Lecture 12: Determinants

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- We will see several nice properties of determinants that interact meaningfully with matrix operations
- Q Learn how row operations alter the determinant of a matrix
- Learn Cramer's rule which allows us to solve matrix equations by computing determinants

Yesterday we saw that the determinant of a square matrix is the number $% \left({{{\mathbf{x}}_{i}}} \right)$

$$\det A = \sum_{i=1}^n a_{1,i} C_{1,i}$$

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or along the *j*th column

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One of our goals is to determine some tricks for computing the determinant of a matrix. $\langle \Box \rangle \langle \Box \rangle \langle \Box \rangle \langle \Box \rangle \langle \Box \rangle$

The elementary row operations each change the determinant in a prescribed way.

Example

Let
$$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$
. The determinant of A is 1. If we interchange the rows of A , we obtain $E_1A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$, which has determinant -1 .

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If
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If $A = \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix}$ then the determinant of A is -1. If we add 4 times row 1 to row 2, then we obtain $E_2A = \begin{bmatrix} 1 & 2 \\ 6 & 11 \end{bmatrix}$.

If $A = \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix}$ then the determinant of A is -1. If we add 4 times row 1 to row 2, then we obtain $E_2A = \begin{bmatrix} 1 & 2 \\ 6 & 11 \end{bmatrix}$. The determinant of the new matrix is -1 also.

Example If $A = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 2 & 7 \\ 0 & 0 & 9 \end{bmatrix}$ then the determinant of A is just the product of the diagonal entries.

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 $E_3A = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 20 & 70 \\ 0 & 0 & 9 \end{bmatrix}$

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Let A be a square matrix.

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Note that if you scale the entire matrix A by the same constant k, that has the same effect as scaling each of the n rows in order: hence det $(kA) = k^n \det A$. You can compute determinants by reducing to echelon form (upper triangular) and then using the property that the determinant of a triangular matrix is the product of the diagonal elements.

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. Then $|A| = 100 \begin{vmatrix} 1 & 2 \\ 2 & 3 \end{vmatrix}$

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Example
Let
$$A = \begin{bmatrix} 100 & 200 \\ 2 & 3 \end{bmatrix}$$
. Then $|A| = 100 \begin{vmatrix} 1 & 2 \\ 2 & 3 \end{vmatrix} = -100$.

More examples: row reducton and determinants

$$A = \begin{vmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 5 \\ 3 & 4 & 5 & 6 \\ 4 & 5 & 6 & 7 \end{vmatrix}$$

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So $|A| = 0$.

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You can compute the determinant of a matrix by row reducing to echelon form.

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Let A be an $n \times n$ matrix and let U be an echelon form computed with exactly r row interchanges. Let $u_{11}, u_{22}, \ldots, u_{nn}$ be the diagonal entries in U.

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Theorem

Let A be an $n \times n$ matrix and let U be an echelon form computed with exactly r row interchanges. Let $u_{11}, u_{22}, \ldots, u_{nn}$ be the diagonal entries in U. Then $|A| = (-1)^r u_{11} u_{22} \ldots u_{nn}$.

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Proof.

None of the row replacement operations change the determinant, so we only have to keep track of the effect of the scaling and interchange operations.

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Proof.

None of the row replacement operations change the determinant, so we only have to keep track of the effect of the scaling and interchange operations. You can put a matrix in echelon form without any scaling. So the only thing to keep track of is the interchanges. So $|A| = (-1)^r |U|$. As the echelon form is upper triangular, its determinant is the product of its diagonal entries $|U| = u_{11}u_{22}\dots u_{nn}$

Suppose that we have a matrix A and we know that A can be reduced to the echelon form

$$U = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 7 & -1 \\ 0 & 0 & 3 \end{bmatrix}$$

using some number of replacement operations and one row interchange.

