Section 4.2

Cofactor Expansions
Orientation

Last time: we learned . . .

▶ . . . the definition of the determinant.
▶ . . . to compute the determinant using row reduction.
▶ . . . all sorts of magical properties of the determinant, like
  ▶ \( \det(AB) = \det(A) \det(B) \)
  ▶ the determinant computes volumes
  ▶ nonzero determinants characterize invertability
  ▶ etc.

Today: we will learn . . .

▶ Special formulas for \( 2 \times 2 \) and \( 3 \times 3 \) matrices.
▶ How to compute determinants using cofactor expansions.
▶ How to compute inverses using determinants.
Determinants of $2 \times 2$ Matrices

We already have a formula in the $2 \times 2$ case:

$$\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc.$$
Determinants of $3 \times 3$ Matrices

Here’s the formula:

$$\text{det} \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} = a_{11}a_{22}a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31} - a_{11}a_{23}a_{32} - a_{12}a_{21}a_{33}$$

How on earth do you remember this? Draw a bigger matrix, repeating the first two columns to the right:

$$+ \begin{vmatrix} a_{11} & a_{12} & a_{13} & a_{11} & a_{12} \\ a_{21} & a_{22} & a_{23} & a_{21} & a_{22} \\ a_{31} & a_{32} & a_{33} & a_{31} & a_{32} \end{vmatrix} - \begin{vmatrix} a_{11} & a_{12} & a_{13} & a_{11} & a_{12} \\ a_{21} & a_{22} & a_{23} & a_{21} & a_{22} \\ a_{31} & a_{32} & a_{33} & a_{31} & a_{32} \end{vmatrix}$$

Then add the products of the downward diagonals, and subtract the product of the upward diagonals. For example,

$$\text{det} \begin{pmatrix} 5 & 1 & 0 \\ -1 & 3 & 2 \\ 4 & 0 & -1 \end{pmatrix} = \begin{vmatrix} 5 & 1 & 0 & 5 & 1 \\ -1 & 3 & 2 & -1 & 3 \\ 4 & 0 & -1 & 4 & 0 \end{vmatrix} = -15 + 8 + 0 - 0 - 0 - 1 = -8$$
Cofactor Expansions

When \( n \geq 4 \), the determinant isn’t just a sum of products of diagonals. The formula is recursive: you compute a larger determinant in terms of smaller ones.

First some notation. Let \( A \) be an \( n \times n \) matrix.

\[ A_{ij} = \text{ij}^{\text{th}} \text{ minor of } A \]

\[ = (n - 1) \times (n - 1) \text{ matrix you get by deleting the } i^{\text{th}} \text{ row and } j^{\text{th}} \text{ column} \]

\[ C_{ij} = (-1)^{i+j} \det A_{ij} \]

\[ = \text{ij}^{\text{th}} \text{ cofactor of } A \]

The signs of the cofactors follow a checkerboard pattern:

\[
\begin{pmatrix}
+ & - & + & - \\
- & + & - & + \\
+ & - & + & - \\
- & + & - & +
\end{pmatrix}
\]

\pm \text{ in the } ij \text{ entry is the sign of } C_{ij}

**Theorem**

The determinant of an \( n \times n \) matrix \( A \) is

\[
\det(A) = \sum_{j=1}^{n} a_{1j} C_{1j} = a_{11} C_{11} + a_{12} C_{12} + \cdots + a_{1n} C_{1n}.
\]

This formula is called **cofactor expansion along the first row**.
This is the beginning of the recursion.

\[ \det(a_{11}) = a_{11}. \]
Cofactor Expansions

\[ A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \]

The minors are:

\[ A_{11} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = (a_{22}) \]

\[ A_{12} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = (a_{21}) \]

\[ A_{21} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = (a_{12}) \]

\[ A_{22} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = (a_{11}) \]

The cofactors are:

\[ C_{11} = + \det A_{11} = a_{22} \]

\[ C_{12} = - \det A_{12} = -a_{21} \]

\[ C_{21} = - \det A_{21} = -a_{12} \]

\[ C_{22} = + \det A_{22} = a_{11} \]

The determinant is:

\[ \det A = a_{11} C_{11} + a_{12} C_{12} = a_{11} a_{22} - a_{12} a_{21}. \]
Cofactor Expansions

$3 \times 3$ Matrices

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}$$

The top row minors and cofactors are:

$$A_{11} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} = \begin{pmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{pmatrix} \quad C_{11} = + \det \begin{pmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{pmatrix}$$

$$A_{12} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} = \begin{pmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{pmatrix} \quad C_{12} = - \det \begin{pmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{pmatrix}$$

$$A_{13} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} = \begin{pmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{pmatrix} \quad C_{13} = + \det \begin{pmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{pmatrix}$$

