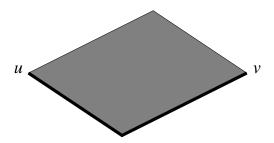
## Mathematics 307—December 6, 1995

## Determinants, areas, and volumes

If u and v are any two 2D vectors, they span a parallelogram. The geometric meaning of determinants is related to the area of this parallelogram.



To understand the relationship exactly, we must first understand what an orientation of the paralellogram is. An orientation is determined by the order in which the vectors u and v are specified. In the picture above, the parallelogram is oriented positively because the smallest angle necessary to rotate u towards v is positive. If it is negative the orientation is negative. If u and v lie on the same line, the parallelogram's orientation is undefined.

**Proposition.** The area of the oriented parallelogram determined by u and v is equal to the determinant of the  $2 \times 2$  matrix

$$M_{u,v} = [u \quad v]$$

whose columns are the coordinates of the vectors u and v.

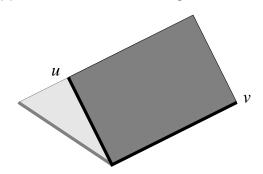
In particular, the determinant vanishes precisely when u and v lie in the same line, or equivalently when u and v are scalar multiples of each other.

The proof of the Proposition is essentially geometrical. We will start with the original pair u and v and transform them by a succession of elementary operations in which we can understand exactly how bothe the area and the determinant is modified. We begin with

**Proposition.** If A and B are two  $2 \times 2$  matrices then det(AB) = det(A) det(B).

This is a messy but straightforward calculation.

Step (1) We shear the vector u along v so as to turn the parallelogram into a rectangle.



Determinants and areas 2

The area remains the same, since a shear does not change the height of the parallelogram spanned by u and v. The vector u is changed to  $u_{\text{new}} = u + cv$  for some constant c, and  $v_{\text{new}}$  isn't changed. So we change  $M_{u,v}$  to

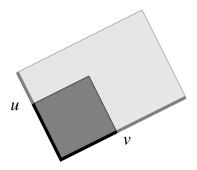
$$\begin{bmatrix} u + cv & v \end{bmatrix} = \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} 1 & 0 \\ c & 1 \end{bmatrix}$$

Because of the multiplication rule for determinants, the determinant doesn't change either.

We can find c explicitly:

$$(u+cv) \cdot v = 0, \quad c = \frac{u \cdot v}{v \cdot v}.$$

Step (2) We scale u and v to make them of unit length.

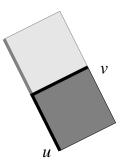


Areas get multiplied by the product of the two scaling factors a and b, which happen to be the inverse of the lengths of u and v. But the new matrix is

$$\begin{bmatrix} au & bv \end{bmatrix} = \begin{bmatrix} u & v \end{bmatrix} \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix}$$

so the determinant also gets multiplied by ab.

Step (3) We can scale u by -1 if necessary to get the orientation positive.



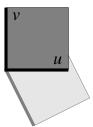
The area changes sign, and the matrix gets multiplied on the right by

$$\begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$

These last two steps can by combined into a single scaling.

Determinants and areas 3

Step (4) We rotate u and v so v lies along the x-axis.



We don't change the area and the matrix gets multiplied by a rotation matrix

$$\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

so the determinant doesn't change either.

Step (4) At the end of this process, u and v span a unit square, and v lies along the x-axis. Typere are two possibilities: either u = (0,1) or v = (0,-1). Here both the oriented area and the determinant are both either  $\pm 1$ , and in any event they agree.

Step (5) So we have changed the pair u, v into a pair where the area and the determinant agree, and we have done this in such a way that the area and the determinant change equally as we proceed. Therefore the area and the determinant were equal to begin with.

There is one simplification we might have made. We could have avoided the possibility of a negative orientation at the end if we had allowed a negative scale factor of u in Step (2). What the description above shows if we start with any matrix M we can find a vertical shear N, a diagonal matrix A, and a rotation matrix K such that MNAK = I. Multiplying by inverses we get

$$M = K^{-1}A^{-1}N^{-1}$$
.

The inverse of a rotation is a rotation, etc. Therefore we have incidentally proved along the way:

**Proposition.** If M is any matrix then we can write

$$M = KAN$$

where N is the matrix of a vertical shear, A diagonal, K a rotation.

There is one problem we have overlooked. The procedure above works only if the vectors we start with are linearly independent. Otherwise we find ourselves trying to divide by 0. In case they are dependent, the Proposition remains valid, but a bit of fiddling is necessary in the argument to allow for vectors which are 0 after the initial shear.

## What happens in 3D

The whole process works also in 3D, once we understand how to perform the original shear. So we are given u, v, w and we want to find a shear that takes these into three mutually perpendicular vectors  $u_*, v_*, w_*$ . We can do this in stages: (1) first of all u stays the same:

Determinants and areas 4

(2) Shear v by a multiple of  $u_*$  to get  $V_*$  perpendicular to  $u_*$ .

$$v_* = v - cu_*$$

where  $cu_*$  is the perpendicular projection of  $v_*$  onto the line through  $u_*$ :

$$c = \frac{v \bullet u_*}{u_* \bullet u_*}$$

(3) Shear  $w_*$  along the plane containing  $u_*$  and  $v_*$ :

$$w_* = w - au_* - bv_*$$

and now  $au_* + bv_*$  is the perpendicular projection of w onto this plane so

$$a = \frac{w \cdot u_*}{u_* \cdot u_*}, \quad b = \frac{w \cdot v_*}{v_* \cdot v_*}$$

These process works in any number of dimensions to change a given set of n linearly independent vectors to a set of mutually perpendicular ones. If we start with  $u_i$  for i = 1 to n we get  $u_{*,i}$  where

$$\begin{aligned} u_{*,1} &= u_1 \\ u_{*,2} &= u_2 - \frac{u_2 \bullet u_{*,1}}{u_{*,1} \bullet u_{*,1}} \ u_{*,1} \\ u_{*,3} &= u_3 - \frac{u_3 \bullet u_{*,1}}{u_{*,1} \bullet u_{*,1}} \ u_{*,1} - \frac{u_3 \bullet u_{*,2}}{u_{*,2} \bullet u_{*,2}} \ u_{*,2} \\ & \dots \\ u_{*,n} &= u_n - \frac{u_n \bullet u_{*,1}}{u_{*,1} \bullet u_{*,1}} \ u_{*,1} - \frac{u_n \bullet u_{*,2}}{u_{*,2} \bullet u_{*,2}} \ u_{*,2} - \dots - \frac{u_n \bullet u_{*,n-1}}{u_{*,n-1} \bullet u_{*,n-1}} \ u_{*,n-1} \end{aligned}$$

This process is called **Gram-Schmid orthogonalization**. Again, it only works completely when the original vectors  $u_i$  are linearly independent.