Suppose that we have a matrix A and we know that A can be reduced to the echelon form

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using some number of replacement operations and one row interchange. What is |A|? We use the previous theorem:

$$|A| = (-1)|U| = -(1)(7)(3) = -21.$$

We see that if A is $n \times n$ and U is an echelon form for A which we obtained using replacements and r interchanges then $|A| = (-1)^r |U| = (-1)^r u_{11} \dots u_{nn}.$

We see that if A is $n \times n$ and U is an echelon form for A which we obtained using replacements and r interchanges then $|A| = (-1)^r |U| = (-1)^r u_{11} \dots u_{nn}$. A has n pivots if and only if all the entries u_{11}, \dots, u_{nn} are nonzero. We see that if A is $n \times n$ and U is an echelon form for A which we obtained using replacements and r interchanges then $|A| = (-1)^r |U| = (-1)^r u_{11} \dots u_{nn}$. A has n pivots if and only if all the entries u_{11}, \dots, u_{nn} are nonzero.

Theorem

Let A be a square matrix. Then A is invertible if and only if $\det A \neq 0$

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Theorem

Let A be a square matrix. Then A is invertible if and only if det $A \neq 0$

Fact

This meshes with the 2×2 case.

Example				
Compute	$\begin{vmatrix} 1 \\ -1 \\ 0 \\ 0 \end{vmatrix}$	2 3 5 —5	0 2 1 2	1 2 1 2 2
<i>A</i>				

Example				
Compute	$\begin{vmatrix} 1 \\ -1 \\ 0 \\ 0 \end{vmatrix}$	2 (3 2 5 2 -5 2) 1 2 2 L 1 2 2	. First clean out beneath the first pivot
$ A = \begin{vmatrix} 1\\0\\0\\0 \end{vmatrix}$	2 5 5 —5	0 1 2 3 1 1 2 2		

Example						
Compute	$\begin{vmatrix} 1 \\ -1 \\ 0 \\ 0 \end{vmatrix}$	2 3 5 —5	0 2 1 2	1 2 1 2	rst c	lean out beneath the first pivot
$ A = \begin{vmatrix} 1\\0\\0\\0 \end{vmatrix}$	2 5 5 —5	0 1 2 3 1 1 2 2	=	$\begin{vmatrix} 1 \\ -5 \end{vmatrix}$	2 1 2	3 1 2

Example							
Compute	$\begin{vmatrix} 1 \\ -1 \\ 0 \\ 0 \end{vmatrix}$	2 3 5 —5	0 2 1 2	1 2 1 2	rst c	lean out be	neath the first pivot
$ A = \begin{vmatrix} 1 \\ 0 \\ 0 \\ 0 \end{vmatrix}$	2 5 5 —5	0 1 2 3 1 1 2 2	=	$\begin{vmatrix} 5\\5\\-5 \end{vmatrix}$	2 1 2	$\begin{vmatrix} 3\\1\\2 \end{vmatrix} = \begin{vmatrix} 0\\0\\-5 \end{vmatrix}$	4 5 3 3 2 2

Example							
Compute	$\begin{vmatrix} 1 \\ -1 \\ 0 \\ 0 \end{vmatrix}$	2 3 5 5	0 2 1 2	1 2 1 2	. First clean out beneath the first pivot		
$ A = \begin{vmatrix} 1 \\ 0 \\ 0 \\ 0 \end{vmatrix}$	2 5 5 -5	0 1 2 3 1 1 2 2	=	1	$\begin{vmatrix} 5 & 2 & 3 \\ 5 & 1 & 1 \\ -5 & 2 & 2 \end{vmatrix} = \begin{vmatrix} 0 & 4 & 5 \\ 0 & 3 & 3 \\ -5 & 2 & 2 \end{vmatrix}$ Now we upon to get		
expand along the first column to get							
$ A = (-5) \begin{vmatrix} 4 & 5 \\ 3 & 3 \end{vmatrix} = (-5)(-3) = 15.$							

Let's compute
$$\begin{vmatrix} -1 & 2 & 3 & 0 \\ 3 & 4 & 3 & 0 \\ 5 & 4 & 6 & 6 \\ 4 & 2 & 4 & 3 \end{vmatrix}$$

