

## Two Determinant Proofs

In this short write-up we will prove two basic identities for determinants:

$$\begin{aligned} |\mathbf{AB}| &= |\mathbf{A}| |\mathbf{B}| \\ |\mathbf{A}| &= |\mathbf{A}^T| \end{aligned}$$

$$|\mathbf{AB}| = |\mathbf{A}| |\mathbf{B}|$$

Strang does something interesting here (proving the  $n \times n$  case by algebra is complicated). If  $\mathbf{B}$  is zero, this equation is certainly true.

If  $\mathbf{B}$  is non-zero then consider this ratio:

$$D(A) = \frac{|\mathbf{AB}|}{|\mathbf{B}|}$$

Does  $D(A)$  have the first three properties that define the determinant of  $A$ ? If  $A = I$  then this becomes

$$D(A) = \frac{|\mathbf{B}|}{|\mathbf{B}|} = 1$$

For property 2, if two rows of  $A$  are exchanged, then so are the same two rows of  $AB$  and therefore  $|AB|$  changes sign, and so does the ratio  $D(A)$ .

Finally property 3 (linearity) also holds (check for yourself).

Since the ratio  $D(A)$  has the same three properties that define  $|A|$ , it equals  $|A|$ .

Note that if this is given then:

$$\begin{aligned} AA^{-1} &= I \\ |A| |A^{-1}| &= |I| = 1 \\ |A| &= \frac{1}{|A^{-1}|} \end{aligned}$$

$$|\mathbf{A}| = |\mathbf{A}^T|$$

The simplest way to see this is use the rules for computing determinants of e.g.  $3 \times 3$

$$\begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = a(ei - hf) - b(di - fg) + c(dh - eg)$$

Now, the transpose is

$$\begin{vmatrix} a & d & g \\ b & e & h \\ c & f & i \end{vmatrix} = a(ei - hf) - d(bi - ch) + g(bf - ec)$$

All the same terms are there. But we haven't proved these rules yet

To Prove:  $|\mathbf{A}| = |\mathbf{A}^T|$

Either  $\mathbf{A}$  is *singular* with determinant zero (and so is its transpose), or it can be factored into triangular matrices  $\mathbf{L}$  and  $\mathbf{U}$ .

To do this, the rows of  $\mathbf{A}$  may need to be reordered first by multiplication with a permutation matrix  $\mathbf{P}$ . Thus

$$\mathbf{PA} = \mathbf{LU}$$

Let's start with the permutation matrix.  $\mathbf{P}$  has determinant either  $\pm 1$ , by rule 2 (from the previous write-up).

For any permutation matrix, the transpose is also the inverse

$$\mathbf{P}^T = \mathbf{P}^{-1}$$

so

$$\mathbf{PP}^T = \mathbf{I}$$

(since any row or column of either has at most a single 1, and those match up in the row by column multiplication because of the transpose.

We proved above that

$$|\mathbf{AB}| = |\mathbf{A}| |\mathbf{B}|$$

Let's call this the product rule. By using this rule and then that the determinant of the identity matrix is 1

$$|\mathbf{P}| |\mathbf{P}^T| = |\mathbf{PP}^T| = |\mathbf{I}| = 1$$

Now, either  $|\mathbf{P}| = 1$  and so, by the last equation,  $|\mathbf{P}^T| = 1$ , or else both determinants are minus 1.

In either case

$$|\mathbf{P}| = |\mathbf{P}^T|$$

Moving to the right-hand side

$$|\mathbf{U}| = \begin{vmatrix} p & a & b \\ 0 & q & c \\ 0 & 0 & r \end{vmatrix}$$

By rule 7 (previous write-up),  $|\mathbf{U}| = pqr$ . Transposition doesn't change the values on the diagonal, so the determinant of a triangular matrix is not changed by the transpose.

$$|\mathbf{U}| = |\mathbf{U}^T|$$

There is a particular decomposition into  $\mathbf{LU}$  in which  $\mathbf{L}$  has 1's on the diagonal, although that's not required for what comes next. But in that case,  $|\mathbf{L}| = 1$ . Otherwise

$$|\mathbf{L}| = \begin{vmatrix} u & 0 & 0 \\ x & v & 0 \\ y & z & w \end{vmatrix} = 1$$

and so (again, since transposition doesn't change the diagonal entries),  $|L| = |L^T|$ .

### putting it all together

We have that

$$\mathbf{PA} = \mathbf{LU}$$

and we have shown that individually, each of  $\mathbf{P}$ ,  $\mathbf{L}$  and  $\mathbf{U}$  has the same determinant as the respective transpose. Further, we have the product rule,  $|\mathbf{AB}| = |\mathbf{A}||\mathbf{B}|$ .

Now since

$$\mathbf{PA} = \mathbf{LU}$$

$$|\mathbf{PA}| = |\mathbf{LU}|$$

and

$$(\mathbf{PA})^T = (\mathbf{LU})^T$$

$$|(\mathbf{PA})^T| = |(\mathbf{LU})^T|$$

Looking first at the right hand side, by the property of the transpose

$$|(\mathbf{LU})^T| = |\mathbf{U}^T \mathbf{L}^T|$$

by the product rule

$$|\mathbf{U}^T \mathbf{L}^T| = |\mathbf{U}^T| |\mathbf{L}^T|$$

by the identities above

$$= |\mathbf{U}| |\mathbf{L}|$$

$$= |\mathbf{L}| |\mathbf{U}|$$

by the product rule

$$= |\mathbf{LU}|$$

In summary

$$|(\mathbf{PA})^T| = |\mathbf{LU}|$$

by our initial formulation

$$|\mathbf{PA}| = |\mathbf{LU}|$$

Thus, we have shown that

$$|\mathbf{PA}| = |(\mathbf{PA})^T|$$

but by the product rule

$$|\mathbf{PA}| = |\mathbf{P}| |\mathbf{A}|$$

and by the property of the transpose

$$(\mathbf{PA})^T = \mathbf{A}^T \mathbf{P}^T$$

so

$$|(\mathbf{PA})^T| = |\mathbf{A}^T \mathbf{P}^T|$$

by the product rule

$$= |\mathbf{A}^T| |\mathbf{P}^T|$$

Thus

$$\begin{aligned} |\mathbf{PA}| &= |(\mathbf{PA})^T| \\ |\mathbf{P}||\mathbf{A}| &= |\mathbf{A}^T||\mathbf{P}^T| \end{aligned}$$

but

$$|\mathbf{P}| = |\mathbf{P}^T|$$

so

$$|\mathbf{A}| = |\mathbf{A}^T|$$