

# Mathematics for AI

## Lecture 4

linear maps, matrix multiplication,  
matrix rank



# Functions



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$f$ : one-to-one & onto (bijective)

یک به یک و پوشا

Invertible  
عکوس پذیر

bijective  $\exists g$  such that  ~~$f \circ g$~~   $g(f(x)) = x \quad \forall x \in X$

$\exists g: Y \rightarrow X$

$f(g(y)) = y \quad \forall y \in Y$

$g = f^{-1}$

# Functions in linear algebra



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- Here, we are interested in functions from a vector space  $V$  to a vector space  $U$

$$(f: U \rightarrow V)$$



# Linear Transformations



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$$f(u+v) = f(u) + f(v)$$

$$f(a u) = a f(u)$$

$\Leftrightarrow$

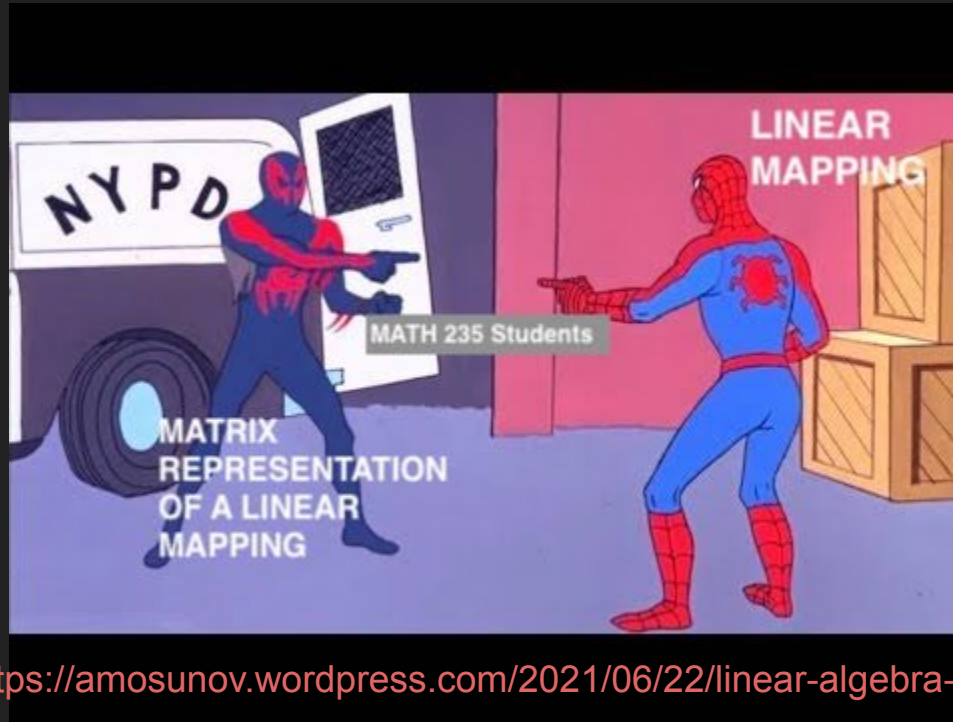
$$f(a u + b v) = a f(u) + b f(v)$$

does not matter if linear combination applied before or after transformation.

# linear map $\Leftrightarrow$ matrix representation



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<https://amosunov.wordpress.com/2021/06/22/linear-algebra-memes/>

# Dot Product as matrix product



$$a = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix}$$

$3 \times 1$

$$b = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$3 \times 1$

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$$a^T b = \begin{bmatrix} a_1 & a_2 & a_3 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \langle a, b \rangle$$

