

Last time: we defined $\det(A)$ using row operations:

L15

- (1) If $A \xrightarrow{R_i \leftarrow c R_i} B$ then $\det(B) = \det(A)$
- (2) If $A \xrightarrow{R_i \leftarrow c R_i} B$ then $\det(B) = c \cdot \det(A)$
- (3) If $A \xrightarrow{R_i \leftrightarrow R_j} B$ then $\det(B) = -\det(A)$
- (4) $\det(I_n) = 1$.

This is the fastest way to compute the determinant of a general matrix with known entries. However, we are going to care a lot about matrices with **unknown entries**. Then elimination becomes tedious because you don't know if an entry is a pivot or not.

Eg: $\det \begin{pmatrix} -\lambda & 1 & 3 \\ 1 & 2-\lambda & 1 \\ 1 & 1 & -\lambda \end{pmatrix} = ?$ Is $-\lambda$ a pivot?

Today: Other ways of computing determinants, with an aside on cross products.

Cofactor Expansion

This is a handy **recursive formula** for \det that works well for matrices with unknown entries.

Recursive: we'll compute $\det(n \times n)$ by computing several $\det((n-1) \times (n-1))$'s.

Def: Let A be an $n \times n$ matrix, and $1 \leq i, j \leq n$.

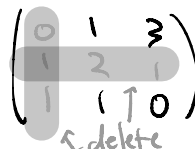
- The (i, j) **minor** A_{ij} is the $(n-1) \times (n-1)$ matrix obtained by deleting the i^{th} row & j^{th} column of A .
- The (i, j) **cofactor** C_{ij} is

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

- The **cofactor matrix** is the matrix C whose (i, j) entry is C_{ij} :

$$C = \begin{pmatrix} C_{11} & C_{12} & \dots \\ C_{21} & C_{22} & \dots \\ \vdots & \vdots & \ddots \end{pmatrix}$$

Eg: $A = \begin{pmatrix} 0 & 1 & 3 \\ 1 & 2 & 1 \\ 1 & 1 & 0 \end{pmatrix}$ $A_{21} = \begin{pmatrix} 0 & 1 & 3 \\ 1 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 3 \\ 1 & 0 \end{pmatrix}$



$$C_{21} = (-1)^{2+1} \det \begin{pmatrix} 1 & 3 \\ 1 & 0 \end{pmatrix} = -(-3) = 3$$

NB: $(-1)^{i+j}$ follows a checkerboard pattern:

$$\begin{pmatrix} + & - & + \\ - & + & - \\ + & - & + \end{pmatrix}$$

$$+ : (-1)^{i+j} = +1$$

$$- : (-1)^{i+j} = -1$$

Thm (Cofactor Expansion):

Let A be an $n \times n$ matrix, $a_{ij} = (i, j)$ -entry of A .

(1) Cofactor expansion along the i^{th} row:

$$\det(A) = \sum_{j=1}^n a_{ij} C_{ij} = a_{i1} C_{i1} + a_{i2} C_{i2} + \dots + a_{in} C_{in}$$

(2) Cofactor expansion along the j^{th} column:

$$\det(A) = \sum_{i=1}^n a_{ij} C_{ij} = a_{1j} C_{1j} + a_{2j} C_{2j} + \dots + a_{nj} C_{nj}$$

Eg: $A = \begin{pmatrix} 0 & 1 & 3 \\ 1 & 2 & 1 \\ 1 & 1 & 0 \end{pmatrix}$

(1) Expand cofactors along the 3rd row:

$$\begin{aligned} \det \begin{pmatrix} 0 & 1 & 3 \\ 1 & 2 & 1 \\ 1 & 1 & 0 \end{pmatrix} &= \underset{a_{31}}{1} \cdot \underset{C_{31}}{\det \begin{pmatrix} 1 & 3 \\ 2 & 1 \end{pmatrix}} + \underset{a_{32}}{1} \cdot (-1) \underset{C_{32}}{\det \begin{pmatrix} 0 & 3 \\ 1 & 1 \end{pmatrix}} + \underset{a_{33}}{0} \cdot \underset{C_{33}}{\det \begin{pmatrix} 0 & 1 \\ 1 & 2 \end{pmatrix}} \\ &= 1 \cdot (1 \cdot 1 - 2 \cdot 3) - 1 \cdot (0 \cdot 1 - 3 \cdot 1) \\ &= -5 + 3 = -2 \end{aligned}$$

