

Determinant and Trace

Area and mappings from the plane to itself: Recall that in the last set of notes we found a linear mapping to take the unit square $S = \{0 \leq x \leq 1, 0 \leq y \leq 1\}$ to any parallelogram P with one corner at the origin. We can write the parallelogram P as

$$P = \{xv + yw : 0 \leq x \leq 1, 0 \leq y \leq 1\},$$

where v and w are the two vectors which form the edges of P starting at the origin $(0,0)$. Then we can write the linear transformation as

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = xv + yw, \quad [T] = \begin{bmatrix} v_1 & w_1 \\ v_2 & w_2 \end{bmatrix},$$

where $v = (v_1, v_2)$ and $w = (w_1, w_2)$ in components. Notice that the mapping T is invertible precisely when it does not collapse S down to a line segment (or a point), which happens precisely when the area of the parallelogram P is non-zero. Also, recall that we said T is invertible precisely when its determinant $\det([T]) = v_1w_2 - v_2w_1$ is nonzero. We have

$$\det([T]) \neq 0 \Leftrightarrow T \text{ invertible} \Leftrightarrow \text{Area}(P) \neq 0.$$

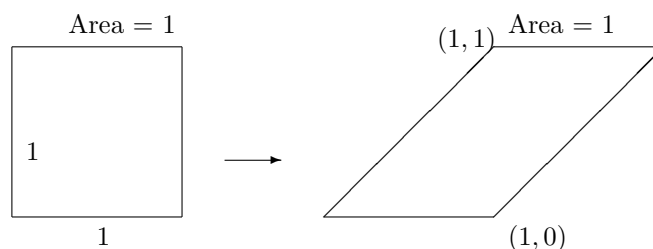
We'll see next that $\det([T])$ is the area of P , up to a sign. This is easiest to see with the shear map we examined in the last set of notes. Start with the shear map T whose matrix representation is

$$[T] = \begin{bmatrix} 1 & b \\ 0 & 1 \end{bmatrix}.$$

(In the earlier set of notes we wrote the entry in the upper right corner of $[T]$ as a , but it will turn out to be convenient to call it b for our later discussion.) In this case, T maps the unit square $S = \{0 \leq x \leq 1, 0 \leq y \leq 1\}$ to the parallelogram P spanned by the two vectors $v = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $w = \begin{bmatrix} b \\ 1 \end{bmatrix}$; in other words,

$$T(S) = P = \{xv + yw : 0 \leq x \leq 1, 0 \leq y \leq 1\} = \left\{x \begin{bmatrix} 1 \\ 0 \end{bmatrix} + y \begin{bmatrix} b \\ 1 \end{bmatrix} : 0 \leq x \leq 1, 0 \leq y \leq 1\right\}.$$

We reproduce a picture here:



We already know that the unit square S has area 1, but let's see that P also has area 1. The area of a parallelogram is equal to its base times its height, and the height and base of P are both 1, so the area of P is $1 \cdot 1 = 1$. On the other hand,

$$\det([T]) = \det\left(\begin{bmatrix} 1 & b \\ 0 & 1 \end{bmatrix}\right) = 1 \cdot 1 - 0 \cdot b = 1 = \text{Area}(P).$$

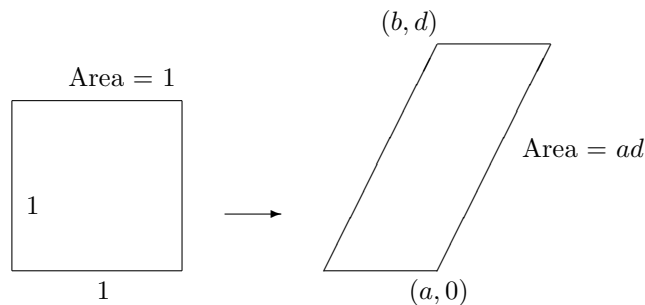
Now we can rescale the shear T by a in the horizontal direction and by d in the vertical direction, to have something more general. This time we have

$$[T] = \begin{bmatrix} a & b \\ 0 & d \end{bmatrix},$$

and

$$T(S) = P = \{xv + yw : 0 \leq x \leq 1, 0 \leq y \leq 1\} = \left\{ x \begin{bmatrix} a \\ 0 \end{bmatrix} + y \begin{bmatrix} b \\ d \end{bmatrix} : 0 \leq x \leq 1, 0 \leq y \leq 1 \right\},$$

and the picture looks like



(In this particular picture $a = 1/2$ and $d = 2$, but this choice of scaling factors is not important.) This time the height of the parallelogram P is d while its base is a , so $\text{Area}(P) = \text{base} \cdot \text{height} = ad$. Again, we have

$$|\det([T])| = \left| \det \left(\begin{bmatrix} a & b \\ 0 & d \end{bmatrix} \right) \right| = |a \cdot d - b \cdot 0| = |ad| = \text{Area}(P).$$

Notice that the absolute value here is necessary, because a and d could have opposite signs.