$1 \times 3$        $3 \times 1$        $1 \times 1$

$$b^T a = a^T b = \langle a, b \rangle$$



# Inner Product



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# Inner Product



$$\langle u+w, v \rangle = \langle u, v \rangle + \langle w, v \rangle$$



$$\left\langle \begin{bmatrix} u_1+w_1 \\ u_2+w_2 \end{bmatrix}, \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right\rangle =$$

$$\langle \alpha u, v \rangle = \alpha \langle u, v \rangle$$

$$\langle u, v+w \rangle = \langle u, v \rangle + \langle u, w \rangle$$

$$\langle u, \alpha v \rangle = \alpha \langle u, v \rangle$$

$$\langle u, v \rangle = f(u, v)$$

$$f(u, v) = f(u+w, v) = f(u, v) + f(w, v)$$

$$f(\alpha u, v) = \alpha f(u, v)$$

# Inner Product



$$f(x, y) = f\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = xy$$

$$f\left(\alpha \begin{bmatrix} x \\ y \end{bmatrix}\right) = f\left(\begin{bmatrix} \alpha x \\ \alpha y \end{bmatrix}\right) = (\alpha x)(\alpha y) = \alpha^2 xy$$



$$f\left(\begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} z \\ t \end{bmatrix}\right) = (x+z)(y+t) \neq xy + zt$$

$$f(y, v) = f\left(\begin{bmatrix} y \\ v \end{bmatrix}\right) = f\left(\alpha \begin{bmatrix} y \\ v \end{bmatrix}\right) = \alpha f\left(\begin{bmatrix} y \\ v \end{bmatrix}\right)$$

$$f\left(\begin{bmatrix} u+w \\ v \end{bmatrix}\right) = f\left(\begin{bmatrix} u \\ v \end{bmatrix}\right) + f\left(\begin{bmatrix} w \\ v \end{bmatrix}\right)$$

$f(x, y)$  is linear in  $x$  } bilinear  
is linear in  $y$  } ~~linear~~  
is not linear in  $(x, y)$

$$\langle u+w, v \rangle$$

# General vector spaces: Inner product space



$$\textcircled{I} \begin{cases} \langle \alpha u, v \rangle = \alpha \langle u, v \rangle \\ \langle u+v, w \rangle = \langle u, w \rangle + \langle v, w \rangle \end{cases}$$

$$\textcircled{II} \langle u, v \rangle = \langle v, u \rangle$$

$$\textcircled{III} \begin{cases} \langle u, u \rangle > 0 & u \neq 0 \\ \langle u, u \rangle = 0 & u = 0 \end{cases}$$

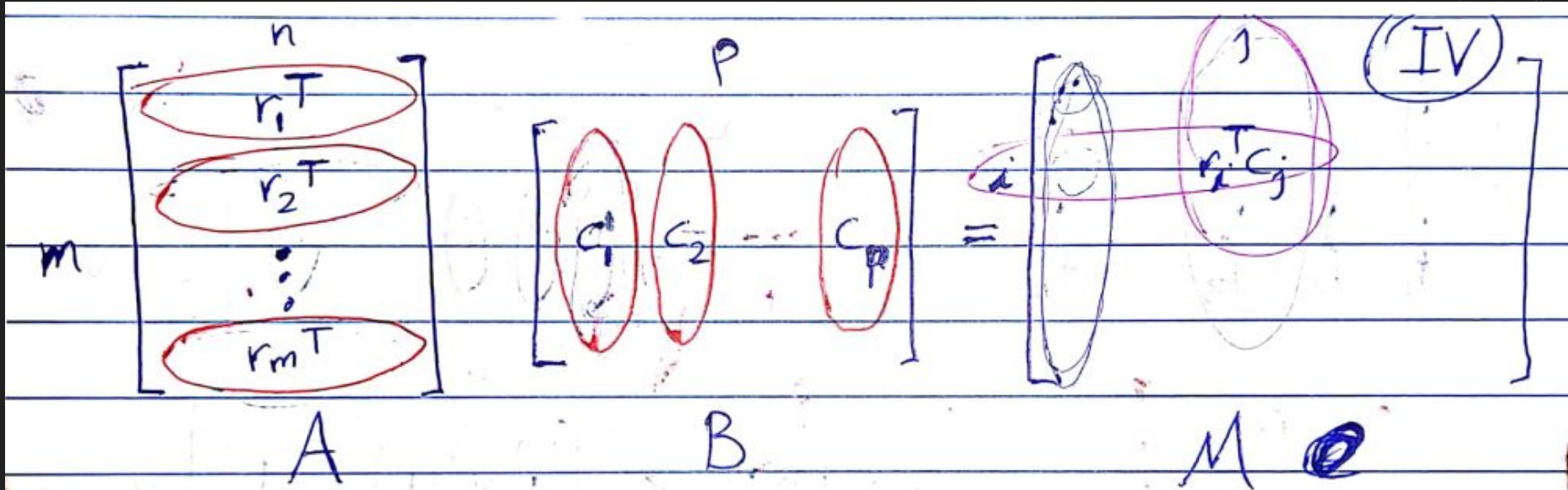
$$\langle 0, u \rangle = \langle \alpha v - v, u \rangle$$

