

## Previous Lecture

- ◆ Matrix Multiplicative Identity
- ◆ Matrix Determinant and Inverses
- ◆ Matrix Exponents
- ◆ Solving Systems w/Matrix Inverses
- ◆ Matrix invertibility



## 3.6: Determinants and Inverses when $n > 2$

Suppose I give you a physical system (e.g., population)  $\mathbf{A}\vec{x}_n = \vec{x}_{n+1}$ , where  $\vec{x}_n$  represents the current population of various species in the forest, matrix  $\mathbf{A}$  encodes how the variables interact (predation, survival rates, migration, reproduction), and  $\vec{x}_{n+1}$  is the population in the following year.



**Problem:** If you tally the population of these species today  $\vec{x}_n$ , can you *uniquely* determine what the population was last year  $\vec{x}_{n-1}$ ? Could *different* past populations levels have led to the same present levels? The determinant is the number that tells us which situation we're in.

The system  $\mathbf{A}\vec{x} = \vec{b}$  always determines the future, but determinants tell us **whether the system remembers the past**.

<https://youtu.be/lp3X9L0h2dk?si=XtKN95m0ybXaDRDU&t=308>

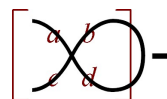
Mathematically, remembering the past means having the ability to calculate an inverse  $\mathbf{A}^{-1}$ .

Below, we'll also learn how to use determinants to solve large linear systems (Cramer's rule).

## Notation/Calculation

The determinant of  $\mathbf{A}$  is a number notated as:  $\det \mathbf{A} = |\mathbf{A}| = |a_{ij}| = \det \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix}$ .

Recall the determinant of a  $2 \times 2$  matrix is calculated as:  $|\mathbf{A}| = ad - bc$ .



**Determinants of  $n > 2$  Square Matrices:** For these, we need some new tools.

e.g.,  $\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$

We'll make use of a "checkerboard" of signs:  $\begin{vmatrix} + & - & + & \dots \\ - & + & - & \dots \\ + & - & + & \vdots \\ \vdots & \vdots & \dots & \ddots \end{vmatrix}$ .

**Definition:** the **minor** ( $M_{ij}$ ) for a matrix's  $a_{ij}$ th entry is found by ignoring the  $i$ th row and the  $j$ th column,

and taking the determinant of the remaining elements. So,  $M_{12} = \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}$ .

**Definition:** the **cofactor** ( $A_{ij}$ ) for a matrix's  $a_{ij}$ th entry is  $A_{ij} := (\text{checkerboard sign})M_{ij}$ .

$$\text{So, } A_{12} = (-1)M_{12} = - \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}.$$

**Definition:** Finally, the determinant of an  $n \times n$  matrix  $\mathbf{A} = [a_{ij}]$  is defined as:

$$\det \mathbf{A} = a_{11}A_{11} + a_{12}A_{12} + \dots + a_{1n}A_{1n}, \text{ where } A_{ij} \text{ is the } ij \text{th cofactor.}$$

**Example:** For a  $3 \times 3$  matrix, this would be...

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}, \text{ and similarly for } n \geq 2.$$

Or, 
$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11}M_{11} - a_{12}M_{12} + a_{13}M_{13} = a_{11}A_{11} + a_{12}A_{12} + a_{13}A_{13}.$$

Notice how easy these calculations would be if some of the  $a_{ij}$  were 0.

Later, we'll find out how to arrange for this to happen.



Above, we "expanded" our determinant along the **top row of the original matrix**.

However, you can calculate the determinant using *any* row or column.

Indeed, if there is a row or column with entries that are zero (or that you can manipulate to be zero), it is strategic to use that row or column to simplify your determinant calculation.

**Corollary:** If  $\mathbf{A}$  contains either a zero column vector or a zero row vector, then  $|\mathbf{A}| = 0$ .

This follows since we can choose this zero row or column to do our expansion, and the resulting expression will have cofactor coefficients of zero.

## Determinant Properties

♦  $|\mathbf{AB}| = |\mathbf{A}||\mathbf{B}|$ , or equivalently,  $\det(\mathbf{AB}) = \det \mathbf{A} \cdot \det \mathbf{B}$ .

