# Chapter 4

**Determinants** 

## Section 4.1

Determinants: Definition

#### Orientation

### Recall: This course is about learning to:

- Solve the matrix equation Ax = b
   We've said most of what we'll say about this topic now.
- Solve the matrix equation  $Ax = \lambda x$  (eigenvalue problem) We are now aiming at this.
- Almost solve the equation Ax = b
  This will happen later.

The next topic is determinants.

This is a completely magical function that takes a square matrix and gives you a number.

It is a very complicated function—the formula for the determinant of a  $10 \times 10$  matrix has 3,628,800 summands—so instead of writing down the formula, we'll give other ways to compute it.

Today is mostly about the *theory* of the determinant; in the next lecture we will focus on *computation*.

## A Definition of Determinant

#### Definition

The determinant is a function determinants are only for square matrices!

$$\mathsf{det} \colon \{n \times n \; \mathsf{matrices}\} \longrightarrow \mathbf{R}$$

with the following properties:

- 1. If you do a row replacement on a matrix, the determinant doesn't change.
- 2. If you scale a row by c, the determinant is multiplied by c.
- 3. If you swap two rows of a matrix, the determinant is multiplied by -1.
- 4.  $\det(I_n) = 1$ .

## Example:

$$\begin{pmatrix} 2 & 1 \\ 1 & 4 \end{pmatrix} \xrightarrow{R_1 \longleftrightarrow R_2} \begin{pmatrix} 1 & 4 \\ 2 & 1 \end{pmatrix}$$

$$R_2 = R_2 - 2R_1$$

$$R_2 = R_2 \div -7$$

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$$R_1 = R_1 - 4R_2$$

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$$R_2 = R_2 \div -7$$

$$R_2 = R_2 \div -7$$

$$R_3 = R_1 - 4R_2$$

$$R_4 = R_1 - 4R_2$$

$$R_5 =$$

## A Definition of Determinant

#### Definition

The **determinant** is a function determinants are only for square matrices!

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with the following properties:

- 1. If you do a row replacement on a matrix, the determinant doesn't change.
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- 3. If you swap two rows of a matrix, the determinant is multiplied by -1.
- 4.  $\det(I_n) = 1$ .

This is a *definition* because it tells you how to compute the determinant: row reduce!

It's not at all obvious that you get the same determinant if you row reduce in two different ways, but this is magically true!

## Special Cases

#### Special Case 1

If A has a zero row, then det(A) = 0.

Why?

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 0 & 0 \\ 7 & 8 & 9 \end{pmatrix} \xrightarrow{R_2 = -R_2} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 0 & 0 \\ 7 & 8 & 9 \end{pmatrix}$$

The determinant of the second matrix is negative the determinant of the first (property 3), so

$$\det\begin{pmatrix} 1 & 2 & 3 \\ 0 & 0 & 0 \\ 7 & 8 & 9 \end{pmatrix} = -\det\begin{pmatrix} 1 & 2 & 3 \\ 0 & 0 & 0 \\ 7 & 8 & 9 \end{pmatrix}.$$

This implies the determinant is zero.

## Special Cases

## Special Case 2

If A is upper-triangular, then the determinant is the product of the diagonal entries:

$$\det\begin{pmatrix} a & \star & \star \\ 0 & b & \star \\ 0 & 0 & c \end{pmatrix} = abc.$$

**Upper-triangular** means the only nonzero entries are on or above the diagonal.

## Why?

- ▶ If one of the diagonal entries is zero, then the matrix has fewer than *n* pivots, so the RREF has a row of zeros. (Row operations don't change whether the determinant is zero.)
- Otherwise,

$$\begin{pmatrix} a & \star & \star \\ 0 & b & \star \\ 0 & 0 & c \end{pmatrix} \xrightarrow{\text{scale by}} \xrightarrow{a^{-1}, b^{-1}, c^{-1}} \begin{pmatrix} 1 & \star & \star \\ 0 & 1 & \star \\ 0 & 0 & 1 \end{pmatrix} \xrightarrow{\text{replacements}} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
 
$$\det = abc \iff \det = 1 \iff \det = 1$$

#### **Theorem**

Let A be a square matrix. Suppose you do some number of row operations on A to get a matrix B in row echelon form. Then

$$det(A) = (-1)^r \frac{(product of the diagonal entries of B)}{(product of the scaling factors)},$$

where r is the number of row swaps.