$$f(u, v) : V \times V \rightarrow \mathbb{R}$$

# Matrix Multiplication in terms of inner products



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$$M_{ij} = \langle r_i, c_j \rangle = r_i^T c_j$$

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# Matrix multiplications in terms of matrix-vector product



$$F = \begin{bmatrix} f_1 & f_2 & f_3 & \dots & f_n \end{bmatrix}$$

$$W: P \times m \quad Y = \begin{bmatrix} y_1 & y_2 & y_3 & \dots & y_n \end{bmatrix} \quad y_i = W f_i$$

$$Y = W F = W \begin{bmatrix} f_1 & f_2 & \dots & f_n \end{bmatrix}$$

$$= \begin{bmatrix} W f_1 & W f_2 & \dots & W f_n \end{bmatrix}$$

$$= \begin{bmatrix} y_1 & y_2 & \dots & y_n \end{bmatrix} = Y$$

The diagram illustrates the matrix multiplication process. At the top, a matrix  $F$  is shown as a collection of column vectors  $f_1, f_2, f_3, \dots, f_n$ , each of size  $m \times 1$ . A weight matrix  $W$  of size  $P \times m$  is multiplied by  $F$  to produce the output matrix  $Y$  of size  $P \times n$ . The output  $Y$  is shown as a collection of column vectors  $y_1, y_2, y_3, \dots, y_n$ , each of size  $P \times 1$ . The relationship  $y_i = W f_i$  is highlighted in orange. A network diagram shows the connections between the input nodes (columns of  $F$ ) and the output nodes (columns of  $Y$ ), with weights  $w_{ij}$  connecting them. The final result is shown as  $Y = W F = W \begin{bmatrix} f_1 & f_2 & \dots & f_n \end{bmatrix} = \begin{bmatrix} W f_1 & W f_2 & \dots & W f_n \end{bmatrix} = \begin{bmatrix} y_1 & y_2 & \dots & y_n \end{bmatrix} = Y$ .

# Transforming a bunch of points data points as columns of a matrix



$$\begin{bmatrix} a_1^T \\ a_2^T \\ \vdots \\ a_m^T \end{bmatrix} \begin{bmatrix} | & | & \dots & | \\ b_1 & b_2 & \dots & b_n \\ | & | & \dots & | \end{bmatrix} = \begin{bmatrix} | & | & \dots & | \\ c_1 & c_2 & \dots & c_n \\ | & | & \dots & | \end{bmatrix} \quad \text{(VI)}$$

$A$   
 $m \times p$

$B$   
 $p \times n$

$C$   
 $m \times n$

$$c_i = A b_i \quad \text{😊}$$

$a_1^T$

# Transforming a bunch of points data points as rows of a matrix



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$$\begin{bmatrix} a_1^T \\ a_2^T \\ \vdots \\ a_m^T \end{bmatrix} \begin{bmatrix} b_1 & b_2 & \dots & b_n \end{bmatrix} = \begin{bmatrix} c_1^T \\ c_2^T \\ \vdots \\ c_m^T \end{bmatrix}$$