As a direct result of the above property, we can prove  $|\mathbf{A}^{-1}| = \frac{1}{|\mathbf{A}|}$ . (recall  $\mathbf{A}$  invertible implies  $|\mathbf{A}| \neq 0$ )

**Proof:** Observe that the result assumes  $\mathbf{A}^{-1}$  and  $\mathbf{A}$  exist, and therefore we can say:

$\mathbf{AA}^{-1} = \mathbf{I}$ . Then, using the determinant property on this equation:

$|\mathbf{AA}^{-1}| = |\mathbf{A}||\mathbf{A}^{-1}| = |\mathbf{I}| = 1$ . And dividing by  $|\mathbf{A}|$  gives us:

$$|\mathbf{A}^{-1}| = \frac{1}{|\mathbf{A}|}. \quad \blacksquare$$

**Recall the Important Property:**  $\mathbf{A}$  is invertible if  $|\mathbf{A}| \neq 0$ .

## Determinant "Method of Elimination": (calculating determinants more easily)

We can: **Pull constants from rows or columns** (bolded text indicates changes, assume  $k$  is some constant.)

$$\begin{vmatrix} \mathbf{ka}_{11} & \mathbf{ka}_{12} & \mathbf{ka}_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \mathbf{k} \begin{vmatrix} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}, \text{ and}$$
$$\begin{vmatrix} a_{11} & \mathbf{ka}_{12} & a_{13} \\ a_{21} & \mathbf{ka}_{22} & a_{23} \\ a_{31} & \mathbf{ka}_{32} & a_{33} \end{vmatrix} = \mathbf{k} \begin{vmatrix} a_{11} & \mathbf{a}_{12} & a_{13} \\ a_{21} & \mathbf{a}_{22} & a_{23} \\ a_{31} & \mathbf{a}_{32} & a_{33} \end{vmatrix}.$$

**Proof for  $n = 3$ :** (done in class).

**Switch columns or rows by changing sign.**

$$\begin{vmatrix} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} \\ \mathbf{a}_{21} & \mathbf{a}_{22} & \mathbf{a}_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = - \begin{vmatrix} \mathbf{a}_{21} & \mathbf{a}_{22} & \mathbf{a}_{23} \\ \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}, \text{ and}$$
$$\begin{vmatrix} \mathbf{a}_{11} & \mathbf{a}_{12} & a_{13} \\ \mathbf{a}_{21} & \mathbf{a}_{22} & a_{23} \\ \mathbf{a}_{31} & \mathbf{a}_{32} & a_{33} \end{vmatrix} = - \begin{vmatrix} \mathbf{a}_{12} & \mathbf{a}_{11} & a_{13} \\ \mathbf{a}_{22} & \mathbf{a}_{21} & a_{23} \\ \mathbf{a}_{32} & \mathbf{a}_{31} & a_{33} \end{vmatrix}.$$

**Proof for  $n = 3$ .** (try this one yourself, take the determinant of the matrices in the above equation).

**Can add a multiple of a row to another row, and can do the same with columns.**

$$\begin{vmatrix} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} \\ b & c & d \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} \mathbf{a}_{11}+\mathbf{kb} & \mathbf{a}_{12}+\mathbf{kc} & \mathbf{a}_{13}+\mathbf{kd} \\ b & c & d \\ a_{31} & a_{32} & a_{33} \end{vmatrix}, \text{ and}$$
$$\begin{vmatrix} \mathbf{a}_{11} & a_{12} & b \\ \mathbf{a}_{21} & a_{22} & c \\ \mathbf{a}_{31} & a_{32} & d \end{vmatrix} = \begin{vmatrix} \mathbf{a}_{11}+\mathbf{kb} & a_{12} & b \\ \mathbf{a}_{21}+\mathbf{kc} & a_{22} & c \\ \mathbf{a}_{31}+\mathbf{kd} & a_{32} & d \end{vmatrix}.$$

**Proof for  $n = 3$ .** (try this one yourself, take the determinant of the matrices in the above equation).

**Identical rows or columns means determinant is zero.**

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ b & c & d \\ b & c & d \end{vmatrix} = 0, \text{ and } \begin{vmatrix} a_{11} & a & a \\ a_{21} & b & b \\ a_{31} & c & c \end{vmatrix} = 0.$$

