

## Lecture 30 §6.4 Product Rule & Cramer's Rule

Recall: Row operations change determinants in specific ways.  $A \longrightarrow B$

Operation	$\det(B)$	Net Effect
(I) $R_i \leftrightarrow R_j$ <small><math>i \neq j</math> (exchange 2 rows)</small>	$-\det(A)$	sign switch
(II) $R_i \rightarrow \alpha R_i$ <small><math>\alpha \neq 0</math> scalar (multiply a row by a nonzero scalar)</small>	$\alpha \det(A)$	multiply by scalar $\alpha$
(III) $R_i \rightarrow R_i + \beta R_j$ <small><math>i \neq j</math> (Add to a row a scalar multiple of a <u>different</u> row)</small>	$\det(A)$	no change

. Assuming we know  $\det(A^T) = \det(A)$ , we can derive similar rules for analogous column operations on A  $\longleftrightarrow$  row operations on A<sup>T</sup>.

TODAY : ①  $\det(AB) = \det A \det B$  (Product Rule)  
 ② Cramer's Rule for solving  $Ax=b$  with A invertible via determinants

## Product Rule

Theorem 1:  $A, B$  of size  $n \times n$ , then  $\det(AB) = \det(A) \cdot \det(B)$ .

Example:

$$A = \begin{bmatrix} 2 & 2 \\ 1 & 3 \end{bmatrix}$$

$$\det A = 2 \cdot 3 - 2 \cdot 1 = 4 \neq 0$$

$$B = \begin{bmatrix} -1 & 3 \\ 1 & -2 \end{bmatrix}$$

$$\det B = (-1)(-2) - 3 \cdot 1 = -1 \neq 0$$

$$AB = \begin{bmatrix} 2 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} -1 & 3 \\ 1 & -2 \end{bmatrix} = \begin{bmatrix} -2+2 & 2 \cdot 3 - 4 \\ -1+3 & 3-6 \end{bmatrix} = \begin{bmatrix} 0 & 2 \\ 2 & -4 \end{bmatrix} \Rightarrow \det(AB) = -4$$

We saw a proof for general  $2 \times 2$  matrices in Lecture 28.

Consequence: If  $A$  invertible then  $AA^{-1} = I_n$ , so  $\det(A) \det(A^{-1}) = 1$ .

Then  $\det A \neq 0$  &  $\det(A^{-1}) = \frac{1}{\det A}$ . However, if  $\det(A) = 0$ , then  $A$  is singular.

Q: Why is the product formula valid in general?

Special case:  $AB$  singular matrix. Then:  $A$  or  $B$  are singular (Lecture 9)

Conclude  $\det(AB) = 0 = \det(A) \det(B)$ . ( $\det(\cdot) = 0$  for the singular one).

Lemma: If  $A$  is singular of size  $n \times n$ , &  $B$  is any matrix of size  $n \times n$ , then  $AB$  is also singular.

Q: Why is this true? Use  $C$  singular if, and only if,  $C^T$  is singular

(Reason: If  $C$  invertible, then  $(C^{-1})^T = (C^T)^{-1}$  so  $C^T$  is also invertible)

• Since  $A$  is singular, so is  $A^T$ . Now  $(AB)^T = B^T A^T$

so  $B^T A^T$  is singular | Pick  $\vec{x} \neq \vec{0}$  in  $\mathbb{R}^n$  with  $A^T \vec{x} = \vec{0}$  Then  
 $(B^T A^T) \vec{x} = B^T(\underbrace{A^T \vec{x}}_{= \vec{0}}) = \vec{0}$ )

• Then  $AB = (B^T A^T)^T$  is singular

Proof of the Product Rule:

① If  $A$  is singular, then  $AB$  is also singular, so  $\det(AB) = 0 = \boxed{\det(A)\det(B)}$ .

② If  $A$  is not singular, then  $A \sim I_n$  with a sequence of row operations.  
 $\Rightarrow \beta \det(A) = \det(I_n) = 1$  for some  $\beta \neq 0$  (coming from rule rules)

(\*)  $B$ UT The same row operations give  $AB \sim B$  so  $\beta \det(AB) = \det(B)$   
 since  $\beta = \frac{1}{\det A}$ , we get  $\det(AB) = \det(B)/\beta = \det A \det B$ .

. Let's check our claim (k) in an example.