$A$   $m \times p$        $B$   $p \times n$        $C$   $m \times n$

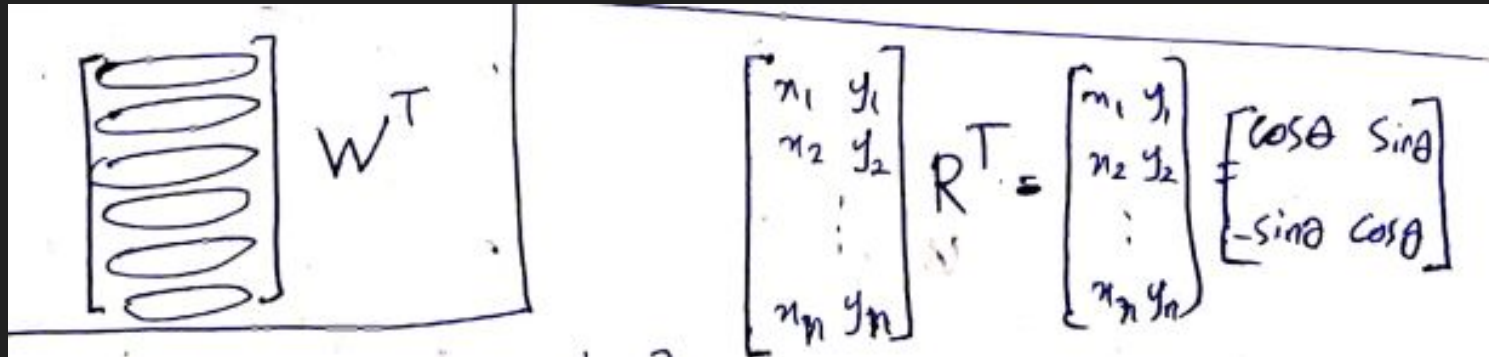
Or  $c_i^T = a_i^T B$  😊



be careful about linear transformations on row vectors!



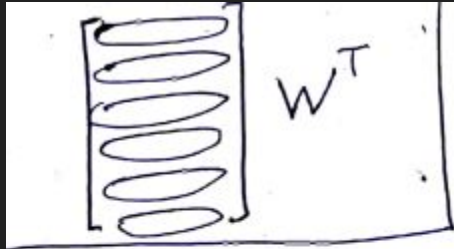
$$R \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$
$$R \left[ \begin{array}{c} \begin{bmatrix} x_1 \\ y_1 \end{bmatrix} \\ \begin{bmatrix} x_2 \\ y_2 \end{bmatrix} \\ \begin{bmatrix} x_3 \\ y_3 \end{bmatrix} \end{array} \right]$$



be careful about linear transformations on row vectors!



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$$W^T$$
$$\begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ \vdots & \vdots \\ x_n & y_n \end{bmatrix} R^T = \begin{bmatrix} x_1 & y_1 \\ x_2 & y_2 \\ \vdots & \vdots \\ x_n & y_n \end{bmatrix} \begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix}$$
$$R \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$
$$R \left[ \begin{matrix} x_1 \\ y_1 \end{matrix} \right] \left[ \begin{matrix} x_2 \\ y_2 \end{matrix} \right] \left[ \begin{matrix} x_3 \\ y_3 \end{matrix} \right] \dots$$

# Outer Product



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`np.outer(u,v)`

`u @ v.T`

How many independent columns?

How many independent rows?

`outer(u,v) = outer(v,u).T`

complex numbers?

# Outer Product



Outer product

$$u = \begin{bmatrix} a \\ b \\ c \end{bmatrix} \quad v = \begin{bmatrix} x \\ y \\ z \\ t \end{bmatrix}$$

$$u \in \mathbb{R}^3$$

$$v \in \mathbb{R}^4$$

(VII)

$$W = \begin{matrix} & \begin{matrix} x & y & z & t \end{matrix} \\ \begin{matrix} a \\ b \\ c \end{matrix} & \begin{bmatrix} ax & ay & az & at \\ bx & by & bz & bt \\ cx & cy & cz & ct \end{bmatrix} \end{matrix} = \begin{bmatrix} a(x,y,z,t) \\ b(x,y,z,t) \\ c(x,y,z,t) \end{bmatrix}$$

