begin{aligned} \det(\mathbf{E}_{(\alpha)i,j}\mathbf{A}) &= (a_{j1} + \alpha a_{i1})C_{j1} + (a_{j2} + \alpha a_{i2})C_{j2} + \\ &\quad \vdots \\ &\quad + (a_{jn} + \alpha a_{in})C_{jn} \\ &= (a_{j1}C_{j1} + a_{j2}C_{j2} \cdots + a_{jn}C_{jn}) \\ &\quad + \alpha(a_{i1}C_{j1} + a_{i2}C_{j2} \cdots + a_{in}C_{jn}) \\ &= \det(\mathbf{A}) + \beta \end{aligned}$$

Proof

- So we need to show the following is true

$$\beta = a_{i1}C_{j1} + a_{i2}C_{j2} \cdots + a_{in}C_{jn} = 0$$

- Consider the matrix \mathbf{A}^* which differs from \mathbf{A} only in row j , specifically,

$$[\mathbf{A}^*]_{pq} = \begin{cases} [\mathbf{A}]_{pq} & \text{if } p \neq j, \\ [\mathbf{A}]_{iq} & \text{if } p = j. \end{cases}$$

- Notice, by construction, we have

$$a_{jq}^* = a_{iq}^* = a_{iq} \quad \text{and} \quad C_{jq}^* = C_{jq}$$

- Since there are two identical rows in \mathbf{A}^* , namely, the i th and the j th row

$$\det(\mathbf{A}^*) = 0$$

Proof

- However, the following cofactor expansion of \mathbf{A}^* along the j th row reveals

$$\begin{aligned} 0 = \det(\mathbf{A}^*) &= a_{j1}^* C_{j1}^* + a_{j2}^* C_{j2}^* + \cdots + a_{jn}^* C_{jn}^* \\ &= a_{i1} C_{j1} + a_{i2} C_{j2} + \cdots + a_{in} C_{jn} = \beta \end{aligned}$$

- This completes the main result,

$$\det(\mathbf{E}_{(\alpha)i,j} \mathbf{A}) = \det(\mathbf{A}) + \beta = \det(\mathbf{A})$$

- Using the fact $\det(\mathbf{I}) = 1$ and the result above, we have

$$\det(\mathbf{E}_{(\alpha)i,j} \mathbf{I}) = \det(\mathbf{I}) = 1$$

$$\implies \det(\mathbf{E}_{(\alpha)i,j}) = 1$$

$$\implies \det(\mathbf{E}_{(\alpha)i,j} \mathbf{A}) = 1 \cdot \det(\mathbf{A}) = \det(\mathbf{E}_{(\alpha)i,j}) \det(\mathbf{A}) \quad \square$$

Q: Why is the following theorem obviously true given what we have discussed?

Theorem

Suppose \mathbf{A} is a square matrix.

1. If \mathbf{A} has a row or column of zeros, then

$$\det(\mathbf{A}) = 0$$

2. If \mathbf{A} has two identical rows or two identical columns, then

$$\det(\mathbf{A}) = 0$$

3. If \mathbf{A} has a row that is a multiple of another row of \mathbf{A} , then

$$\det(\mathbf{A}) = 0$$

4. If \mathbf{A} has a column that is a multiple of another column of \mathbf{A} , then

$$\det(\mathbf{A}) = 0$$

- Interchanging rows

$$\det(\mathbf{E}_{i,j}) = -1$$
$$\det(\mathbf{E}_{i,j}\mathbf{A}) = \det(\mathbf{E}_{i,j}) \det(\mathbf{A})$$

- Multiplying a row

$$\det(\mathbf{E}_{(\alpha)i}) = \alpha$$
$$\det(\mathbf{E}_{(\alpha)i}\mathbf{A}) = \det(\mathbf{E}_{(\alpha)i}) \det(\mathbf{A})$$

- Adding a multiple of a row to another

$$\det(\mathbf{E}_{(\alpha)i,j}) = 1$$
$$\det(\mathbf{E}_{(\alpha)i,j}\mathbf{A}) = \det(\mathbf{E}_{(\alpha)i,j}) \det(\mathbf{A})$$

