

# LECTURE 14 PROPERTIES OF THE DETERMINANT

## Calculating $\det M$

$$\det M = \sum_{\sigma}^1 \operatorname{sgn}(\sigma) m_{\sigma(1)}^1 m_{\sigma(2)}^2 \cdots m_{\sigma(n)}^n$$

$$= m_1^1 \sum_{\hat{\sigma}} \operatorname{sgn}(\hat{\sigma}) m_{\hat{\sigma}(2)}^2 \cdots m_{\hat{\sigma}(n)}^n$$

$\hat{\sigma}$  are permutations over  $n-1$  objects

$$- m_2^1 \sum_{\hat{\sigma}} \operatorname{sgn}(\hat{\sigma}) m_{\hat{\sigma}(1)}^1 \cdots m_{\hat{\sigma}(n)}^n$$

$$+ m_3^1 \sum_{\hat{\sigma}} \operatorname{sgn}(\hat{\sigma}) m_{\hat{\sigma}(1)}^1 m_{\hat{\sigma}(2)}^2 \cdots m_{\hat{\sigma}(n)}^n$$



:



Expanding sum in 1<sup>st</sup> row.

Notice signs alternate

Each sum is the determinant of the matrix left over after removing the 1<sup>st</sup> row and a column corresponding to which  $m_i^1$  is factored out front.

## EXAMPLE

$$\det \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

$$= 1 \det \begin{pmatrix} 5 & 6 \\ 8 & 9 \end{pmatrix} - 2 \det \begin{pmatrix} 4 & 6 \\ 7 & 9 \end{pmatrix} + 3 \det \begin{pmatrix} 4 & 5 \\ 7 & 8 \end{pmatrix}$$

expand  
in 2nd  
row

$$= (5 \cdot 9 - 6 \cdot 8) - 2(4 \cdot 9 - 6 \cdot 7) + 3(4 \cdot 8 - 5 \cdot 7) = 0$$

so  $\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}^{-1}$  does NOT exist.

## EXAMPLE

$$\det \begin{pmatrix} 1 & 2 & 3 \\ 4 & 0 & 0 \\ 5 & 6 & 7 \end{pmatrix} \xrightarrow{\text{swapped rows}} - \det \begin{pmatrix} 4 & 0 & 0 \\ 1 & 2 & 3 \\ 5 & 6 & 7 \end{pmatrix}$$

$$= 4 \det \begin{pmatrix} 2 & 3 \\ 6 & 7 \end{pmatrix} = -16$$

The same method works expanding in any column because

$$\boxed{\det M^T = \det M}$$

This is true because

$$\sum_{\sigma} \text{sgn}(\sigma) m'_{\sigma(1)} \dots m'_{\sigma(n)} = \sum_{\sigma} \text{sgn}(\sigma) m''_1 \dots m''_n$$

i.e. commuting rows or columns  
entries gives same sum of terms.

### DETERMINANT OF THE INVERSE

Let  $A, B$  be  $n \times n$  matrices

Recall we found

$$\det AB = \det A \det B$$

$$\det I = 1$$

$$\Rightarrow 1 = \det A^{-1}A = \det A^{-1} \det A$$

$$\Rightarrow \boxed{\det A^{-1} = \frac{1}{\det A}}$$

Writing matrices as products of  
simpler ones is a good trick  
in computing the determinant.

## INVERSES AGAIN

The adjoint of a matrix  $M = (m_{j|i})$

$$\text{adj } M = (\text{cofactor}(m_{j|i}))^T$$

where the cofactor  $(m_{j|i})$  is  
the determinant of  $M$  with  
row  $i$  and column  $j$  removed and  
multiplied by  $(-1)^{i+j}$

Ex  $\text{adj} \begin{pmatrix} 3 & -1 & -1 \\ 1 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix}$

$$= \begin{pmatrix} \det(2, 0) & -\det(1, 0) & \det(1, 2) \\ -\det(-1, -1) & \det(3, -1) & -\det(3, -1) \\ \det(-1, -1) & -\det(3, 0) & \det(3, 2) \end{pmatrix}^T$$

Notice That the dot product of  
the  $i^{\text{th}}$  row of  $M$  & the  $i^{\text{th}}$  row of  
 $(\text{adj } M)^T$  is just the expansion of  
 $\det M$  in the  $i^{\text{th}}$  row.