(2) Expand cofactors along the 2nd column:

$$\begin{aligned} \det \begin{pmatrix} 0 & 1 & 3 \\ 1 & 2 & 1 \\ 1 & 1 & 0 \end{pmatrix} &= \underset{a_{12}}{1} \cdot (-1) \underset{C_{12}}{\det \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}} + \underset{a_{22}}{2} \cdot \underset{C_{22}}{\det \begin{pmatrix} 0 & 3 \\ 1 & 0 \end{pmatrix}} + \underset{a_{32}}{1} \cdot (-1) \underset{C_{32}}{\det \begin{pmatrix} 0 & 3 \\ 1 & 1 \end{pmatrix}} \\ &= -(-1) + 2(-3) - (-3) = -2 \end{aligned}$$

Remarks:

- (1) This is a **recursive** formula: $C_{ij} = \pm \det((n-1) \times (n-1))$
- (2) You can compute $C_{ij} = (-1)^{i+j} \det(A_{ij})$ however you like — you don't have to use cofactor expansion every time.
- (3) Expanding along any row or column gives you $\det(A)$: you always get the **same number**.
- (4) This is **ridiculously slow**: $O(n! \cdot n)$. It's only useful when your matrix has
 - **unknown entries**, or
 - a row/column with **lots of zeros**

Eg: $\det \begin{pmatrix} -\lambda & 1 & 3 \\ 1 & 2-\lambda & 1 \\ 1 & 1 & -\lambda \end{pmatrix}$

$$\begin{aligned} & \xrightarrow[\text{1st col}]{\text{expand}} (-\lambda) \cdot \det \begin{pmatrix} 2-\lambda & 1 \\ 1 & -\lambda \end{pmatrix} + 1 \cdot (-1) \det \begin{pmatrix} 1 & 3 \\ 1 & -\lambda \end{pmatrix} + 1 \cdot \det \begin{pmatrix} 1 & 3 \\ 2-\lambda & 1 \end{pmatrix} \\ &= -\lambda((2-\lambda)(-\lambda)-1) - (-\lambda-3) + (1-3(2-\lambda)) \\ &= -\lambda(\lambda^2-2\lambda-1) + \lambda+3 + 1-6+3\lambda \\ &= -\lambda^3+2\lambda^2+5\lambda-2 \end{aligned}$$

In fact, for 3×3 matrices it's not too hard to compute the determinant even if all of the entries are unknown (like for 2×2).


Eg: $\det \begin{pmatrix} a & b & c \\ d & e & f \\ g & h & i \end{pmatrix} = a \cdot \det \begin{pmatrix} e & f \\ h & i \end{pmatrix} + b(-1) \det \begin{pmatrix} d & f \\ g & i \end{pmatrix} + c \det \begin{pmatrix} d & e \\ g & h \end{pmatrix}$

$$= a(ei - fh) - b(di - fg) + c(dh - eg)$$

$$= aei + bfg + cdh - afh - bdi - ceg$$

Of course, this is only useful if you can remember it.

Sarrus' Scheme:



$$\det = aei + bfg + cdh - afh - bdi - ceg$$

This says: to compute $\det(3 \times 3)$ matrix:


- (1) repeat the first 2 columns on the right
- (2) sum the products of the forward diagonals and subtract the products of the backward diagonals

Warning: This only works for 3×3 matrices!

If you try it on a bigger matrix, you won't get the determinant!

→ See the **big formula** at the end for an $n \times n$ version.

Eg: $\det \begin{pmatrix} 0 & 1 & 3 \\ 1 & 2 & 1 \\ 1 & 1 & 0 \end{pmatrix} = 0 \cdot 2 \cdot 0 + 1 \cdot 1 \cdot 1 + 3 \cdot 1 \cdot 1 - 3 \cdot 2 \cdot 1 - 0 \cdot 1 \cdot 1 - 1 \cdot 1 \cdot 0$



$$= 4 - 6 = -2 \quad (\text{again})$$

Computing 3×3 determinants using Sarrus' scheme vs. cofactor expansion is largely a matter of personal preference.

Here's a matrix with a column with lots of zeros:

Eg: $\det \begin{pmatrix} 2 & 5 & -3 & -1 \\ -2 & -3 & 2 & -5 \\ 1 & 3 & -2 & 0 \\ -1 & 6 & 4 & 0 \end{pmatrix}$

$$= (-1)(-1) \det \begin{pmatrix} -2 & -3 & 2 \\ 1 & 3 & -2 \\ -1 & 6 & 4 \end{pmatrix} - 5 \cdot \det \begin{pmatrix} 2 & 5 & -3 \\ 1 & 3 & -2 \\ -1 & 6 & 4 \end{pmatrix}$$

$$+ 0 \cdot (-1) \det \begin{pmatrix} \text{don't} \\ \text{care} \end{pmatrix} + 0 \cdot \det \begin{pmatrix} \text{don't} \\ \text{care} \end{pmatrix}$$

$$= -24 - 5(11) = -79$$

We only had to compute two 3×3 determinants!