Example  $A = \begin{bmatrix} 2 & 2 \\ 1 & 3 \end{bmatrix}$  &  $B = \begin{bmatrix} -1 & 3 \\ 1 & -2 \end{bmatrix} \Rightarrow AB = \begin{bmatrix} 0 & 2 \\ 2 & -3 \end{bmatrix}$

We write the operations needed to turn  $A$  into  $I_2$  & check the same sequence of elementary row operations turns  $AB$  into  $B$ .

$$A = \begin{bmatrix} 2 & 2 \\ 1 & 3 \end{bmatrix} \xrightarrow[\text{R}_2 \leftrightarrow R_1]{(I)} \begin{bmatrix} 1 & 3 \\ 2 & 2 \end{bmatrix} \xrightarrow[R_2 \rightarrow R_2 - 2R_1]{(III)} \begin{bmatrix} 1 & 3 \\ 0 & -4 \end{bmatrix} \xrightarrow[R_2 \rightarrow \frac{R_2}{-4}]{(II)} \begin{bmatrix} 1 & 3 \\ 0 & 1 \end{bmatrix} \xrightarrow[R_1 \rightarrow R_1 - 3R_2]{(III)} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_2$$

$\det A$        $-\det A$        $-\det A$        $(\frac{-1}{4})(-\det A)$        $\frac{1}{4} \det A$

$$AB = \begin{bmatrix} 0 & 2 \\ 2 & -3 \end{bmatrix} \xleftarrow[R_2 \leftrightarrow R_1]{(I)} \begin{bmatrix} 2 & -3 \\ 0 & 2 \end{bmatrix} \xrightarrow[R_2 \rightarrow R_2 - 2R_1]{(III)} \begin{bmatrix} 2 & -3 \\ -4 & 8 \end{bmatrix} \xrightarrow[R_2 \rightarrow \frac{R_2}{-4}]{(II)} \begin{bmatrix} 2 & -3 \\ 1 & -2 \end{bmatrix} \xrightarrow[R_1 \rightarrow R_1 - 3R_2]{(III)} \begin{bmatrix} -1 & 3 \\ 1 & -2 \end{bmatrix} = B$$

$\det AB$        $-\det AB$        $-\det(AB)$        $-\frac{1}{4}(-\det AB)$        $\frac{1}{4} \det AB$

In general: Need To see that  $A \xrightarrow[\text{op.}]{ } A'$  gives  $AB \xrightarrow[\text{same op.}]{ } A'B$

for any of the 3 elementary row operations.

After this fact, we concatenate all operations and get  $A' = I_n$  so  $A'B = B$ .

## Cramer's Rule

GOAL Given a non-singular  $n \times n$  matrix  $A$ , we want to use determinants to find the unique solution to  $A \underline{x} = \underline{b}$  for a fixed  $\underline{b}$  in  $\mathbb{R}^n$ .

Theorem 2 (Cramer's Rule) Write  $A = [A_1 \cdots A_n]$  &  $\underline{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$

For each  $i=1, 2, \dots, n$  we build a new matrix

$$B_i = [A_1 \cdots A_{i-1} \underline{b} \ A_{i+1} \cdots A_n] \quad (\text{replace } i^{\text{th}} \text{ col of } A \text{ with } \underline{b})$$

Then, the unique solution to  $A \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$  satisfies:

$$x_1 = \frac{\det(B_1)}{\det(A)}, \quad x_2 = \frac{\det(B_2)}{\det(A)}, \quad \dots, \quad x_n = \frac{\det(B_n)}{\det(A)}.$$

Example: ① Solve  $\begin{bmatrix} 2 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

$$B_1 = \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix} \rightsquigarrow \det(B_1) = 3 - 4 = -1$$

$$B_2 = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \rightsquigarrow \det(B_2) = 4 - 1 = 3$$

$$\det A = 6 - 2 = 4 \neq 0$$

$$x_1 = \frac{-1}{4}, \quad x_2 = \frac{3}{4}$$

$$\left. \begin{array}{l} \text{Check: } \begin{bmatrix} 2 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} -\frac{1}{4} \\ \frac{3}{4} \end{bmatrix} = \begin{bmatrix} \frac{-2}{4} + \frac{6}{4} \\ \frac{1}{4} + \frac{9}{4} \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \end{array} \right. \checkmark$$

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$$B_i = [A_1 \cdots A_{i-1} \ b_i \ A_{i+1} \cdots A_n] \quad (\text{replace } i^{\text{th}} \text{ col of } A \text{ with } b)$$