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$$\begin{vmatrix} -1 & 2 & 3 & 0 \\ 3 & 4 & 3 & 0 \\ 5 & 4 & 6 & 6 \\ 4 & 2 & 4 & 3 \end{vmatrix}$$
 We see that the fourth column has

some zeros. We can get more by subtracting 2 times the fourth fow from the third row. Then we take a cofactor expansion

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$$= 3 \begin{vmatrix} -1 & 2 & 3 \\ 3 & 4 & 3 \\ -3 & 0 & -2 \end{vmatrix}$$

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If A is a square matrix then det $A = \det A^T$. That is, a matrix and its transpose have the same determinant.

Proof.

You just flip the cofactor expansions: if you compute the determinant of A by cofactor expansion along row j, compute the determinant of A^T along its jth column, and vice versa.

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Any column operation (interchange columns, replace column, scale column) has the same effect on the determinant as the corresponding row operation. So replacing columns does not change the determinant, interchanging two columns multiplies the determinant by -1, scaling a column scales the determinant.

Theorem

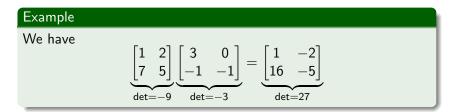
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$$\underbrace{\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}}_{\text{det}=-1} + \underbrace{\begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}}_{\text{det}=-1} = \underbrace{\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}}_{\text{det}=0}$$

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Also it is not true that det $cA = c \det A$ for all matrices A and scalars c. Instead det $cA = c^n \det A$.

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Theorem (Cramer's rule)

Let A be an invertible $n \times n$ matrix. For any $\mathbf{b} \in \mathbb{R}^n$, the unique solution $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$ of $A\mathbf{x} = \mathbf{b}$ has entries given by $x_i = \frac{\det A_i(\mathbf{b})}{\det A}$

Example

Let
$$A = \begin{bmatrix} 1 & 8 \\ 2 & 17 \end{bmatrix}$$
 (which has determinant 1) and $\mathbf{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$.

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Thus the unique solution is $\mathbf{x} = \begin{bmatrix} 9 \\ -1 \end{bmatrix}$.

When you are studying geometry you study parallelograms in \mathbb{R}^2 and parallelepipeds in \mathbb{R}^3 . A pair of vectors determines a parallelogram and a triple of vectors determines a parallelepiped. When you are studying geometry you study parallelograms in \mathbb{R}^2 and parallelepipeds in \mathbb{R}^3 . A pair of vectors determines a parallelogram and a triple of vectors determines a parallelepiped.

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Let A be 2×2 . If we treat the columns of A as the vertices in a paralleogram that are adjacent to the origin, then the area of that parallelogram is $|\det A|$.

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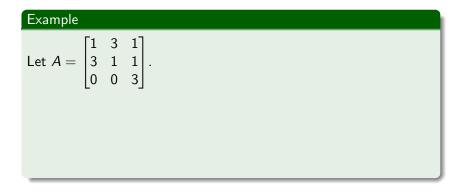
Theorem

Let A be 2×2 . If we treat the columns of A as the vertices in a paralleogram that are adjacent to the origin, then the area of that parallelogram is $|\det A|$. If A is 3×3 and we treat the columns of A as the vertices in a parallelepiped that are adjacent the origin, then the area of that parallelepiped is $|\det A|$.

Let $A = \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix}$. Then the columns of A define a parallelogram with vertices at (0,0), (1,2), (1,0), (2,2).

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Let
$$A = \begin{bmatrix} 1 & 3 & 1 \\ 3 & 1 & 1 \\ 0 & 0 & 3 \end{bmatrix}$$
. The columns of A define a parallelepiped with vertices $(0, 0, 0), (1, 3, 0), (3, 1, 0), (4, 4, 0), (1, 1, 3), (2, 4, 3), (4, 2, 3), (5, 5, 3)$

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$$\det A | = 24.$$