Why? Since B is in REF, it is upper-triangular, so its determinant is the product of its diagonal entries. You changed the determinant by  $(-1)^r$  and the product of the scaling factors when going from A to B.

#### Remark

This is generally the fastest way to compute a determinant of a large matrix, either by hand or by computer.

Row reduction is  $O(n^3)$ ; cofactor expansion (next time) is  $O(n!) \sim O(n^n \sqrt{n})$ .

This is important in real life, when you're usually working with matrices with a gazillion columns.

# Computing Determinants Example

# Computing Determinants 2 × 2 Example

Let's compute the determinant of  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ , a general  $2 \times 2$  matrix.

▶ If a = 0, then

$$\det\begin{pmatrix} a & b \\ c & d \end{pmatrix} = \det\begin{pmatrix} 0 & b \\ c & d \end{pmatrix} = -\det\begin{pmatrix} c & d \\ 0 & b \end{pmatrix} = -bc.$$

Otherwise,

$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = a \cdot \det \begin{pmatrix} 1 & b/a \\ c & d \end{pmatrix} = a \cdot \det \begin{pmatrix} 1 & b/a \\ 0 & d - c \cdot b/a \end{pmatrix}$$
$$= a \cdot 1 \cdot (d - bc/a) = ad - bc.$$

In both cases, the determinant magically turns out to be

$$\det\begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc.$$

Poll

True or false:

- (a) Row operations can change the determinant of a matrix.
- (b) Row operations can change whether the determinant of a matrix is equal to zero.

- (a) True: scaling and row swaps change the determinant by a nonzero number and by -1, respectively.
- (b) False: all row operations multiply the determinant by a nonzero number.

## Determinants and Invertibility

#### **Theorem**

A square matrix A is invertible if and only if det(A) is nonzero.

#### Why?

- ▶ If A is invertible, then its reduced row echelon form is the identity matrix, which has determinant equal to 1.
- ▶ If A is not invertible, then its reduced row echelon form has a zero row, hence has zero determinant.
- ▶ Doing row operations doesn't change whether the determinant is zero.

## **Determinants and Products**

#### **Theorem**

If A and B are two  $n \times n$  matrices, then

$$\det(AB) = \det(A) \cdot \det(B).$$

Why? If B is invertible, we can define

$$f(A) = \frac{\det(AB)}{\det(B)}.$$

Note  $f(I_n) = \det(I_n B)/\det(B) = 1$ . Check that f satisfies the same properties as det with respect to row operations. So

$$\det(A) = f(A) = \frac{\det(AB)}{\det(B)} \implies \det(AB) = \det(A)\det(B).$$

What about if B is not invertible?

#### **Theorem**

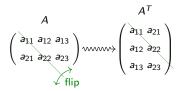
If A is invertible, then  $det(A^{-1}) = \frac{1}{det(A)}$ .

Why? 
$$I_n = AB \implies 1 = \det(I_n) = \det(AB) = \det(A) \det(B)$$
.

## Transposes

Review

Recall: The transpose of an  $m \times n$  matrix A is the  $n \times m$  matrix  $A^T$  whose rows are the columns of A. In other words, the ij entry of  $A^T$  is  $a_{ji}$ .



## **Determinants and Transposes**

#### Theorem

If A is a square matrix, then

$$\det(A) = \det(A^T),$$

where  $A^T$  is the transpose of A.

Example: 
$$\det \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} = \det \begin{pmatrix} 1 & 3 \\ 2 & 4 \end{pmatrix}$$
.

As a consequence, det behaves the same way with respect to *column* operations as row operations.

If A has a zero column, then det(A) = 0.

## Corollary

The determinant of a *lower*-triangular matrix is the product of the diagonal entries.

(The transpose of a lower-triangular matrix is upper-triangular.)

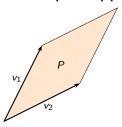
## Section 4.3

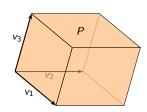
Determinants and Volumes

Now we discuss a completely different description of (the absolute value of) the determinant, in terms of volumes.

This is a crucial component of the change-of-variables formula in multivariable calculus.

The columns  $v_1, v_2, \ldots, v_n$  of an  $n \times n$  matrix A give you n vectors in  $\mathbb{R}^n$ . These determine a **parallelepiped** P.