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Example: ② solve  $\begin{bmatrix} 2 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$

$$B_1 = \begin{bmatrix} 3 & 2 \\ 4 & 3 \end{bmatrix} \rightsquigarrow \det(B_1) = 9 - 8 = 1$$

$$B_2 = \begin{bmatrix} 2 & 3 \\ 1 & 4 \end{bmatrix} \rightsquigarrow \det(B_2) = 8 - 3 = 5$$

$$\det A = 6 - 2 = 4 \neq 0$$

$$x_1 = \frac{1}{4}, \quad x_2 = \frac{5}{4}$$

$$\left. \begin{array}{l} \text{Check: } \begin{bmatrix} 2 & 2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} \frac{1}{4} \\ \frac{5}{4} \end{bmatrix} = \begin{bmatrix} \frac{2}{4} + \frac{10}{4} \\ \frac{1}{4} + \frac{15}{4} \end{bmatrix} = \begin{bmatrix} 3 \\ 4 \end{bmatrix} \end{array} \right\} \checkmark$$

## Proof of Cramer's Rule

$$A = [A_1 \cdots A_n] \text{ non-singular}$$

$$A \underline{x} = \underline{b}$$

We check  $x_1 = \frac{\det([b \ A_2 \cdots A_n])}{\det(A)}$ . Others are similar.

$$\bullet A \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = x_1 \text{ col}_1 A + x_2 \text{ col}_2 A + \cdots + x_n \text{ col}_n A = x_1 A_1 + \cdots + x_n A_n.$$

$$\underline{b} = \underbrace{x_1 A_1}_{\text{scalars!}} + \cdots + \underbrace{x_n A_n}_{\text{scalars!}}$$

$$\text{so } \underline{b}^T = (x_1 A_1 + \cdots + x_n A_n)^T = x_1 A_1^T + \cdots + x_n A_n^T$$

$1 \times n$  matrix

$$\Rightarrow [b_1 \cdots b_n] = [x_1 \cdots x_n] \begin{bmatrix} A_1^T \\ \vdots \\ A_n^T \end{bmatrix} = [x_1 \cdots x_n] A^T.$$

Next, we build a new matrix  $C$  from  $A^T$ , by operation (II)  $R_i \rightarrow x_i R_i$

$$A^T \rightarrow C = \begin{bmatrix} x_1 A_1^T \\ A_2^T \\ \vdots \\ A_n^T \end{bmatrix}$$

$R_i \rightarrow x_i R_i$

$$\Rightarrow \det C = x_1 \det(A^T) = x_1 \det(A).$$

$$\text{But } x_1 A_1^T = \underline{b}^T - x_2 A_2^T - \cdots - x_n A_n^T$$

$$\text{So far : } \left\{ \begin{array}{l} x_1 A_1^T = b^T - x_2 A_2^T - \cdots - x_n A_n^T \\ \det \begin{bmatrix} x_1 A_1^T \\ A_2^T \\ \vdots \\ A_n^T \end{bmatrix} = x_1 \det(A) \end{array} \right.$$

$$\begin{bmatrix} x_1 A_1^T \\ A_2^T \\ \vdots \\ A_n^T \end{bmatrix} = \begin{bmatrix} b^T - x_2 A_2^T - \cdots - x_n A_n^T \\ A_2^T \\ \vdots \\ A_n^T \end{bmatrix} \xrightarrow[\substack{R_1 \rightarrow R_1 - x_n R_n \\ (\text{III})}]{} \begin{bmatrix} b^T - x_2 A_2^T - \cdots - x_{n-1} A_{n-1}^T \\ A_2^T \\ \vdots \\ A_{n-1}^T \\ A_n^T \end{bmatrix}$$

$x_1 \det(A)$

$x_1 \det(A)$

$$\xrightarrow[\substack{R_1 \rightarrow R_1 - x_{n-1} R_{n-1}^T \\ (\text{II})}]{} \begin{bmatrix} b^T - x_2 A_2^T - \cdots - x_{n-2} A_{n-2}^T \\ A_2^T \\ \vdots \\ A_n^T \end{bmatrix} \xrightarrow[\substack{R_1 \rightarrow R_1 - x_2 R_2 \\ (\text{III})}]{} \cdots \xrightarrow[\substack{R_1 \rightarrow R_1 - x_2 R_2 \\ (\text{III})}]{} \begin{bmatrix} b^T \\ A_2^T \\ \vdots \\ A_n^T \end{bmatrix} = B_1^T$$

$x_1 \det(A)$

$$\text{So } \det B_1 = \det B_1^T = x_1 \det(A) \quad \text{true}$$

$$x_1 = \frac{\det B_1}{\det A}$$

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