Lecturer: Barwick

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#### So here's an $n \times n$ matrix

$$A = \left( \begin{array}{ccc} \vec{v}_1 & \cdots & \vec{v}_n \end{array} \right),$$

and we have this number

$$\det(A) = \det\left(\vec{v}_1, \dots, \vec{v}_n\right) \in \mathbf{R}$$

that measures the *signed n-dimensional volume* of the parallelopiped spanned by  $\vec{v}_1, \ldots, \vec{v}_n$ .

Let's list the things we know about det(A).



## (1) *Normalization*. The identity matrix has determinant 1:

$$\det(I) = \det\left(\hat{e}_1, \cdots, \hat{e}_n\right) = 1.$$

## (2) *Multilinearity*. For any real numbers $r, s \in \mathbf{R}$ ,

$$\det \left( \vec{v}_1, \cdots, r\vec{x}_i + s\vec{y}_i, \cdots, \vec{v}_n \right) = r \det \left( \vec{v}_1, \cdots, \vec{x}_i, \cdots, \vec{v}_n \right) \\ + s \det \left( \vec{v}_1, \cdots, \vec{y}_i, \cdots, \vec{v}_n \right).$$



### (3) Alternation. The determinant

$$\det\left(\vec{v}_1,\cdots,\cdots,\vec{v}_n\right)=0$$

if any two of the  $\vec{v}_i$ s are equal.

These are the core, defining properties of det.



Here are more properties, which we *deduced* from the three core properties above:

- (4) Multiplying a row or column by a number  $r \in \mathbf{R}$  multiplies the determinant by that r.
- (5) Swapping two rows or columns in A multiplies the determinant by a -1.
- (6) Adding a multiple of a row or column onto another row or column doesn't change the determinant.



And here are some general computational facts we extract from these properties.

- (8)  $det(A) \neq 0$  *if and only if A* is invertible.
- (9) det(MN) = det(M) det(N).
- (10)  $det(A^{T}) = det(A)$ . (Why?)
- (11) det diag $(\lambda_1, \ldots, \lambda_n) = \prod_{i=1}^n \lambda_i$ .
- (12) More generally, if *A* is triangular, then det(*A*) is the product of the entries along the diagonal. (Why?)



## Let's compute the determinant of the following matrices:

$$\begin{pmatrix} 1 & 2 & 4 \\ 1 & 3 & 9 \\ 1 & 4 & 16 \end{pmatrix}, \quad \begin{pmatrix} 1 & 1 & 1 & 2 \\ 1 & 1 & 2 & 2 \\ 1 & 2 & 2 & 3 \\ 2 & 2 & 3 & 4 \end{pmatrix}$$
$$\begin{pmatrix} 1 & 1 & 2 & 2 \\ 1 & 2 & 2 & 3 \\ 2 & 2 & 3 & 4 \end{pmatrix}$$
$$\begin{pmatrix} 1 & 1 & 2 & 5 & 14 \\ 2 & 5 & 14 & 2 \\ 5 & 14 & 42 & 132 \end{pmatrix}, \quad \begin{pmatrix} 1 & 2 & 5 & 14 \\ 2 & 5 & 5 & 14 \\ 5 & 14 & 42 & 132 \\ 14 & 42 & 132 & 429 \end{pmatrix}$$



I was half-joking about the last two; this is actually a neat little piece of mathematics: there's only one sequence of integers  $c_0, c_1, c_2, ...$  such that for any  $n \ge 1$ ,

$$\det(c_{i+j-2}) = \det(c_{i+j-1}) = 1.$$

These are called the *Catalan numbers*, and your mathematical life isn't complete until you've read about them! (I was going to put a problem about these on the homework, but I thought that might be too much.)

Suppose  $\sigma$ :  $\{1, ..., n\} \longrightarrow \{1, ..., n\}$  a *permutation* of  $\{1, ..., n\}$  – i.e., a bijection from that set to itself.

One can express a permutation very compactly, by writing down the matrix

$$P_{\sigma} = \left( \begin{array}{ccc} \hat{e}_{\sigma(1)} & \cdots & \hat{e}_{\sigma(n)} \end{array} \right)$$

called the *permutation matrix* corresponding to  $\sigma$ .

This is great, because matrix multiplication corresponds to composition of permutations:

$$P_{\sigma \circ \tau} = P_{\sigma} P_{\tau}$$
 and  $P_{id} = I$ .



## Also, these matrices are orthogonal; in fact,

$$P_{\sigma}^{\mathsf{T}} = P_{\sigma}^{-1} = P_{\sigma^{-1}}.$$

## Here's a permutation matrix for n = 5:

$$\left(\begin{array}{c} \hat{e}_2\\ \hat{e}_4\\ \hat{e}_1\\ \hat{e}_3\\ \hat{e}_5\end{array}\right) = \left(\begin{array}{cccccc} 0 & 1 & 0 & 0 & 0\\ 0 & 0 & 0 & 1 & 0\\ 1 & 0 & 0 & 0 & 0\\ 0 & 0 & 1 & 0 & 0\\ 0 & 0 & 0 & 0 & 1\end{array}\right).$$

What's its determinant?