**Proof for  $n = 3$ :** (done in class).

**Determinant of (upper or lower) triangular matrices is the product of the main diagonal.**

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22} & a_{23} \\ 0 & 0 & a_{33} \end{vmatrix} = \begin{vmatrix} a_{11} & 0 & 0 \\ a_{21} & a_{22} & 0 \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11} \cdot a_{22} \cdot a_{33}.$$

**Proof for  $n = 3$ :** (done in class).

Therefore  $\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ 0 & 0 & a_{23} \\ 0 & 0 & a_{33} \end{vmatrix} = 0.$

**Example:** Use the method of elimination (*determinant* row/column operations) to evaluate:  $\begin{vmatrix} -2 & 5 & 4 \\ 5 & 3 & 1 \\ 1 & 4 & 5 \end{vmatrix}.$

$$\Rightarrow c_2 + (-4c_1) \Rightarrow \begin{vmatrix} -2 & 13 & 4 \\ 5 & -17 & 1 \\ 1 & 0 & 5 \end{vmatrix} \Rightarrow c_3 + (-5c_1) \Rightarrow \begin{vmatrix} -2 & 13 & 14 \\ 5 & -17 & -24 \\ 1 & 0 & 0 \end{vmatrix}$$

(note that the above operations can only be performed on determinants, not matrices!)

$$\Rightarrow - \begin{vmatrix} -2 & 13 & 14 \\ 1 & 0 & 0 \\ 5 & -17 & -24 \end{vmatrix} \Rightarrow + \begin{vmatrix} 1 & 0 & 0 \\ -2 & 13 & 14 \\ 5 & -17 & -24 \end{vmatrix} = +1 \begin{vmatrix} 13 & 14 \\ -17 & -24 \end{vmatrix} = -74. \text{ (Checkerboard and Fish)}$$

(don't need these steps here, but thought I'd show row-swapping/sign-changes)

In addition to the  $[\mathbf{A} \mid \mathbf{I}] \rightarrow [\mathbf{I} \mid \mathbf{A}^{-1}]$  method, there's another method to find inverses when  $n > 2$ , but we need another tool: *transposes*.

**Matrix Transpose**

If  $\mathbf{A} = \begin{bmatrix} \mathbf{c} & \mathbf{a} & \mathbf{t} \\ p & e & n \\ \mathbf{m} & \mathbf{o} & \mathbf{m} \end{bmatrix}$ , then  $\mathbf{A}^T = \begin{bmatrix} \mathbf{c} & p & \mathbf{m} \\ \mathbf{a} & e & \mathbf{o} \\ \mathbf{t} & n & \mathbf{m} \end{bmatrix}.$

**Transpose Properties**

- ◆  $(\mathbf{A}^T)^T = \mathbf{A}$
- ◆  $(\mathbf{A} + \mathbf{B})^T = \mathbf{A}^T + \mathbf{B}^T$
- ◆  $(c\mathbf{A})^T = c\mathbf{A}^T$
- ◆  $(\mathbf{AB})^T = \mathbf{B}^T\mathbf{A}^T$  (recall:  $(\mathbf{AB})^{-1} = \mathbf{B}^{-1}\mathbf{A}^{-1}$ )
- ◆  $\det(\mathbf{A}^T) = \det \mathbf{A}$

# Cramer's Rule

It's a way to easily solve a system  $\mathbf{A}\vec{x} = \vec{b}$  using determinants.

Given:  $\mathbf{A}^{3 \times 3} := \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$  with  $|\mathbf{A}| \neq 0$ , and given  $\vec{b} := \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$ ,

Define:  $|\mathbf{B}_1| := \begin{vmatrix} \mathbf{b}_1 & a_{12} & a_{13} \\ \mathbf{b}_2 & a_{22} & a_{23} \\ \mathbf{b}_3 & a_{32} & a_{33} \end{vmatrix}$ ,  $|\mathbf{B}_2| := \begin{vmatrix} a_{11} & \mathbf{b}_1 & a_{13} \\ a_{21} & \mathbf{b}_2 & a_{23} \\ a_{31} & \mathbf{b}_3 & a_{33} \end{vmatrix}$ ,  $|\mathbf{B}_3| := \begin{vmatrix} a_{11} & a_{12} & \mathbf{b}_1 \\ a_{21} & a_{22} & \mathbf{b}_2 \\ a_{31} & a_{32} & \mathbf{b}_3 \end{vmatrix}$ .