Now that we know $|\det([T])|$ gives the area of the image of the unit square if $[T] = \begin{bmatrix} a & b \\ 0 & d \end{bmatrix}$, it's not too hard to see this is true for any linear map. We'll first need a technical fact.

Exercise: Prove $\det(AB) = \det(A)\det(B)$ for 2×2 matrices A and B . (Really, just multiply it out.) Notice that this means $\det(AB) = \det(BA)$ for any pair of 2×2 matrices.

Exercise: Show that for any angle θ we have

$$\det([R_\theta]) = \det \left(\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \right) = 1.$$

Now let $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be a linear mapping of the plane to itself, and suppose

$$[T] = \begin{bmatrix} a & b \\ c & d \end{bmatrix}.$$

This means $T(e_1) = \begin{bmatrix} a \\ c \end{bmatrix}$ and $T(e_2) = \begin{bmatrix} b \\ d \end{bmatrix}$, where $e_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $e_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ as before.

Now, the vector $T(e_1) = \begin{bmatrix} a \\ c \end{bmatrix}$ makes some angle θ with the positive x axis, so we apply the rotation $R_{-\theta}$ to T to get a new mapping

$$\tilde{T} = R_{-\theta} \circ T, \quad [\tilde{T}] = [R_{-\theta}][T] = \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} \tilde{a} & \tilde{b} \\ 0 & \tilde{d} \end{bmatrix},$$

and

$$\det([\tilde{T}]) = \det([R_{-\theta}][T]) = \det([R_{-\theta}]) \det([T]) = \det([T]).$$

By the computation we did above, $\text{Area}(\tilde{P}) = |\det([\tilde{T}])|$. We also have that T sends the unit square S to a parallelogram P , and $\tilde{T}$ sends S to a parallelogram $\tilde{P}$. These two parallelograms P and $\tilde{P}$ differ by a rotation, so they have the same area. Thus we see

$$\text{Area}(P) = \text{Area}(\tilde{P}) = |\det([\tilde{T}])| = |\det([T])|.$$

In particular, we have just proven that $\det([T]) \neq 0$ precisely when T is invertible, because this is precisely when the image parallelogram P has nonzero area.

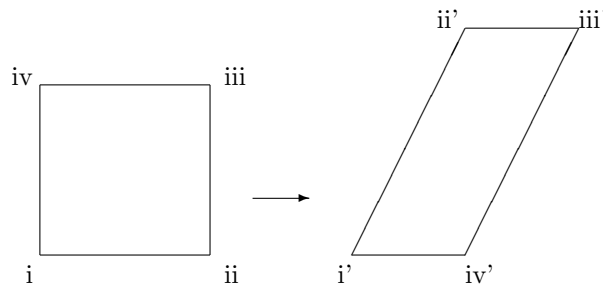
Orientation and the sign of the determinant: As we saw in the previous notes, there are actually two linear transformations which map the unit square S onto this parallelogram P , we can also have

$$T\left(\begin{bmatrix} 1 \\ 0 \end{bmatrix}\right) = \begin{bmatrix} b \\ d \end{bmatrix}, \quad T\left(\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right) = \begin{bmatrix} a \\ 0 \end{bmatrix}, \quad [T] = \begin{bmatrix} b & a \\ d & 0 \end{bmatrix}.$$

In this case we see that

$$\det([T]) = -ad = -\text{Area}(P).$$

Why do we have the minus sign? To understand what's going on, it will help to label the corners of the unit square S and the parallelogram P as in the picture below.