Better idea: do a row operation first!

$$\det \begin{pmatrix} 2 & 5 & -3 & -1 \\ -2 & -3 & 2 & -5 \\ 1 & 3 & -2 & 0 \\ -1 & 6 & 4 & 0 \end{pmatrix} \xrightarrow{R_2 \leftarrow -R_1} \det \begin{pmatrix} 2 & 5 & -3 & -1 \\ -12 & -28 & 17 & 0 \\ 1 & 3 & -2 & 0 \\ -1 & 6 & 4 & 0 \end{pmatrix}$$

$$= (-1)(-1) \det \begin{pmatrix} -12 & -28 & 17 \\ 1 & 3 & -2 \\ -1 & 6 & 4 \end{pmatrix} = -79$$

Now we only had to compute one 3×3 determinant!

Summary: Methods for Computing Determinants

(1) Special Formulas (2×2 , 3×3)

best for small matrices, although cofactors is still faster for a 3×3 matrix with lots of 0's

(2) Cofactor Expansion

best for matrices with unknown entries, or a row/column with lots of zeros

(3) Row (& column) operations

best if you have a big matrix with known entries & no row/column with lots of zeros

(4) Any Combination of the Above

eg. do a row operation to make a row with lots of zeros, then expand cofactors, using Sarrus' scheme to compute C_{ij} 's ...

Aside: Cofactor Formula for A^{-1}

You can actually compute A^{-1} using cofactors:

Thm: Let C be the cofactor matrix of A . Then

$$AC^T = C^T A = \det(A) \cdot I_n$$

In particular, if $\det(A) \neq 0$ then

$$A^{-1} = \frac{1}{\det(A)} C^T \quad (\text{see the supplement})$$

NB: This is ridiculously inefficient computationally.
Don't compute A^{-1} this way!

It does tell you that A^{-1} only has $\det(A)$ in the denominators — more on the HW.

One reason I like the cofactor formula is it helps me remember the formula for the 2×2 inverse.

Eg: $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \Rightarrow C = \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$

$$\Rightarrow A^{-1} = \frac{1}{\det(A)} C^T = \frac{1}{ad-bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

Cross Products

This is an operation that takes two vectors in \mathbb{R}^3 and gives you a vector that's orthogonal to both of them.

Recall: The unit coordinate vectors in \mathbb{R}^3 are

$$e_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \quad e_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad e_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

Def: Let $v = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$, $w = \begin{pmatrix} x \\ y \\ z \end{pmatrix} \in \mathbb{R}^3$. Their cross product is

$$v \times w = \begin{pmatrix} bz - cy \\ cx - az \\ ay - bx \end{pmatrix} \in \mathbb{R}^3$$

NB: (vector) \times (vector) = (vector)

but (vector) \cdot (vector) = (scalar)

More than a Mnemonic:

$$\begin{aligned} \begin{pmatrix} a \\ b \\ c \end{pmatrix} \times \begin{pmatrix} x \\ y \\ z \end{pmatrix} &= \text{"det"} \begin{pmatrix} e_1 & e_2 & e_3 \\ -v_1 & - & - \\ -w_1 & - & - \end{pmatrix} \quad \left(\begin{array}{l} \text{interpreted as} \\ \text{cofactor expansion} \\ \text{along the 1st row} \end{array} \right) \\ &= e_1 \det \begin{pmatrix} b & c \\ y & z \end{pmatrix} - e_2 \det \begin{pmatrix} a & c \\ x & z \end{pmatrix} + e_3 \det \begin{pmatrix} a & b \\ x & y \end{pmatrix} \\ &= (bz - cy) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} - (az - cx) \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + (ay - bx) \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} bz - cy \\ cx - az \\ ay - bx \end{pmatrix} \end{aligned}$$

$$\begin{aligned}
 \text{Eg: } \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} \times \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} &= \text{"det } \begin{pmatrix} e_1 & e_2 & e_3 \\ 1 & 2 & 1 \\ 1 & 1 & 0 \end{pmatrix} \text{"} \\
 &= e_1 \det \begin{pmatrix} 2 & 1 \\ 1 & 0 \end{pmatrix} - e_2 \det \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix} + e_3 \det \begin{pmatrix} 1 & 2 \\ 1 & 1 \end{pmatrix} \\
 &= - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} - \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \\ -1 \end{pmatrix}
 \end{aligned}$$

$$\text{NB: } \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} -1 \\ 1 \\ -1 \end{pmatrix} = 0 \quad \text{and} \quad \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \cdot \begin{pmatrix} -1 \\ 1 \\ -1 \end{pmatrix} = 0 \dots$$