#### **Theorem**

Let A be an  $n \times n$  matrix with columns  $v_1, v_2, \ldots, v_n$ , and let P be the parallelepiped determined by A. Then

(volume of 
$$P$$
) =  $|\det(A)|$ .

#### **Theorem**

Let A be an  $n \times n$  matrix with columns  $v_1, v_2, \ldots, v_n$ , and let P be the parallelepiped determined by A. Then

(volume of 
$$P$$
) =  $|\det(A)|$ .

Sanity check: the volume of P is zero  $\iff$  the columns are *linearly dependent* (P is "flat")  $\iff$  the matrix A is not invertible.

Why is the theorem true? You only have to check that the volume behaves the same way under row operations as | det | does.

Note that the volume of the unit cube (the parallelepiped defined by the identity matrix) is  $1. \,$ 

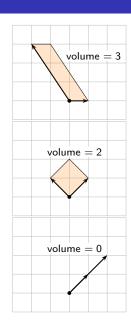
Examples in  $\ensuremath{\mbox{\bf R}}^2$ 

$$\det\begin{pmatrix}1 & -2\\0 & 3\end{pmatrix} = 3$$

$$\det\begin{pmatrix} -1 & 1 \\ 1 & 1 \end{pmatrix} = -2$$

(Should the volume really be -2?)

$$\det\begin{pmatrix}1&2\\1&2\end{pmatrix}=0$$



#### **Theorem**

Let A be an  $n \times n$  matrix with columns  $v_1, v_2, \ldots, v_n$ , and let P be the parallelepiped determined by A. Then

(volume of 
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) =  $|\det(A)|$ .

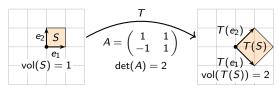
This is even true for curvy shapes, in the following sense.

#### **Theorem**

Let A be an  $n \times n$  matrix, and let T(x) = Ax. If S is any region in  $\mathbb{R}^n$ , then

(volume of 
$$T(S)$$
) =  $|\det(A)|$  (volume of  $S$ ).

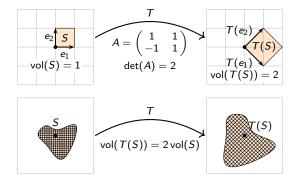
If S is the unit cube, then T(S) is the parallelepiped defined by the columns of A, since the columns of A are  $T(e_1), T(e_2), \ldots, T(e_n)$ . In this case, the second theorem is the same as the first.



#### **Theorem**

Let A be an  $n \times n$  matrix, and let T(x) = Ax. If S is any region in  $\mathbf{R}^n$ , then (volume of T(S)) =  $|\det(A)|$  (volume of S).

For curvy shapes, you break S up into a bunch of tiny cubes. Each one is scaled by  $|\det(A)|$ ; then you use *calculus* to reduce to the previous situation!



# Determinants and Volumes Example

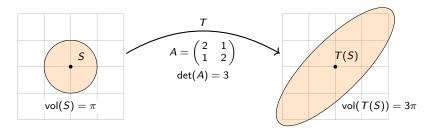
#### **Theorem**

Let A be an  $n \times n$  matrix, and let T(x) = Ax. If S is any region in  $\mathbf{R}^n$ , then  $(\text{volume of } T(S)) = |\det(A)|$  (volume of S).

Example: Let S be the unit disk in  $\mathbb{R}^2$ , and let T(x) = Ax for

$$A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}.$$

Note that det(A) = 3.



## Magical Properties of the Determinant

- 1. There is one and only one function det: {square matrices}  $\rightarrow$  R satisfying the properties (1)–(4) on the second slide.
- 2. A is invertible if and only if  $det(A) \neq 0$ .
- 3. The determinant of an upper- or lower-triangular matrix is the product of the diagonal entries.
- 4. If we row reduce A to row echelon form B using r swaps, then

$$det(A) = (-1)^r \frac{(product of the diagonal entries of B)}{(product of the scaling factors)}.$$

- 5. det(AB) = det(A) det(B) and  $det(A^{-1}) = det(A)^{-1}$ .
- 6.  $\det(A) = \det(A^T)$ .
- 7.  $|\det(A)|$  is the volume of the parallelepiped defined by the columns of A.
- 8. If A is an  $n \times n$  matrix with transformation T(x) = Ax, and S is a subset of  $\mathbb{R}^n$ , then the volume of T(S) is  $|\det(A)|$  times the volume of S. (Even for curvy shapes S.)