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## Matrix Multiplication

Let  $A \in \mathbb{F}^{m \times n}$  and  $B \in \mathbb{F}^{n \times p}$ , then  $AB \in \mathbb{F}^{m \times p}$

$$(AB)_{ik} = \sum_{j=1}^n a_{ij} b_{jk}$$

### Column Multiplication

$$A = (\mathbf{a}_1 \cdots \mathbf{a}_n)$$

$$(AB)_{:r} = b_{1r}\mathbf{a}_1 + b_{2r}\mathbf{a}_2 + \cdots + b_{nr}\mathbf{a}_n$$

### Row Multiplication

$$B = \begin{pmatrix} \mathbf{b}_1 \\ \vdots \\ \mathbf{b}_n \end{pmatrix}$$

$$(AB)_{r:} = a_{r1}\mathbf{b}_1 + a_{r2}\mathbf{b}_2 + \cdots + a_{rn}\mathbf{b}_n$$

## Uniqueness of Reduced Row Echelon Form

$$A' = EA \Rightarrow \text{row}(A') = \text{row}(A)$$

The row operations are:

1. interchange different rows
2. multiply rows by nonzero scalar
3. add a nonzero multiple of another row

We show  $A$  has equivalent row space under row operations.

Type 1 is immediate.

Type 2 replaces  $\mathbf{a}_i$  by  $r\mathbf{a}_i$ , so we just rescale by  $1/r$ .

$$c_1\mathbf{a}_1 + \cdots + c_n\mathbf{a}_n = \frac{c_1}{r}\mathbf{a}'_1 + \cdots + c_n\mathbf{a}_n$$

Type 3 replaces  $\mathbf{a}_i$  by  $\mathbf{a}_i + r\mathbf{a}_j$

$$\begin{aligned} c_1\mathbf{a}_1 + c_2\mathbf{a}_2 + \cdots + c_n\mathbf{a}_n &= c_1(\mathbf{a}_1 + r\mathbf{a}_2) + (c_2 - rc_1)\mathbf{a}_2 + \cdots + c_n\mathbf{a}_n \\ &= c_1\mathbf{a}'_1 + (c_2 - rc_1)\mathbf{a}'_2 + \cdots + c_n\mathbf{a}'_n \end{aligned}$$

So  $A$  and  $A'$  have the same row space.

$$A = B : A, B \in \mathbf{Red} \Leftrightarrow \mathbf{row}(A) = \mathbf{row}(B)$$

$A = B \Rightarrow \mathbf{row}(A) = \mathbf{row}(B)$  is obvious so we prove the reverse direction.

Label the rows of  $A, B$  like so starting from the bottom.

$$A = \begin{pmatrix} \mathbf{a}_n \\ \vdots \\ \mathbf{a}_1 \end{pmatrix}, \quad B = \begin{pmatrix} \mathbf{b}_n \\ \vdots \\ \mathbf{b}_1 \end{pmatrix}$$

We induct on the pivots starting with  $\mathbf{a}_1, \mathbf{b}_1$ .

1. the pivots for  $\mathbf{a}_1, \mathbf{b}_2$  must be the same otherwise  $\mathbf{a}_1 \notin \mathbf{row}(B)$ .
2. By symmetry, the pivots of  $\mathbf{a}_1$  and  $\mathbf{b}_1$  are in the same component.
3.  $\mathbf{b}_1 = r_1 \mathbf{a}_1 + \dots + r_n \mathbf{a}_n$  but the other components don't share pivots  $\Rightarrow \mathbf{b}_1 = r_1 \mathbf{a}_1$ .
4.  $r_1 = 1$

Keep applying the same argument to see  $A = B$ .