The solution to  $\mathbf{A}\vec{x} = \vec{b}$  is  $\vec{x} = \langle x_1, x_2, x_3 \rangle$ , where:

$x_1 = \frac{|\mathbf{B}_1|}{|\mathbf{A}|}$ ,  $x_2 = \frac{|\mathbf{B}_2|}{|\mathbf{A}|}$ ,  $x_3 = \frac{|\mathbf{B}_3|}{|\mathbf{A}|}$ , and similarly with matrices sized  $n > 3$ .

**Example:** Use Cramer's rule to solve the system:

$$\begin{aligned} 2x_1 - 5x_3 &= -3 \\ 4x_1 + 3x_3 - 5x_2 &= 3 \\ -2x_1 + x_2 + x_3 &= 1 \end{aligned}$$

$\mathbf{A}\vec{x} = \vec{b}$        $|\mathbf{A}| = \begin{vmatrix} 2 & 0 & -5 \\ 4 & -5 & 3 \\ -2 & 1 & 1 \end{vmatrix}$  (let's further simplify the column with the zero in it)

$|\mathbf{A}| = \begin{vmatrix} 2 & 0 & -5 \\ -6 & 0 & 8 \\ -2 & 1 & 1 \end{vmatrix} = -1(16 - 30) = 14 \neq 0. \quad \checkmark$

$x_1 = \frac{|\mathbf{B}_1|}{|\mathbf{A}|} = \frac{1}{14} \begin{vmatrix} -3 & 0 & -5 \\ 3 & -5 & 3 \\ 1 & 1 & 1 \end{vmatrix} = -\frac{8}{7}, \quad x_2 = \frac{|\mathbf{B}_2|}{|\mathbf{A}|} = \frac{1}{14} \begin{vmatrix} 2 & -3 & -5 \\ 4 & 3 & 3 \\ -2 & 1 & 1 \end{vmatrix} = -\frac{10}{7}, \quad x_3 = \frac{1}{14} \begin{vmatrix} 2 & 0 & -3 \\ 4 & -5 & 3 \\ -2 & 1 & 1 \end{vmatrix} = \frac{1}{7}.$

So,  $\vec{x} = [x_1 \ x_2 \ x_3]^T = \frac{1}{7}[-8 \ -10 \ 1]^T$ .

## Matrix Inverse $\mathbf{A}^{-1}$ when $n > 3$

In addition to the  $[\mathbf{A} \mid \mathbf{I}] \rightarrow [\mathbf{I} \mid \mathbf{A}^{-1}]$  method, we have the following.

Recall we learned about cofactors  $A_{mn}$ , which are signed sub-determinants of  $\mathbf{A}$ . Let's create a new matrix  $[A_{mn}]$ , which consists of these cofactors (determinants), called the *cofactor* matrix of  $\mathbf{A}$ . It turns out that if you multiply  $\mathbf{A}$  by the transpose of the cofactor matrix  $[A_{mn}]^T$ , you get:  $\mathbf{A}[A_{mn}]^T = |\mathbf{A}|\mathbf{I}$ .

This is amazing, as we now have access to an inverse. That is, all we must do is divide by  $|\mathbf{A}|$ :  $\mathbf{A} \frac{1}{|\mathbf{A}|} [A_{mn}]^T = \mathbf{I}$ . In other words,  $\frac{1}{|\mathbf{A}|} [A_{mn}]^T$  is the inverse of  $\mathbf{A}^{-1}$ . This also suggests the inverse doesn't exist if  $|\mathbf{A}| = 0$  (that is, if the transformation loses dimensions).

**Cofactor Matrix Inverse Method Theorem:** If  $\mathbf{A}$  is invertible (so first check the determinant), we can construct a **cofactor matrix**  $[A_{mn}]$ , which consists of the **cofactors** of  $\mathbf{A}$  (NOT  $\mathbf{A}$ 's components). And then:  $\mathbf{A}^{-1} = \frac{1}{|\mathbf{A}|} [A_{mn}]^T$ .

