Week 4 Part 1: Determinants

While determinants aren't traditionally covered in this course, they do make understanding eigenvalues and eigenvectors much easier.

The determinant is a way to measure the "size" of a matrix. Think of it as the analog of absolute value in real numbers, or the modulus in complex numbers. Lets look at examples of 2×2 . I won't expect you to find 3×3 determinants by hand, but we will do a lot of that on the computer.

$$2 \times 2$$

Here is the generic 2×2 determinant.

$$\det \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21} \tag{1}$$

Now lets look at some examples

Ex:

$$\begin{vmatrix} 2 & 1 \\ 3 & 4 \end{vmatrix} = 8 - 3 = 5.$$

Ex:

$$\begin{vmatrix} 5 & 2 \\ -6 & 3 \end{vmatrix} = 15 + 12 = 27$$

If anyone happens to be interested in the properties of determinants and how to calculate higher dimensional determinants, you can continue below.

 3×3

For higher order determinants such as 3×3 , we use cofactor expansion. While there are some "shortcuts", I personally don't see them saving much time.

$$\det \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{13} & a_{32} & a_{33} \end{pmatrix} = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{13} & a_{32} & a_{33} \end{vmatrix} = \mathbf{a_{11}} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - \mathbf{a_{21}} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + \mathbf{a_{13}} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$
(2)

Then just take the determinants of each 2×2 matrix and you are done! For a 4×4 matrix you do the same thing, but you would get 3×3 cofactors.

Ex:

$$\begin{vmatrix} 1 & 4 & -2 \\ 3 & 2 & 0 \\ -1 & 4 & 3 \end{vmatrix} = 1 \begin{vmatrix} 2 & 0 \\ 4 & 3 \end{vmatrix} - 4 \begin{vmatrix} 3 & 0 \\ -1 & 3 \end{vmatrix} + (-2) \begin{vmatrix} 3 & 2 \\ -1 & 4 \end{vmatrix} = -58$$

Ex:

$$\begin{vmatrix} 2 & 4 & 6 \\ 0 & 3 & 1 \\ 0 & 0 & -5 \end{vmatrix} = 2 \begin{vmatrix} 3 & 1 \\ 0 & -5 \end{vmatrix} = 2 \cdot 3 \cdot (-5).$$

Notice that since we have an upper triangular matrix, all we need to do is take the product of the diagonal: $\det = 2 \cdot 3 \cdot (-5) = -30$

Properties of Determinants

We will list these generically, but illustrate with 2×2 matrices.

(1) $\det(I) = 1$. E.g.,

$$\begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} = 1; \qquad \begin{vmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{vmatrix} = 1; \qquad \text{etc.}$$

(2) Row exchange changes sign of the determinant.

E.g.,

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc = -(bc - ad) = - \begin{vmatrix} c & d \\ a & b \end{vmatrix}$$

(3) If two rows of A are equal, then det(A) = 0E.g.,

$$\begin{vmatrix} a & b \\ a & b \end{vmatrix} = ab - ba = 0.$$

(4) If A has a row (or column) of zeros, then det(A) = 0.

$$\begin{vmatrix} 0 & 0 \\ c & d \end{vmatrix} = 0; \qquad \begin{vmatrix} 0 & b \\ 0 & d \end{vmatrix} = 0.$$

(5) If $A_{n\times n}$ is triangular (either lower or upper) then $\det(A) = a_{11}a_{22}a_{33}\cdots a_{nn}$; i.e., take the product of the diagonal elements.

E.g.,

$$\begin{vmatrix} a & 0 \\ c & d \end{vmatrix} = ad; \qquad \begin{vmatrix} a & b \\ 0 & d \end{vmatrix} = ad.$$

Can we prove this for $n \times n$ matrices?

Handwavey proof. Consider,

$$D = \begin{bmatrix} a_{11} & \cdots & \cdots & \cdots & \cdots \\ & a_{22} & \cdots & \cdots & \cdots \\ & & \ddots & & & \\ & & 0 & & \ddots & & \\ & & & & a_{nn} \end{bmatrix}$$

Then

- (6) If A is singular, then $\det(A) = 0$; otherwise $\det(A) \neq 0$, then A is invertible. If A is in row-echelon form (not to be confused with the full augmented matrix), but not upper triangular, then at least one of the pivots will be zero.
- (7) Product rule: |A||B| = |AB|.

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} \begin{vmatrix} e & f \\ g & h \end{vmatrix} = (ad - bc)(eh - fg) = (ae + bg)(cf + dh) - (af + bh)(ce + dg) = \begin{vmatrix} ae + bg & af + bh \\ ce + dg & cf + dh \end{vmatrix}$$

However, for scalar multiplication, $|cD_{n\times n}|=c^n|D_{n\times n}|$ because this can be written as

$$cD_{n\times n} = \begin{bmatrix} c & & & \\ & c & 0 & \\ 0 & & \ddots & \\ & & & c \end{bmatrix}$$

A special case is $|A||A^{-1}|=|AA^{-1}|=|I|=1\Rightarrow |A^{-1}|=1/|A|$ (8) Transpose: $|A|=|A^T|.$

E.g.,

$$|A| = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = \begin{vmatrix} a & c \\ b & d \end{vmatrix} = |A^T|.$$