$$W \in \mathbb{R}^{3 \times 4}$$

$$W = u \otimes v$$

$$\mathbb{R}^{3 \times 4}$$

# Outer Product



$$W = u \otimes v$$

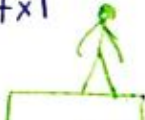
$v \otimes u = W^T$

$$W = \begin{bmatrix} x \begin{bmatrix} a \\ b \\ c \end{bmatrix} & y \begin{bmatrix} a \\ b \\ c \end{bmatrix} & z \begin{bmatrix} a \\ b \\ c \end{bmatrix} & t \begin{bmatrix} a \\ b \\ c \end{bmatrix} \end{bmatrix} = \begin{bmatrix} xu & yu & zu & tu \end{bmatrix}$$

$$W = \begin{bmatrix} a \\ b \\ c \end{bmatrix} \begin{bmatrix} x & y & z & t \end{bmatrix} = \begin{matrix} 3 \times 1 \\ 1 \times 4 \end{matrix} u v^T = W$$

$u \in \mathbb{R}^{3 \times 1}$     $v \in \mathbb{R}^{4 \times 1}$

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# Inner product vs outer product



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inner product  $u^T v \in \mathbb{R}$

$$u, v \in \mathbb{R}^{n \times 1}$$

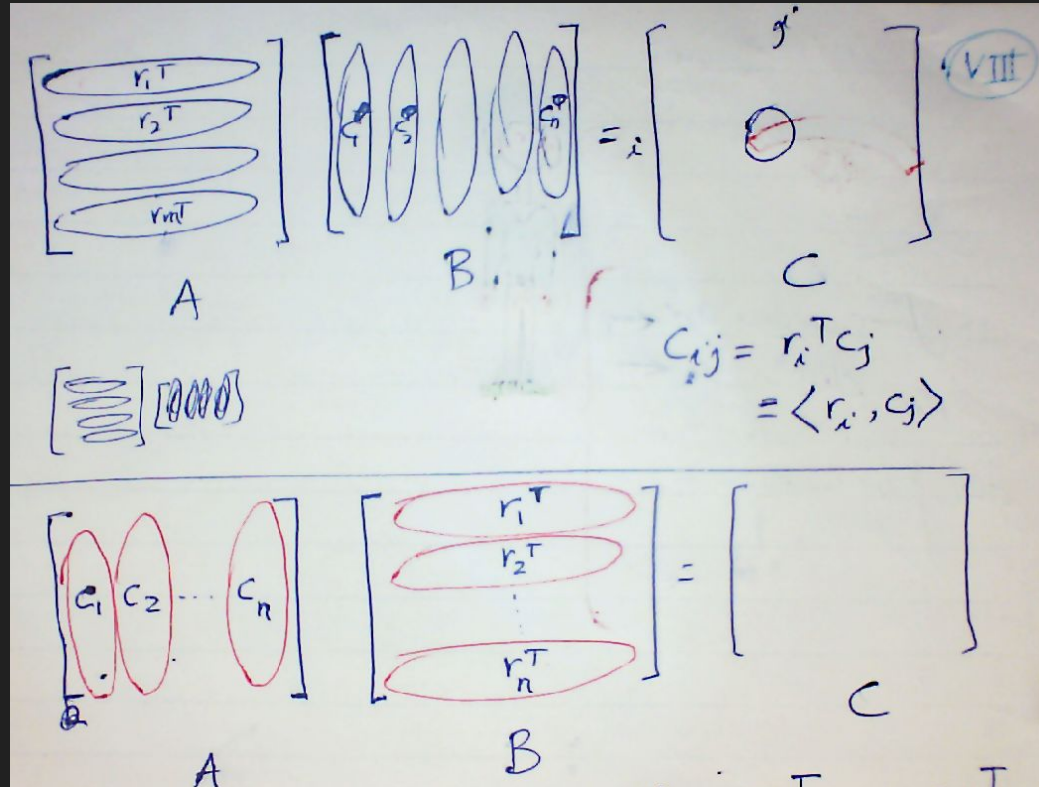
$$u \cdot v = \langle u, v \rangle = u^T v$$