## Reduced Form is Unique

If two different sequences of elementary matrices corresponding to row operations yield two different reduced row echelon forms  $B$  and  $C$  for  $A$ , then by the previous propositions we get:

1.  $\mathbf{row}(A) = \mathbf{row}(B) = \mathbf{row}(C)$
2.  $B = C$

## Exercises

### Ex 3.1.2

$$\begin{aligned} A &= (a_{ij}), \quad A^T = (a_{ij})^T = a_{ji} \\ (A + B)^T &= ((a_{ij}) + (b_{ij}))^T = a_{ji} + b_{ji} = A^T + B^T \end{aligned}$$

### Ex 3.1.5

We use these simple rules:

$$(XY)^T = Y^T X^T$$

$$(X_{:,k})^T = (X^T)_{,k}$$

and the column notation

$$(XY)_{:,k} = Y_{1,k} X_{:,1} + \dots + Y_{n,k} X_{:,n}$$

Putting this all together

$$\begin{aligned} (AB)_{:,k}^T &= (B^T A^T)_{:,k} = (A^T)_{1,k} (B^T)_{:,1} + \dots + (A^T)_{n,k} (B^T)_{:,n} \\ &= A_{k,1} B_{:,1} + \dots + A_{k,n} B_{:,n}, \end{aligned}$$

but  $(AB)_{:,k}^T = (AB)_{:,k}$ ,

## Forms of Matrix Multiplication

### Column and Row Form

$$A = (\mathbf{a}_1 \cdots \mathbf{a}_n), \mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

$$A\mathbf{x} = x_1 \mathbf{a}_1 + \dots + x_n \mathbf{a}_n$$

$$A = \begin{pmatrix} \mathbf{a}_1 \\ \vdots \\ \mathbf{a}_n \end{pmatrix}, \mathbf{x} = (x_1 \cdots x_n)$$

$$\mathbf{x}A = x_1\mathbf{a}_1 + \cdots + x_n\mathbf{a}_n$$

## As a Dot Product

$$A = \begin{pmatrix} \mathbf{a}_1 \\ \vdots \\ \mathbf{a}_n \end{pmatrix}, \mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

$$A\mathbf{x} = \begin{pmatrix} \langle \mathbf{a}_1, \mathbf{x} \rangle \\ \vdots \\ \langle \mathbf{a}_n, \mathbf{x} \rangle \end{pmatrix}$$

$$A = \begin{pmatrix} \mathbf{a}_1 \\ \vdots \\ \mathbf{a}_n \end{pmatrix}, B = (\mathbf{b}_1 \cdots \mathbf{b}_m)$$

$$AB = \begin{pmatrix} \langle \mathbf{a}_1, \mathbf{b}_1 \rangle \cdots \langle \mathbf{a}_1, \mathbf{b}_m \rangle \\ \vdots \\ \langle \mathbf{a}_n, \mathbf{b}_1 \rangle \cdots \langle \mathbf{a}_n, \mathbf{b}_m \rangle \end{pmatrix}$$

A consequence of this is that  $A^T A$ , where  $A = (\mathbf{a}_1 \cdots \mathbf{a}_n)$  is

$$A^T A = \begin{pmatrix} \langle \mathbf{a}_1, \mathbf{a}_1 \rangle \cdots \langle \mathbf{a}_n, \mathbf{a}_n \rangle \\ \vdots \\ \langle \mathbf{a}_n, \mathbf{a}_1 \rangle \cdots \langle \mathbf{a}_n, \mathbf{a}_n \rangle \end{pmatrix}$$

and likewise for  $AA^T$  when  $A = \begin{pmatrix} \mathbf{a}_1 \\ \vdots \\ \mathbf{a}_n \end{pmatrix}$ .

## Matrix as Map on Columns and Rows

$$A = \begin{pmatrix} \mathbf{a}_1 \\ \vdots \\ \mathbf{a}_n \end{pmatrix}, B = (\mathbf{b}_1 \cdots \mathbf{b}_m)$$

$$AB = (A\mathbf{b}_1 \cdots A\mathbf{b}_m)$$

$$= \begin{pmatrix} \mathbf{a}_1 B \\ \vdots \\ \mathbf{a}_n B \end{pmatrix}$$