**Proof:** We need to solve  $\mathbf{A}\mathbf{X} = \mathbf{I}$  for  $\mathbf{X}$ . So, for each column vector  $\vec{x}_i$  in  $\mathbf{X}$ , use Cramer's rule to solve  $\mathbf{A}\vec{x}_i = \vec{e}_i$ . ■

Row reduction  $[\mathbf{A} \mid \mathbf{I}] \rightarrow [\mathbf{I} \mid \mathbf{A}^{-1}]$  is our calculator. It's how we *usually* compute inverses.

The cofactor (determinant) method is our microscope. It reveals the structure behind invertibility.

It tells us that inverses exist because determinants encode a matrix's geometry (how it transforms the vector space).

$$\text{If } \mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}, \text{ then } \mathbf{A}^{-1} = \frac{1}{\det \mathbf{A}} [A_{mn}]^T = \frac{1}{\det \mathbf{A}} \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}^T = \frac{1}{\det \mathbf{A}} \begin{bmatrix} A_{11} & A_{21} & A_{31} \\ A_{12} & A_{22} & A_{32} \\ A_{13} & A_{23} & A_{33} \end{bmatrix},$$

... where  $A_{mn}$  are the *cofactors* (NOT the components) of  $\mathbf{A}$ .

**Example:** Use the cofactor method to find the inverse ( $\mathbf{A}^{-1}$ ) of:  $\mathbf{A} = \begin{bmatrix} 2 & 4 & -3 \\ 2 & -3 & -1 \\ -5 & 0 & -3 \end{bmatrix}$ .

Then, use it to solve  $\mathbf{A}\vec{x} = \vec{b}$  where  $\vec{b} = [0, 1, 1]^T$ . (HINT  $\det \mathbf{A} = 107$ )

Recall:  $\mathbf{A}^{-1} = \frac{1}{|\mathbf{A}|} [A_{mn}]^T$ . **Cofactors:**  $A_{mn} = (\text{Checkerboard})(\text{Fish})$ .

So,  $A_{11} = (+)(9 - 0) = 9$ ,  $A_{12} = (-)(-6 - 5) = 11$ ,  $A_{13} = (+)(0 - 15) = -15$ ... So,  $[A_{mn}] = \begin{bmatrix} 9 & 11 & -15 \\ 12 & -21 & -20 \\ -13 & -4 & -14 \end{bmatrix}$ .

And the transpose:  $[A_{mn}]^T$  is  $\begin{bmatrix} 9 & 12 & -13 \\ 11 & -21 & -4 \\ -15 & -20 & -14 \end{bmatrix}$ , so  $\mathbf{A}^{-1} = \frac{1}{|\mathbf{A}|} [A_{mn}]^T = \frac{1}{107} \begin{bmatrix} 9 & 12 & -13 \\ 11 & -21 & -4 \\ -15 & -20 & -14 \end{bmatrix}$ .

Solution:  $\vec{x} = \frac{1}{107} \begin{bmatrix} 9 & 12 & -13 \\ 11 & -21 & -4 \\ -15 & -20 & -14 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{107} \begin{bmatrix} -1 \\ -25 \\ -34 \end{bmatrix}$ .

**Video Tutorials** (visually rich and intuitive): <https://youtu.be/uQhTuRIWMxw> <https://youtu.be/Ip3X9LOh2dk>

**Exercises 3.6** 🐟

## What did we learn?

- ◆ Calculating Determinants  $|\mathbf{A}|$  when  $n > 3$
- ◆ Calculating Determinants More Easily
- ◆ Matrix Transpose  $\mathbf{A}^T$
- ◆ Cramer's Rule:  $\mathbf{A}\vec{x} = \vec{b} \Rightarrow x_i = \frac{|\mathbf{B}_i|}{|\mathbf{A}|}$ .
- ◆ Inverting Matrices using Cofactors  $\mathbf{A}^{-1} = \frac{1}{|\mathbf{A}|}[\mathbf{A}_{mn}]^T$



Prepared by Dr. Jodin Morey.

Materials for Other Courses Found at **MathTalker.org**