outer product  $u v^T \in \mathbb{R}^{m \times n}$

$$u \in \mathbb{R}^{m \times 1} \quad v \in \mathbb{R}^{n \times 1}$$

$$u \otimes v = u v^T$$

# Two ways of looking at matrix product



# Matrix Multiplication in terms of outer products



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$$\begin{bmatrix} a & b \\ d & e \end{bmatrix} \cdot \begin{bmatrix} x & y \\ z & t \end{bmatrix} = \begin{bmatrix} ax+by & ay+bt \\ dx+ez & dy+et \end{bmatrix}$$
$$\begin{bmatrix} ax & ay \\ dx & dy \end{bmatrix} + \begin{bmatrix} bz & bt \\ ez & et \end{bmatrix} = \begin{bmatrix} a \\ d \end{bmatrix} \begin{bmatrix} x & y \end{bmatrix} + \begin{bmatrix} b \\ e \end{bmatrix} \begin{bmatrix} z & t \end{bmatrix}$$
$$\begin{bmatrix} a \\ d \end{bmatrix} \otimes \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} b \\ e \end{bmatrix} \otimes \begin{bmatrix} z \\ t \end{bmatrix}$$



# Matrix Multiplication in terms of outer products



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$$A = [c_1 \ c_2 \ \dots \ c_n]$$
$$B = \begin{bmatrix} r_1^T \\ r_2^T \\ \vdots \\ r_n^T \end{bmatrix} = C$$

$$C = c_1 r_1^T + \dots + c_n r_n^T$$
$$C = \sum_{i=1}^n c_i r_i^T$$
$$= \sum_{i=1}^n c_i \otimes r_i$$

# Block-wise multiplication



$$\begin{array}{l} m_1 + m_2 = m \\ A \rightarrow m \times n \quad B \rightarrow n \times p \\ \left[ \begin{array}{c} m_1 \\ m_2 \end{array} \right] \left[ \begin{array}{c} A_1 \\ A_2 \end{array} \right] \left[ \begin{array}{cc} B_1 & B_2 \end{array} \right] = C \quad n \times p \\ \underbrace{\hspace{10em}}_{p_1 + p_2 = p} \end{array}$$
$$\left[ \begin{array}{cc} A_1 B_1 & A_1 B_2 \\ A_2 B_1 & A_2 B_2 \end{array} \right] = \left[ \begin{array}{cc} m_1 \times p_1 & m_1 \times p_2 \\ m_2 \times p_1 & m_2 \times p_2 \end{array} \right]$$

$\downarrow$   $m_2 \times n$   $n \times p_1$   
 $m_2 \times p_1$

# Block-wise multiplication



$$\begin{array}{c} m_1 \times n_1 \\ \swarrow \\ A \\ \left[ \begin{array}{cc} A_1 & A_2 \\ A_3 & A_4 \end{array} \right] \begin{array}{l} n_1 \\ n_2 \end{array} \left[ \begin{array}{cc} B_1 & B_2 \\ B_3 & B_4 \end{array} \right] \\ \underbrace{\hspace{1.5cm}}_{n_1} \quad \underbrace{\hspace{1.5cm}}_{n_2} \end{array}$$
  
$$\left[ \begin{array}{cc} A_1 B_1 + A_2 B_3 & A_1 B_2 + A_2 B_4 \\ A_3 B_1 + A_4 B_3 & A_3 B_2 + A_4 B_4 \end{array} \right]$$