What does this labeling mean? The mapping T sends the vector e_1 , which goes from i to ii in the square on the left to the vector $\begin{bmatrix} b \\ d \end{bmatrix}$, which also goes from i' to ii' in the parallelogram on the right. Similarly, the mapping T sends the vector e_2 , which goes from i to iv in the square on the left to the vector $\begin{bmatrix} a \\ 0 \end{bmatrix}$, which also goes from i' to iv' in the parallelogram on the right. Now, if we follow the labeling of the corners of the square in order, as in i to ii to iii to iv , then we traverse along the boundary of the square counter-clockwise. However, if we follow the labeling of the corners of the parallelogram in order, as in i' to ii' to iii' to iv' , we traverse along the boundary of the parallelogram clockwise. This means the mapping T reversed the direction we traversed along the boundary of the shape. In other words, T reversed the orientation. We have discovered the following general principle:

$$\det([T]) < 0 \Leftrightarrow T \text{ reverses orientation.}$$

This principle is exactly why we wrote $|\det([T])| = \text{Area}(P)$ before. In general, if $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ preserves orientation then $\det([T]) = \text{Area}(P)$, but if T reverses orientation then $\det([T]) = -\text{Area}(P)$.

Higher dimensions: So far we've seen that the determinant of a 2×2 matrix is the area (up to a sign) of the parallelogram which is the image of the unit square. In fact, a similar thing

is true in higher dimensions. Let $[T]$ be an $n \times n$ matrix, which we've seen corresponds to a linear map $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$. Then T sends the unit cube $S = \{0 \leq x_i \leq 1 : i = 1, 2, \dots, n\}$ to a parallelepiped P , which is spanned by the columns of $[T]$. Then $|\det([T])| = \text{Vol}(P)$, where Vol gives the n -dimensional volume.

We begin with a quick illustrative example. Consider

$$[T] = \begin{bmatrix} a & b & 0 \\ c & d & 0 \\ 0 & 0 & e \end{bmatrix}, \quad e > 0.$$

Then the image of the unit cube S under T is

$$P = \{(x, y, z) : (x, y) \in \bar{P}, 0 \leq z \leq e\},$$

where

$$[\bar{T}] = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \quad \bar{S} = \{0 \leq x \leq 1, 0 \leq y \leq 1\}, \quad \bar{P} = T(\bar{S}).$$

By slicing P with horizontal slices, we see

$$\text{Vol}(P) = e \text{Area}(\bar{P}) = e |\det([\bar{T}])|.$$

So, by any reasonable definition of the determinant for 3×3 matrices which fits with our definition for 2×2 matrices, we must have

$$\det([T]) = e \det([\bar{T}]) = e(ad - bc).$$

Exercise: Let $[T]$ be a 3×3 matrix. Show that you can always perform a rotation to make the last column of $[T]$ into $\begin{bmatrix} 0 \\ 0 \\ e \end{bmatrix}$. (Hint: what are the columns of $[T]$?)

At this point, we can write down a reasonable formula for the determinant of a 3×3 matrix. Let

$$[T] = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix},$$

then

$$\det[T] = c \det \begin{pmatrix} d & e \\ g & h \end{pmatrix} - f \det \begin{pmatrix} a & b \\ g & h \end{pmatrix} + i \det \begin{pmatrix} a & b \\ d & e \end{pmatrix}.$$

Here we've singled out the last column, but we can do the same thing by picking out any row or column. To do this properly, we need some notation. Let $[T] = [A_{ij}]$, so that the entry of $[T]$ in the i th row, j th column is A_{ij} . Also, let $[\bar{T}_{ij}]$ be the 2×2 matrix you get from $[T]$ by crossing out the i th row and j th column. Then for any choice of $j = 1, 2, 3$ we can write

$$\det([T]) = (-1)^{1+j} A_{1j} \det([\bar{T}_{1j}]) + (-1)^{2+j} A_{2j} \det([\bar{T}_{2j}]) + (-1)^{3+j} A_{3j} \det([\bar{T}_{3j}]),$$

which computes $\det([T])$ by expanding along the j th column. Alternatively, for any choice of $i = 1, 2, 3$ we can write

$$\det([T]) = (-1)^{i+1} A_{i1} \det([\bar{T}_{i1}]) + (-1)^{i+2} A_{i2} \det([\bar{T}_{i2}]) + (-1)^{i+3} A_{i3} \det([\bar{T}_{i3}]),$$

which computes $\det([T])$ by expanding along the i th row.