This is a general pattern: the determinant of a permutation matrix  $P_{\sigma}$  is called the *sign* of the permutation  $\sigma$ :

 $\operatorname{sgn}(\sigma) \coloneqq \operatorname{det}(P_{\sigma})$ 

In effect, it's

 $(-1)^{\text{number of swaps in }\sigma}$ .

What's weird about this is that you can imagine performing more or fewer swaps to get sigma. The magic of determinants is telling you that the *parity* of the number of swaps stays the same!

Note that  $sgn(\sigma) = sgn(\sigma^{-1})$ . (Why?)



It turns out that permutations give you a formula for the determinant of any matrix  $A = \begin{pmatrix} \vec{v}_1 & \cdots & \vec{v}_n \end{pmatrix}$ . Let's see why.

First, the *j*-th column  $\vec{v}_j$  can be written as

$$\vec{v}_j = \sum_{k=1}^n a_{k,j} \hat{e}_k.$$

The multilinearity of det can then be deployed:

$$\det(A) = \det\left(\sum_{k(1)=1}^{n} a_{k(1),j}\hat{e}_{k(1)}, \dots, \sum_{k(n)=1}^{n} a_{k(n),j}\hat{e}_{k(n)}\right)$$
$$= \sum_{k(1)=1}^{n} \cdots \sum_{k(n)=1}^{n} \left(\prod_{i=1}^{n} a_{k(i),i}\right) \det(\hat{e}_{k(1)}, \dots, \hat{e}_{k(n)})$$

Now all those sums can be combined into one sum. You're summing over the set  $E_n$  of all maps  $k: \{1, ..., n\} \longrightarrow \{1, ..., n\}$ :

$$\det(A) = \sum_{k \in E_n} \left( \prod_{i=1}^n a_{k(i),i} \right) \det(\hat{e}_{k(1)}, \dots, \hat{e}_{k(n)}).$$



$$\det(A) = \sum_{k \in E_n} \left( \prod_{i=1}^n a_{k(i),i} \right) \det(\hat{e}_{k(1)}, \dots, \hat{e}_{k(n)}).$$

Now we use the alternatingness: if any two columns are equal, then the determinant is zero. So any summand in which  $k: \{1, ..., n\} \longrightarrow \{1, ..., n\}$  is not injective doesn't appear:

$$\det(A) = \sum_{\sigma \in \Sigma_n} \left( \prod_{i=1}^n a_{\sigma(i),i} \right) \det(\hat{e}_{\sigma(1)}, \dots, \hat{e}_{\sigma(n)}),$$

where  $\Sigma_n$  is the set of permutations of  $\{1, \ldots, n\}$ .

## Now we have:

$$\det(A) = \sum_{\sigma \in \Sigma_n} \left( \prod_{i=1}^n a_{\sigma(i),i} \right) \det(\hat{e}_{\sigma(1)}, \dots, \hat{e}_{\sigma(n)})$$
$$= \sum_{\sigma \in \Sigma_n} \left( \prod_{i=1}^n a_{\sigma(i),i} \right) \det(P_{\sigma})$$
$$= \sum_{\sigma \in \Sigma_n} \operatorname{sgn}(\sigma) \left( \prod_{i=1}^n a_{\sigma(i),i} \right).$$

$$\det(A) = \sum_{\sigma \in \Sigma_n} \operatorname{sgn}(\sigma) \left( \prod_{i=1}^n a_{\sigma(i),i} \right)$$

There it is - the Leibniz formula for the determinant.

Do you care? Well, if you're trying to program a computer to compute determinants, no. Evaluating this formula involves  $\Omega(n! n)$  operations. Gaussian elimination uses  $O(n^3)$  operations. We have a winner.

On the other hand, the fact that there *is* a formula is vaguely reassuring. But there's another advantage ...



Think of the determinant as a function from  $\mathbb{R}^{n^2} \longrightarrow \mathbb{R}$ ; this formula expresses that function as a *polynomial* in  $n^2$  variables. That means that it's continuous and infinitely differentiable. So this leads us to the following result:

**Proposition.** Suppose  $A = (a_{i,j})$  an invertible  $n \times n$  matrix. Then there exists an  $\varepsilon > 0$  such that if  $A' = (a'_{i,j})$  is an  $n \times n$  matrix such that  $|a'_{i,j} - a_{i,j}| < \varepsilon$ , then A' is invertible too.

That is, invertible matrices are stable under small perturbations.



Here's another wacky-sounding consequence. Suppose  $L \subseteq \mathbb{R}^{n^2}$  is a line. Then if there exists one point on *L* that corresponds to an invertible matrix, then all but finitely many points on *L* correspond to invertible matrices.



## **Question.** Suppose *A* an $n \times n$ matrix. For how many real numbers $t \in \mathbf{R}$ is A + tI is invertible (none, finitely many, infinitely many, all)?