# Column Rank

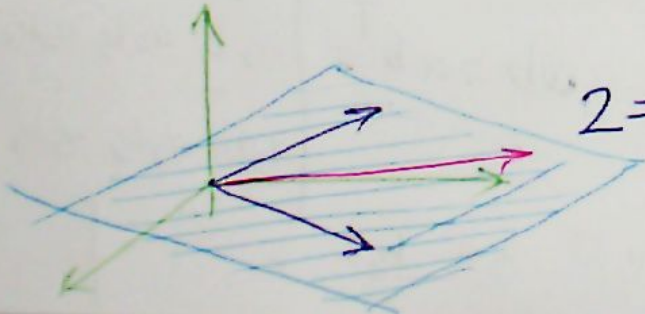


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Column Rank

$$\begin{bmatrix} 1 & 4 & 3 \\ 2 & 5 & 3 \\ 3 & 6 & 3 \end{bmatrix}$$

Column Rank = 2



$2 = \dim(\text{column space})$

# Column Rank and Row Rank



$$\begin{bmatrix} 1 & -2 \\ 2 & -4 \\ 3 & -6 \\ 4 & -8 \end{bmatrix}$$

$$\text{column rank} = 1$$

$$\text{row rank} = 1$$

$$\begin{bmatrix} 1 & 0 & 4 \\ 2 & 0 & 5 \\ 3 & 0 & 6 \end{bmatrix}$$

$$\text{column rank} = 2$$

$$\text{row rank} = 2$$

# Column Rank and Row Rank



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$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

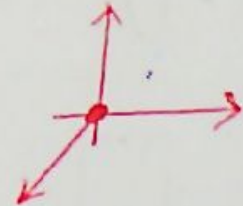
column rank = 3

row rank = 3

$$\begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

column rank = 0

row rank = 0



# Column Rank and Row Rank



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$$\vec{a} = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \end{bmatrix} \quad \vec{b} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} \quad \vec{a} \otimes \vec{b} = \vec{a} \vec{b}^T = \begin{bmatrix} a_1 b_1 & a_1 b_2 & a_1 b_3 & a_1 b_4 \\ a_2 b_1 & a_2 b_2 & a_2 b_3 & a_2 b_4 \\ a_3 b_1 & a_3 b_2 & a_3 b_3 & a_3 b_4 \end{bmatrix}$$
  
$$\vec{a} \otimes \vec{b} = \vec{a} \vec{b}^T = \begin{bmatrix} b_1 \vec{a} & b_2 \vec{a} & b_3 \vec{a} & b_4 \vec{a} \end{bmatrix} = \begin{bmatrix} a_1 b^T \\ a_2 b^T \\ a_3 b^T \end{bmatrix}$$

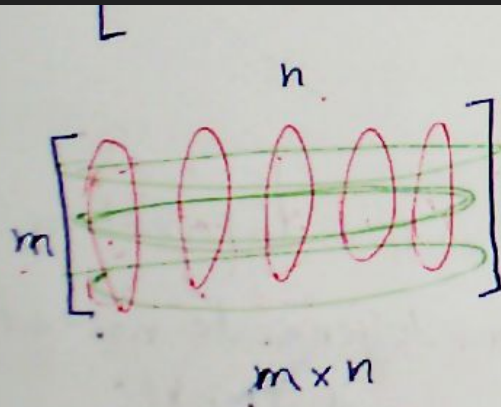
(at most) 1 independent column      1 independent row

$$\text{RowRank}(\vec{a} \vec{b}^T) = \text{ColumnRank}(\vec{a} \vec{b}^T) \leq 1$$

# Column Rank and Row Rank



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$$\text{column rank} = \dim(\text{column space}) \leq n$$

$$\text{row rank} = \dim(\text{row space}) \leq m$$

$$\text{column rank} \leq \min(m, n)$$

$$\text{row rank} \leq \min(m, n)$$



# Column Rank and Row Rank



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$$\begin{bmatrix} 1 & 2 & 3 & 0 \\ 2 & 3 & 5 & 1 \\ 3 & 4 & 7 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 2 & 3 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 1 & -1 \end{bmatrix}$$