The same idea will compute the determinant of any square matrix inductively. That is, you write the determinant of an $n \times n$ matrix as a sum of determinants of $(n-1) \times (n-1)$ matrices. We write the general formula as follows. Again, we let A_{ij} be the entry of $[T]$ in the i th row, j th

column, and we let $[\bar{T}_{ij}]$ be the $(n-1) \times (n-1)$ matrix you get from $[T]$ by crossing out the i th row and the j th column. Then for any choice of $j = 1, 2, \dots, n$ we compute $\det([T])$ by expanding along the j th column using the formula

$$\det([T]) = (-1)^{1+j} A_{1j} \det([\bar{T}_{1j}]) + (-1)^{2+j} A_{2j} \det([\bar{T}_{2j}]) + \dots + (-1)^{n+j} A_{nj} \det([\bar{T}_{nj}]).$$

Alternatively, for any choice of $i = 1, 2, \dots, n$ we compute $\det([T])$ by expanding along the i th row using the formula

$$\det([T]) = (-1)^{i+1} A_{i1} \det([\bar{T}_{i1}]) + (-1)^{i+2} A_{i2} \det([\bar{T}_{i2}]) + \dots + (-1)^{i+n} A_{in} \det([\bar{T}_{in}]).$$

We summarize some important properties of the determinant here.

1. The determinant is linear in each row and column. That is, if A is an $n \times n$ matrix and $\tilde{A}$ is the same as A except that you multiply the i th row by c , then $\det(\tilde{A}) = c \det(A)$. Also, A_1 and A_2 are the same except at the i th row and A is what you get by adding together the i th row of A_1 and A_2 then $\det(A) = \det(A_1) + \det(A_2)$. The same goes for columns.
2. Consequently, if A is an $n \times n$ matrix and c is a number then $\det(cA) = c^n \det(A)$.
3. An $n \times n$ matrix A is invertible if and only if $\det(A) \neq 0$.
4. In fact, $|\det(A)|$ is the n -dimensional volume of the parallelepiped P which is the image of the unit cube $S = \{0 \leq x_1 \leq 1, \dots, 0 \leq x_n \leq 1\}$ under the linear transformation associated to A . (You can prove this by induction, in a very similar way we got the geometric interpretation for three dimensions from the two-dimensional version.)
5. Let A and B be $n \times n$ matrices, then $\det(AB) = \det(A) \det(B)$.
6. Let A be an $n \times n$ matrix and let $\tilde{A}$ be the matrix you get by swapping adjacent two rows of A (or by swapping two adjacent columns). Then $\det(\tilde{A}) = -\det(A)$.

Trace: Another important number associated to an $n \times n$ matrix is its trace. We let $[T]$ be an $n \times n$ matrix with A_{ij} being the entry in the i th row, j th column. Then the trace of $[T]$ is given by

$$\text{tr}([T]) = A_{11} + A_{22} + \dots + A_{nn},$$

the sum of the entries of $[T]$ on the diagonal running from the top left of $[T]$ to its bottom right. Later on, we'll see that under some conditions (for instance, if $A_{ij} = A_{ji}$) that the trace measures an average distortion of T as a mapping. That is, if the trace is close to n then T doesn't distort distances too much, but if the trace is very different from n then it distorts distances a lot. Remember that this guideline only holds if T is symmetric, that is if $A_{ij} = A_{ji}$. The trace satisfies the following properties:

1. If I is the $n \times n$ identity matrix then $\text{tr}(I) = n$.
2. If A and B are square matrices and c is a number then

$$\text{tr}(A + B) = \text{tr}(A) + \text{tr}(B), \quad \text{tr}(cA) = c \text{tr}(A), \quad \text{tr}(AB) = \text{tr}(BA).$$

Some special matrices: We close with some special types of square matrices. Let $[T]$ be a square matrix, with entries A_{ij} in the i th row, j th column as above. We say $[T]$ is *diagonal* if $A_{ij} = 0$ for $i \neq j$. We say $[T]$ is *upper triangular* if $A_{ij} = 0$ for $i > j$ and that $[T]$ is *lower triangular* if $A_{ij} = 0$ for $i < j$.

Exercise: Why is a diagonal matrix called diagonal? How about upper (or lower) triangular matrices?

Exercise: Let $[T]$ be diagonal, and show that

$$\det([T]) = A_{11} A_{22} \dots A_{nn}, \quad \text{tr}([T]) = A_{11} + A_{22} + \dots + A_{nn}$$

Exercise: Show that the same formula holds for a upper triangular and lower triangular matrices.