For the  $i^{\text{th}}$  and  $j^{\text{th}}$  rows ( $i \neq j$ ) the dot product amounts to expanding  $\det M$  in the  $i^{\text{th}}$  row save that the  $j^{\text{th}}$  row of  $M$  is replaced by its  $i^{\text{th}}$  row. The determinants of a matrix with two equal rows vanish, so these dot products do as well.

Finally note that computing dot products of the rows of  $M \& N^T$  amounts to computing the elements of  $MN$ .

Hence

$$M \operatorname{adj}(M) = \det M \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & & \vdots \\ & & \ddots & 0 \\ 0 & \cdots & 0 & 1 \end{pmatrix}$$

Thus computing  $\operatorname{adj}(M)$  gives us a formula for  $M^{-1}$  when it exists.

$$M^{-1} = \frac{1}{\det M} \text{adj}(M)$$

when  $\det M \neq 0$ .

Ex In the above example

$$\text{adj} \begin{pmatrix} 3 & -1 & -1 \\ 1 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 0 & 2 \\ -1 & 3 & -1 \\ 1 & -3 & 7 \end{pmatrix}$$

and

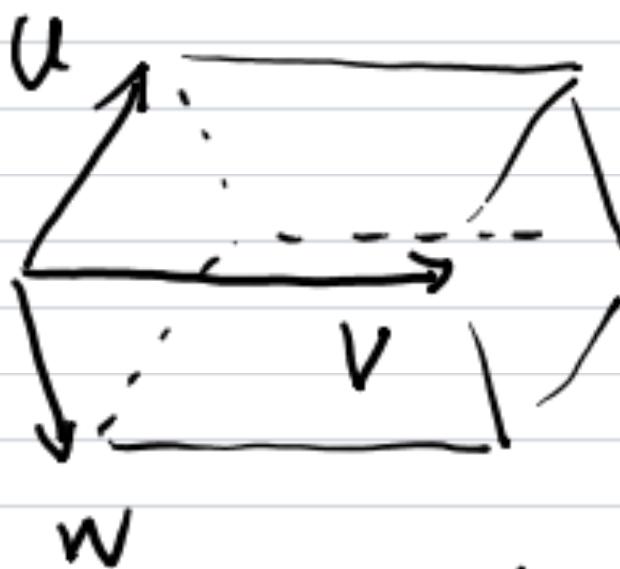
$$\begin{pmatrix} 2 & 0 & 2 \\ -1 & 3 & -1 \\ 1 & -3 & 7 \end{pmatrix} \begin{pmatrix} 3 & -1 & -1 \\ 1 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix} = \begin{pmatrix} 6 & 0 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 6 \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} 3 & -1 & -1 \\ 1 & 2 & 0 \\ 0 & 1 & 1 \end{pmatrix}^{-1} = \frac{1}{6} \begin{pmatrix} 2 & 0 & 2 \\ -1 & 3 & -1 \\ 1 & -3 & 7 \end{pmatrix} //$$

This is sometimes called Cramer's rule.

## APPLICATION

Volume of parallelepiped



$$\text{Volume} = |\det(u \ v \ w)|$$

This comes from the formula you know which says the volume is  $|u \cdot (v \times w)|$  and using the fact that this amounts to computing the determinant by expanding in the first column.

Do you think a similar formula could work for computing higher-dimensional volumes?

## Lecture 14 Review Questions:

① Let  $M = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$

Show

$$\det M = \frac{1}{2} \operatorname{tr} M^2 - \frac{1}{2} (\operatorname{tr} M)^2$$

Suppose  $M$  is a  $3 \times 3$  matrix.

Find and verify a similar formula for  $\det M$  in terms of  $\operatorname{tr} M^3$ ,  $\operatorname{tr} M \cdot \operatorname{tr} M^2$ ,  $(\operatorname{tr} M)^3$ .

② Suppose  $M = LU$  is an LU decomposition. Explain how you would compute  $\det M$  efficiently in this case.