Matrix Factorization

$L = [ ]$   $3 \times 2$   $2 \times 4$

$\text{Row Rank} = \text{Col Rank}$

$\begin{bmatrix} c_1 & c_2 & \dots & c_n \end{bmatrix}$

for  $i = 1 \dots n$   
if  $c_i \notin \text{span}(L)$   
L.append( $c_i$ )



# \* Column Rank and Row Rank

$$\begin{bmatrix} 1 & 2 & 3 & 10 \\ 2 & 3 & 5 & 1 \\ 3 & 4 & 7 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 2 & 3 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 1 & -1 \end{bmatrix}$$

Matrix Factorization

$L = [ ]$   $3 \times 2$   $2 \times 4$

$\text{Row Rank} = \text{Col Rank}$

for  $i = 1 \dots n$   
if  $c_i \notin \text{span}(c_1, \dots, c_{i-1})$   
L.append( $c_i$ )

$\begin{bmatrix} c_1 & c_2 & \dots & c_n \end{bmatrix}$

Column Rank = Row Rank

# Column Rank and Row Rank



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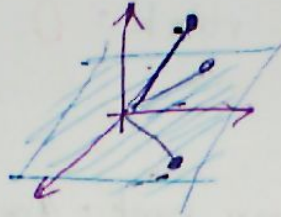
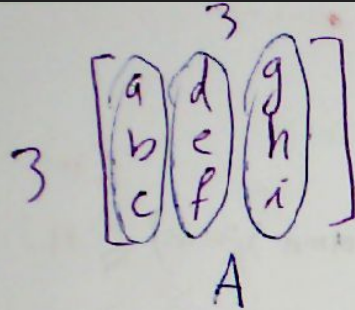
$$m \begin{bmatrix} & n \\ & A \\ & \end{bmatrix} = m \begin{bmatrix} & r \\ & \text{columns} \\ & \end{bmatrix} r \begin{bmatrix} & n \\ & \text{rows} \\ & \end{bmatrix}$$

column rank =  $r \leq \min(m, n)$       a basis for column space      a basis for row space

**Row Rank(A) = Column Rank(A)**  
**def = Rank(A)** رتبه

$\text{Rank}(A) = \dim(\text{Colspace}(A)) = \dim(\text{Row Space}(A))$

# "Most" matrices have full rank



if  $A$  is a random  $3 \times 3$  matrix (uniform distribution)  
nondegenerate normal distribution

$\text{rank}(A) = 3$  with probability = 1

almost Always

# thin and fat matrices



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$$A = \begin{bmatrix} \phantom{0} \\ \phantom{0} \\ \phantom{0} \end{bmatrix} \in \mathbb{R}^{m \times n} \quad m < n$$

Fat matrix

چاق

$$A = \begin{bmatrix} \phantom{0} \\ \phantom{0} \\ \phantom{0} \end{bmatrix} \in \mathbb{R}^{m \times n} \quad m > n$$

thin matrix لاغیر

$$A = \begin{bmatrix} \phantom{0} \\ \phantom{0} \\ \phantom{0} \end{bmatrix} \in \mathbb{R}^{m \times n} \quad m = n$$

مربعی Square matrix

# full-rank and rank-deficient



$$A \in \mathbb{R}^{m \times n}$$

$$\text{rank}(A) \leq \min(m, n)$$

①9

$$\left\{ \begin{array}{l} \text{rank}(A) = \min(m, n) \\ \text{rank}(A) < \min(m, n) \end{array} \right.$$

full-rank

rank-deficient

رتبه کم

A is full-rank

A has full rank

# full-rank and rank-deficient



$$A \in \mathbb{R}^{m \times n}$$

$$\text{rank}(A) \leq \min(m, n)$$

①9

$$\left\{ \begin{array}{l} \text{rank}(A) = \min(m, n) \\ \text{rank}(A) < \min(m, n) \end{array} \right.$$

full-rank

rank-deficient  
رتبه ناقص

A is full-rank

A has full rank

$$A \in \mathbb{R}^{n \times n}$$

$$\text{rank}(A) = n \quad \text{full-rank}$$

$$\text{rank}(A) < n \quad \text{rank-deficient}$$