- Hence we have essentially shown that

$$\det(\mathbf{EA}) = \det(\mathbf{E}) \det(\mathbf{A}), \quad \text{where } \mathbf{E} \text{ is an elementary matrix.}$$

- A natural question to ask next is what if \mathbf{E} is not an elementary matrix.

$$\det(\mathbf{AB})$$

- In order to show

$$\det(\mathbf{AB}) = \det(\mathbf{A}) \det(\mathbf{B})$$

we need the following result, which is surely not a surprise!

Theorem

An $n \times n$ matrix \mathbf{A} is singular if and only if

$$\det(\mathbf{A}) = 0$$

Proof

- Consider a sequence of elementary matrices such that

$$\mathbf{E}_k \mathbf{E}_{k-1} \mathbf{E}_{k-2} \dots \mathbf{E}_1 \mathbf{A} = \text{rref}(\mathbf{A}) = \mathbf{R}$$

Proof

- Since \mathbf{E}_i for $i = 1, \dots, k$ are elementary matrices, we have,

$$\det(\mathbf{E}_k) \det(\mathbf{E}_{k-1}) \det(\mathbf{E}_{k-2}) \dots \det(\mathbf{E}_1) \det(\mathbf{A}) = \det(\mathbf{R})$$

- Since the determinant of any elementary matrix is non-zero

$$\det(\mathbf{A}) = 0 \iff \det(\mathbf{R}) = 0$$

- Recall the rref of a square matrix can either be \mathbf{I} or have a row of zeros.
- If \mathbf{A} is singular, then $\mathbf{R} = \text{rref}(\mathbf{A})$ has an entire row of 0, which leads us to

$$\det(\mathbf{R}) = 0 \implies \det(\mathbf{A}) = 0$$

- If $\det(\mathbf{A}) = 0$, then

$$\det(\mathbf{R}) = 0 \implies \mathbf{R} \neq \mathbf{I}$$

which means \mathbf{A} is singular. \square

- We can now formally add one extra statement to our equivalent theorem.

Equivalence Theorem

If \mathbf{A} is an $n \times n$ matrix, then the following statements are **equivalent**,

1. \mathbf{A} is invertible.
2. $\mathbf{Ax} = \mathbf{0}$ has only the trivial solution.
3. The reduced echelon form of \mathbf{A} is \mathbf{I}_n .
4. \mathbf{A} is expressible as a product of elementary matrices.
5. $\mathbf{Ax} = \mathbf{b}$ is consistent for every $n \times 1$ matrix \mathbf{b} .
6. $\mathbf{Ax} = \mathbf{b}$ has exactly one solution for every $n \times 1$ matrix \mathbf{b} .
7. $\det(\mathbf{A}) \neq 0$.

Theorem

Suppose \mathbf{A} and \mathbf{B} are $n \times n$ matrices, then

$$\det(\mathbf{AB}) = \det(\mathbf{A}) \det(\mathbf{B})$$

Proof

- If \mathbf{A} is singular, then \mathbf{AB} is singular.

$$\begin{aligned}\det(\mathbf{AB}) &= 0 \\ &= \det(\mathbf{A}) \det(\mathbf{B})\end{aligned}$$

- If \mathbf{A} is invertible, then \mathbf{A} can be written as a product of elementary matrices

$$\mathbf{A} = \mathbf{E}_k \mathbf{E}_{k-1} \mathbf{E}_{k-2} \dots \mathbf{E}_1$$

- Therefore,

$$\begin{aligned}\det(\mathbf{AB}) &= \det(\mathbf{E}_k \mathbf{E}_{k-1} \mathbf{E}_{k-2} \dots \mathbf{E}_1 \mathbf{B}) \\ &= \det(\mathbf{E}_k) \det(\mathbf{E}_{k-1}) \det(\mathbf{E}_{k-2}) \dots \det(\mathbf{E}_1) \det(\mathbf{B}) \\ &= \det(\mathbf{E}_k \mathbf{E}_{k-1} \mathbf{E}_{k-2} \dots \mathbf{E}_1) \det(\mathbf{B}) \\ &= \det(\mathbf{A}) \det(\mathbf{B})\end{aligned}$$

Q: Is there any further connection between determinants and the solution to

$$\mathbf{Ax} = \mathbf{b}$$

besides giving us some idea on the number of solutions?