The determinant is magically the same formula as before:

$$\det A = a_{11} C_{11} + a_{12} C_{12} + a_{13} C_{13}$$

$$= a_{11} \det \begin{pmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{pmatrix} - a_{12} \det \begin{pmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{pmatrix} + a_{13} \det \begin{pmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{pmatrix}$$
Cofactor Expansions

Example

\[
\begin{vmatrix}
5 & 1 & 0 \\
-1 & 3 & 2 \\
4 & 0 & -1 \\
\end{vmatrix} = 5 \cdot \begin{vmatrix}
5 & 1 & 0 \\
-1 & -3 & 2 \\
4 & 0 & -1 \\
\end{vmatrix} - 1 \cdot \begin{vmatrix}
-1 & 3 & 2 \\
4 & 0 & -1 \\
5 & 1 & 0 \\
\end{vmatrix} \\
+ 0 \cdot \begin{vmatrix}
-1 & 3 & 2 \\
4 & 0 & -1 \\
5 & 1 & 0 \\
\end{vmatrix} \\
= 5 \cdot \begin{vmatrix}
3 & 2 \\
0 & -1 \\
\end{vmatrix} - 1 \cdot \begin{vmatrix}
-1 & 2 \\
4 & -1 \\
\end{vmatrix} + 0 \cdot \begin{vmatrix}
-1 & 3 \\
4 & 0 \\
\end{vmatrix} \\
= 5 \cdot (-3 - 0) - 1 \cdot (1 - 8) \\
= -15 + 7 = -8
\]
2n – 1 More Formulas for the Determinant

Recall: the formula

$$\det(A) = \sum_{j=1}^{n} a_{1j} C_{1j} = a_{11} C_{11} + a_{12} C_{12} + \cdots + a_{1n} C_{1n}. $$

is called **cofactor expansion along the first row**. Actually, you can expand cofactors along any row or column you like!

**Magical Theorem**

The determinant of an $n \times n$ matrix $A$ is

$$\det A = \sum_{j=1}^{n} a_{ij} C_{ij} \quad \text{for any fixed } i$$

$$\det A = \sum_{i=1}^{n} a_{ij} C_{ij} \quad \text{for any fixed } j$$

These formulas are called **cofactor expansion along the } i\text{th row}, respectively, } j\text{th column}.

In particular, you get the same answer whichever row or column you choose.

Try this with a row or a column with a lot of zeros.
Cofactor Expansion

Example

\[
A = \begin{pmatrix}
2 & 1 & 0 \\
1 & 1 & 0 \\
5 & 9 & 1
\end{pmatrix}
\]

It looks easiest to expand along the third column:

\[
det A = 0 \cdot det \begin{pmatrix} \\
\text{(don’t care)} \\
\text{(don’t care)} \\
\end{pmatrix} - 0 \cdot det \begin{pmatrix} \\
\text{(don’t care)} \\
\text{(don’t care)} \\
\end{pmatrix} + 1 \cdot det \begin{pmatrix}
2 & 1 & 0 \\
1 & 1 & 0 \\
5 & 9 & 1
\end{pmatrix}
\]

\[
= det \begin{pmatrix} 2 & 1 \\
1 & 1
\end{pmatrix} = 2 - 1 = 1
\]
Cofactor Expansion

Advice

- In general, computing a determinant by cofactor expansion is slower than by row reduction.
- It makes sense to expand by cofactors if you have a row or column with a lot of zeros.
- Also if your matrix has unknowns in it, since those are hard to row reduce (you don’t know where the pivots are).

You can also use more than one method; for example:

- Use cofactors on a $4 \times 4$ matrix but compute the minors using the $3 \times 3$ formula.
- Do row operations to produce a row/column with lots of zeros, then expand cofactors (but keep track of how you changed the determinant!).

Example:

\[
\begin{vmatrix}
5 & 1 & 0 \\
-1 & 3 & 2 \\
4 & 0 & -1
\end{vmatrix}
\]

\[
\rightarrow R_2 = R_2 + 2R_3
\]

\[
\begin{vmatrix}
5 & 1 & 0 \\
7 & 3 & 0 \\
4 & 0 & -1
\end{vmatrix}
\]

\[
\rightarrow 3rd \text{ column}
\]

\[
(-1) \begin{vmatrix}
5 & 1 \\
7 & 3
\end{vmatrix} = -8
\]
Repeatedly expanding along the first row, you get:

$$\det \begin{pmatrix} 0 & 7 & 2 & 9 & 8 \\ 1 & 3 & 2 & 7 & 4 \\ 0 & 0 & 0 & 0 & 3 \\ 0 & 0 & 2 & 1 & 1 \\ 0 & 0 & 0 & 2 & 5 \end{pmatrix} = ?$$