Here's an important relationship between cross products and determinants:

Def: Let $u, v, w \in \mathbb{R}^3$. Their **triple product** is

$$u \cdot (v \times w) = \det \begin{pmatrix} -u^T- \\ -v^T- \\ -w^T- \end{pmatrix}$$

Check this formula: if $u = \begin{pmatrix} i \\ j \\ k \end{pmatrix}$, $v = \begin{pmatrix} a \\ b \\ c \end{pmatrix}$, $w = \begin{pmatrix} x \\ y \\ z \end{pmatrix}$ then

$$\begin{aligned}
 u \cdot (v \times w) &= \begin{pmatrix} i \\ j \\ k \end{pmatrix} \cdot \left[e_1 \det \begin{pmatrix} b & c \\ y & z \end{pmatrix} - e_2 \det \begin{pmatrix} a & c \\ x & z \end{pmatrix} + e_3 \det \begin{pmatrix} a & b \\ x & y \end{pmatrix} \right] \\
 &= \begin{pmatrix} i \\ j \\ k \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \det \begin{pmatrix} b & c \\ y & z \end{pmatrix} - \begin{pmatrix} i \\ j \\ k \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \det \begin{pmatrix} a & c \\ x & z \end{pmatrix} + \begin{pmatrix} i \\ j \\ k \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \det \begin{pmatrix} a & b \\ x & y \end{pmatrix} \\
 &= i \det \begin{pmatrix} b & c \\ y & z \end{pmatrix} - j \det \begin{pmatrix} a & c \\ x & z \end{pmatrix} + k \det \begin{pmatrix} a & b \\ x & y \end{pmatrix} \\
 &= \det \begin{pmatrix} i & j & k \\ a & b & c \\ x & y & z \end{pmatrix} \quad \checkmark
 \end{aligned}$$

Eg: $u = \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix}$, $v = \begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}$, $w = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$

$\hookrightarrow v \times w = \begin{pmatrix} -1 \\ 1 \\ -1 \end{pmatrix}$ (from before)

$\hookrightarrow u \cdot (v \times w) = \begin{pmatrix} 0 \\ 1 \\ 3 \end{pmatrix} \cdot \begin{pmatrix} -1 \\ 1 \\ -1 \end{pmatrix} = 1 - 3 = -2$

$\Rightarrow \det \begin{pmatrix} 0 & 1 & 3 \\ 1 & 2 & 1 \\ 1 & 0 & 0 \end{pmatrix} = -2$ (again)

Properties of Cross Products:

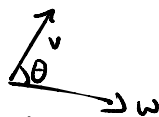
(1) $v \times w \perp v$ and $v \times w \perp w$

\rightarrow because $v \cdot (v \times w) = \det \begin{pmatrix} - & v^T & - \\ - & v^T & - \\ - & w^T & - \end{pmatrix} = 0$

(2) $v \times w = -w \times v$

\rightarrow because " $\det \begin{pmatrix} e_1 & e_2 & e_3 \\ - & v^T & - \\ - & w^T & - \end{pmatrix}$ " $\xrightarrow{\text{row swap}}$ " $\det \begin{pmatrix} e_1 & e_2 & e_3 \\ - & w^T & - \\ - & v^T & - \end{pmatrix}$ "

(3) $\|v \times w\| = \|v\| \cdot \|w\| \cdot \sin(\theta)$

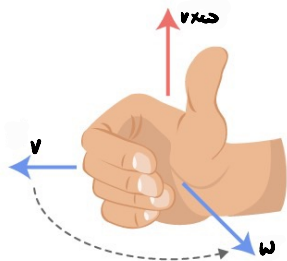


\rightarrow compare $v \cdot w = \|v\| \cdot \|w\| \cdot \cos(\theta)$

(4) $v \times w = 0 \iff v$ and w are **collinear**

$\rightarrow \sin(\theta) = 0 \iff \theta = 0^\circ$ or $\theta = 180^\circ$

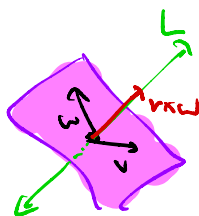
(5) $v \times w$ points in the direction determined by the **right hand rule**.