Definition

If every element in an $n \times n$ matrix \mathbf{A} is replaced by its cofactor, the resulting matrix is called the **matrix of cofactors** and is denoted \mathbf{C} . The transpose of the matrix of cofactors, is called the **adjoint** of \mathbf{A} and is denoted

$$\text{adj}(\mathbf{A}) = \mathbf{C}^T$$

Theorem

Suppose \mathbf{A} is a square matrix. If $\det(\mathbf{A}) \neq 0$, then

$$\mathbf{A}^{-1} = \frac{1}{\det(\mathbf{A})} \text{adj}(\mathbf{A})$$

Proof

- Recall we have show the following result in the cofactor proof,

$$\sum_{k=1}^n a_{ik} C_{jk} = \begin{cases} \det(\mathbf{A}) & \text{if } i = j, \\ 0 & \text{if } i \neq j. \end{cases}$$

- Consider the elements of the following product

$$\begin{aligned} \left[\mathbf{A} \frac{1}{\det(\mathbf{A})} \text{adj}(\mathbf{A}) \right]_{ij} &= \frac{1}{\det(\mathbf{A})} [\mathbf{A} \mathbf{C}^T]_{ij} = \frac{1}{\det(\mathbf{A})} \sum_{k=1}^n a_{ik} C_{jk} \\ &= \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases} \end{aligned}$$

- Therefore,

$$\mathbf{A} \frac{1}{\det(\mathbf{A})} \text{adj}(\mathbf{A}) = \mathbf{I} \implies \mathbf{A}^{-1} = \frac{1}{\det(\mathbf{A})} \text{adj}(\mathbf{A}) \quad \square$$

Cramer's rule

Let \mathbf{A} be an $n \times n$ matrix. If $\det(\mathbf{A}) \neq 0$, then the unique solution to $\mathbf{A}\mathbf{x} = \mathbf{b}$ is

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_k \\ \vdots \\ x_n \end{bmatrix} = \frac{1}{\det(\mathbf{A})} \begin{bmatrix} \det(\mathbf{B}_1) \\ \det(\mathbf{B}_2) \\ \vdots \\ \det(\mathbf{B}_k) \\ \vdots \\ \det(\mathbf{B}_n) \end{bmatrix}$$

where \mathbf{B}_k denotes the matrix obtained by replacing the k th column of \mathbf{A} by \mathbf{b} .

Proof

- Since $\det(\mathbf{A}) \neq 0$, using the theorem on page 16, we have

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b} = \frac{1}{\det(\mathbf{A})} \text{adj}(\mathbf{A}) \mathbf{b} = \frac{1}{\det(\mathbf{A})} \mathbf{C}^T \mathbf{b}$$

where \mathbf{C} is the cofactor matrix.

Proof

- Considering the k th term, we have

$$x_k = \frac{1}{\det(\mathbf{A})} \sum_{j=1}^n b_j C_{jk} = \frac{\det(\mathbf{B}_k)}{\det(\mathbf{A})}$$

where the sum is the cofactor expansion along the k column of

$$\det(\mathbf{B}_k) = \begin{bmatrix} a_{11} & a_{12} & \cdots & b_1 & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & b_2 & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & b_n & \cdots & a_{nn} \end{bmatrix} \quad \square$$

Exercise

Find $\det(\mathbf{A})$, where $\mathbf{A} = \begin{bmatrix} 2 & 1 & 3 \\ 4 & 2 & 1 \\ 6 & -3 & 4 \end{bmatrix}$, first by using the definition, then by the cofactor expansion. Count the number of additions and multiplications used.