A. $-84$  B. $-28$  C. $-7$  D. $0$  E. $7$  F. $28$  G. $84$

$$= (-1) \cdot 7 \cdot \det \begin{pmatrix} 0 & 0 & 3 \\ 2 & 1 & 1 \\ 0 & 2 & 5 \end{pmatrix} = (-1) \cdot 7 \cdot 3 \cdot \det \begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix}$$

$$= (-1) \cdot 7 \cdot 3 \cdot 2 \cdot 2 = -84.$$
A General Formula for the Inverse

Just for fun!

For $2 \times 2$ matrices we had a nice formula for the inverse:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \iff A^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix} = \frac{1}{\det A} \begin{pmatrix} C_{11} & C_{21} \\ C_{12} & C_{22} \end{pmatrix}.$$ 

**Theorem**

This last formula works for any $n \times n$ invertible matrix $A$:

$$A^{-1} = \frac{1}{\det A} \begin{pmatrix} C_{11} & C_{21} & C_{31} & \cdots & C_{n1} \\ C_{12} & C_{22} & C_{32} & \cdots & C_{n2} \\ C_{13} & C_{23} & C_{33} & \cdots & C_{n3} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ C_{1n} & C_{2n} & C_{3n} & \cdots & C_{nn} \end{pmatrix} = \frac{1}{\det A} (C_{ij})^T$$ 

Note that the cofactors are “transposed”: the $(i,j)$ entry of the matrix is $C_{ji}$. 

The proof uses Cramer’s rule.
A Formula for the Inverse

Example

Compute $A^{-1}$, where $A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix}$.

The minors are:

- $A_{11} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$
- $A_{12} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$
- $A_{13} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}$
- $A_{21} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$
- $A_{22} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$
- $A_{23} = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}$
- $A_{31} = \begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}$
- $A_{32} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$
- $A_{33} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$

The cofactors are (don’t forget to multiply by $(-1)^{i+j}$):

- $C_{11} = -1$
- $C_{12} = 1$
- $C_{13} = -1$
- $C_{21} = 1$
- $C_{22} = -1$
- $C_{23} = -1$
- $C_{31} = -1$
- $C_{32} = -1$
- $C_{33} = 1$

The determinant is (expanding along the first row):

$$\det A = 1 \cdot C_{11} + 0 \cdot C_{12} + 1 \cdot C_{13} = -2$$
Compute $A^{-1}$, where $A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix}$.

The inverse is

$$A^{-1} = \frac{1}{\det A} \begin{pmatrix} C_{11} & C_{21} & C_{31} \\ C_{12} & C_{22} & C_{32} \\ C_{13} & C_{23} & C_{33} \end{pmatrix} = -\frac{1}{2} \begin{pmatrix} -1 & 1 & -1 \\ 1 & -1 & -1 \\ -1 & -1 & 1 \end{pmatrix}.$$

Check:

$$\begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix} \cdot -\frac{1}{2} \begin{pmatrix} -1 & 1 & -1 \\ 1 & -1 & -1 \\ -1 & -1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$
A Formula for the Inverse

Why?

\[
A^{-1} = \frac{1}{\det A} \begin{pmatrix}
C_{11} & C_{21} & C_{31} & \cdots & C_{n1} \\
C_{12} & C_{22} & C_{32} & \cdots & C_{n2} \\
C_{13} & C_{23} & C_{33} & \cdots & C_{n3} \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
C_{1n} & C_{2n} & C_{3n} & \cdots & C_{nn}
\end{pmatrix}
\]

That was a lot of work! It’s way easier to compute inverses by row reduction.

- The formula is good for error estimates: the only division is by the determinant, so if your determinant is tiny, your error bars are large.

- It’s also useful if your matrix has unknowns in it.

- It’s part of a larger picture in the theory.
We have several ways to compute the determinant of a matrix.

- Special formulas for $2 \times 2$ and $3 \times 3$ matrices.
  These work great for small matrices.

- Cofactor expansion.
  This is perfect when there is a row or column with a lot of zeros, or if your matrix has unknowns in it.

- Row reduction.
  This is the way to go when you have a big matrix which doesn't have a row or column with a lot of zeros.

- Any combination of the above.

  Cofactor expansion is recursive, but you don’t have to use cofactor expansion to compute the determinants of the minors! Or you can do row operations and then a cofactor expansion.