NB: (1)+(3)+(5) give a geometric characterization of the cross product:

→ If v, w are collinear then $v \times w = 0$

→ otherwise $\text{Span}\{v, w\}$ is a plane, so $\text{Span}\{v, w\}^\perp$ is a line L .



(1) says $v \times w$ lies on L

(3) says how long $v \times w$ is

(5) says which direction $v \times w$ points.

Why are Cross Products Useful?

They are ubiquitous in multivariable calculus & physics.

For us, it's a shortcut for computing orthogonal complements in \mathbb{R}^3 .

$$\text{Span}\{v, w\}^\perp = \text{Span}\{v \times w\}$$

if $v, w \in \mathbb{R}^3$ are noncollinear

Eg: Find a basis for $\text{Span}\left\{\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}\right\}^\perp$

We eyeball $\begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix} \in \text{Span}\left\{\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}\right\}^\perp$, but we need one more:

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \times \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} -3 \\ -6 \\ 5 \end{pmatrix} \rightarrow \text{basis } \left\{ \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -3 \\ -6 \\ 5 \end{pmatrix} \right\}$$

The Big Formula (JUST FOR FUNSIES)

This is an explicit, non-recursive formula for $\det(A)$.
It's not important for this class, but it's nice to know it exists.

Def: A **permutation** of $\{1, 2, \dots, n\}$ is a re-ordering
 $\sigma: \{1, 2, \dots, n\} \rightarrow \{1, 2, \dots, n\}$
 $\sigma(i)$ = new number in the i^{th} position.

So a permutation of $\{1, 2, \dots, 52\}$ is just a way of shuffling a deck of cards.

Eg: Here are all of the permutations of $\{1, 2, 3\}$:

123 132 213 231 312 321

↪ i.e. $\sigma(1)=1$ $\sigma(2)=3$ $\sigma(3)=2$

Q: How many permutations of $\{1, 2, \dots, n\}$ are there?

- n choices for $\sigma(1)$
- $n-1$ choices for $\sigma(2)$
- $n-2$ choices for $\sigma(3)$
- \vdots
- 2 choices for $\sigma(n-1)$
- 1 choice for $\sigma(n)$

$$\left. \begin{array}{l} n(n-1)(n-2)\dots 2 \cdot 1 \\ = n! \end{array} \right\} \text{ total}$$

NB: $n!$ is a big number! eg. $52! \approx 8 \times 10^{67}$

→ there are only about 10^{24} stars in the universe!

→ so every time you shuffle a deck of cards, you make history - nobody's ever done that before!

→ The Big Formula for a 52×52 matrix has $52!$ summands! It's not computable!

Def: A **transposition** is a permutation that just swaps two numbers.

Eg: $1\overset{\curvearrowright}{3}2$ and $2\overset{\curvearrowright}{1}3$ are transpositions
but 231 is not

Fact: Any permutation can be obtained by doing some number of transpositions.

Eg: $1\overset{\curvearrowright}{2}3 \rightsquigarrow 2\overset{\curvearrowright}{1}3 \rightsquigarrow 231$ $1\overset{\curvearrowright}{2}3 \rightsquigarrow 1\overset{\curvearrowright}{3}2 \rightsquigarrow 312$

Def: The **sign** of a permutation σ is **$\text{sign}(\sigma) =$**

- $+1$ if σ can be obtained by doing an **even** number of transpositions
- -1 if σ can be obtained by doing an **odd** number of transpositions

Eg: $231 = (2 \text{ transpositions}) \Rightarrow \text{sign}(231) = +1$
 $132 = (1 \text{ transposition}) \Rightarrow \text{sign}(132) = -1$

Amazingly, this is well-defined — there's no permutation that can be obtained as both an even and an odd number of transpositions.

The Big Formula:

Let A be an $n \times n$ matrix with (i,j) entry a_{ij} .

$$\det(A) = \sum_{\substack{\text{all} \\ \text{permutations} \\ \sigma}} \text{sign}(\sigma) a_{1\sigma(1)} a_{2\sigma(2)} \cdots a_{n\sigma(n)}$$

In other words, for every way of choosing exactly one entry in each row & column, multiply those entries together, and sum with signs ± 1 .

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

PERMUTATION	#TRANSPOSITIONS	SIGN	SUMMAND
123	0	+1	$a_{11}a_{22}a_{33}$
132	1	-1	$-a_{11}a_{23}a_{32}$
213	1	-1	$-a_{12}a_{21}a_{33}$
231	2	+1	$+a_{12}a_{23}a_{31}$
312	2	+1	$+a_{13}a_{21}a_{32}$
321	1	-1	$-a_{13}a_{22}a_{31}$
$= \det(A)$			