Solution

- In general, by definition, we have $n!$ nonzero terms

$$\begin{aligned}\det(\mathbf{A}) &= \sum_{k_1=1}^3 \sum_{k_2=1}^3 \sum_{k_3=1}^3 \varepsilon_{k_1 k_2 k_3} a_{1k_1} a_{2k_2} a_{3k_3} \\ &= 2 \cdot 2 \cdot 4 - 2 \cdot 1 \cdot (-3) \\ &\quad - 1 \cdot 4 \cdot 4 + 1 \cdot 1 \cdot 6 \\ &\quad 3 \cdot 4 \cdot (-3) - 3 \cdot 2 \cdot 6 = -60\end{aligned}$$

- Notice 5 additions and 12 multiplications are used for this 3×3 matrix.
- According to the expansion by cofactors along the first row, we have

$$\det(\mathbf{A}) = 2 \cdot (2 \cdot 4 + 1 \cdot 3) - 1 \cdot (4 \cdot 4 - 1 \cdot 6) + 3 \cdot (-4 \cdot 3 - 2 \cdot 6) = -60$$

- Notice 5 additions and 9 multiplications are used in this approach.

- That illustrates why we don't use the definition to compute determinants.

Q: Is there a better way than the cofactor expansion if we are forced to compute

$$\det(\mathbf{A})$$

Q: What do we know regarding the determinant of an upper triangular matrix?

Q: How can we exploit the fact that we only need 2 multiplications to compute

$$\det(\mathbf{U}) \quad \text{where } \mathbf{U} \text{ is upper triangular and } \mathbf{U} \sim \mathbf{A},$$

and its link to $\det(\mathbf{A})$ when we need to compute the determinant of

$$\mathbf{A} = \begin{bmatrix} 2 & 1 & 3 \\ 4 & 2 & 1 \\ 6 & -3 & 4 \end{bmatrix}$$

Q: How many additions and multiplications do we need in the elimination step?

- The elimination step consists of finding the necessary row operations

$$\begin{aligned} \mathbf{A} \sim \mathbf{E}_{(-3)1,3} \mathbf{E}_{(-2)1,2} \begin{bmatrix} 2 & 1 & 3 \\ 4 & 2 & 1 \\ 6 & -3 & 4 \end{bmatrix} &\sim \mathbf{E}_{2,3} \begin{bmatrix} 2 & 1 & 3 \\ 0 & 0 & -5 \\ 0 & -6 & -5 \end{bmatrix} \\ &\sim \begin{bmatrix} 2 & 1 & 3 \\ 0 & -6 & -5 \\ 0 & 0 & -5 \end{bmatrix} = \mathbf{U} \\ \implies \mathbf{A} &= \mathbf{E}_{(2)1,2} \mathbf{E}_{(3)1,3} \mathbf{E}_{3,2} \mathbf{U} \end{aligned}$$

in which 4 additions and 6 multiplications are needed for this particular case.

- The next step involves finding $\det(\mathbf{U})$, and making “corrections”

$$\det(\mathbf{A}) = -\det(\mathbf{U}) = -2 \cdot (-6) \cdot (-5) = -60$$

- In this case, only 4 additions and 8 multiplications are needed in total.

Q: How many multiplications do we need in total for an **arbitrary** 3×3 matrix?

- Many theorems to do with determinants can be proved using of the cofactor expansion. But it is **also not** used to find the determinant of a big matrix.
- Below gives the number of arithmetic operations involved in each method.

n	Leibiniz		Laplace		Gauss	
	+	\times	+	\times	+	\times
2	1	2	1	2	1	3
3	5	12	5	9	5	10
4	23	72	23	40	14	23
5	119	480	119	205	30	44
10	3,628,799	32,659,200	3,628,799	6,235,300	285	339

- Note computing the determinant of a 50×50 matrix directly would take a computer 10^{40} years! that is, more than 10^{30} times the age of the universe!
- So don't blindly do it on your laptop! Often it can be avoided altogether.

Q: How about finding the solution of a large square nonsingular linear system?

$$\mathbf{Ax} = \mathbf{b}$$