# Chapter 1

#### Matrices, Vectors, and Systems of Linear Equations

除了標註※之簡報外,其餘採用李宏毅教授之投影片教材

# Vectors and Matrices (Chapter 1.1)

## What is a vector?

• Physics student: Vectors have lengths and directions.



- Math student: Vectors satisfy a set of rules: u + v, 3u are vectors, u + v = v + u, (u + v) + w = u + (v + w), ...
- EE/CS student: A vector **v** is a sequence of numbers  $\mathbf{v} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$

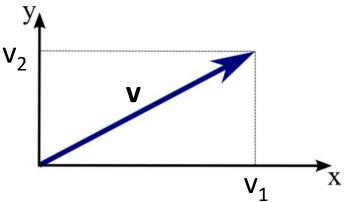


# Vectors (from EE/CS viewpoint)

- A vector **v** is a sequence of numbers
- Components: the entries of a vector.
  - The i-th component of vector **v** refers to v<sub>i</sub>

• If a vector only has less than four components, you can visualize it.

NOTE: Later we will see a more general definition of a "vector"

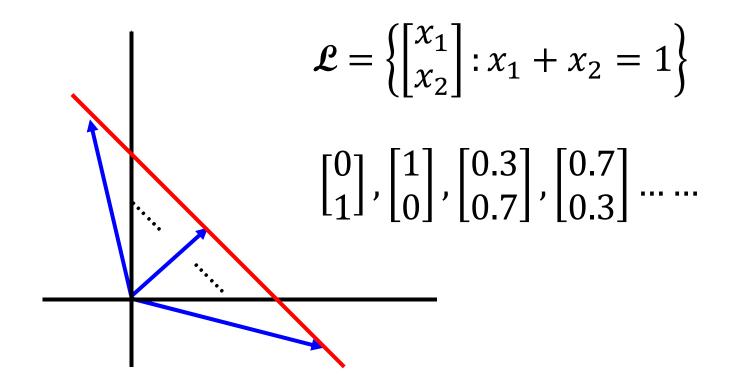


2

**v** =

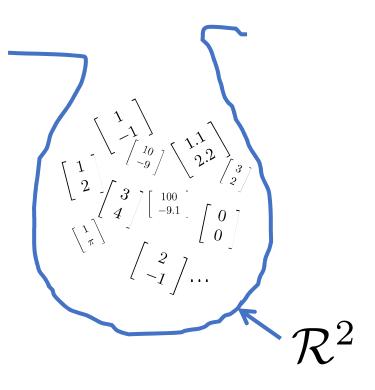
Vector Set 
$$\begin{cases} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}, \begin{bmatrix} 6 \\ 8 \\ 9 \end{bmatrix}, \begin{bmatrix} 9 \\ 0 \\ 2 \end{bmatrix} \end{cases}$$
 A vector set with 4 elements

• A vector set can contain infinite elements

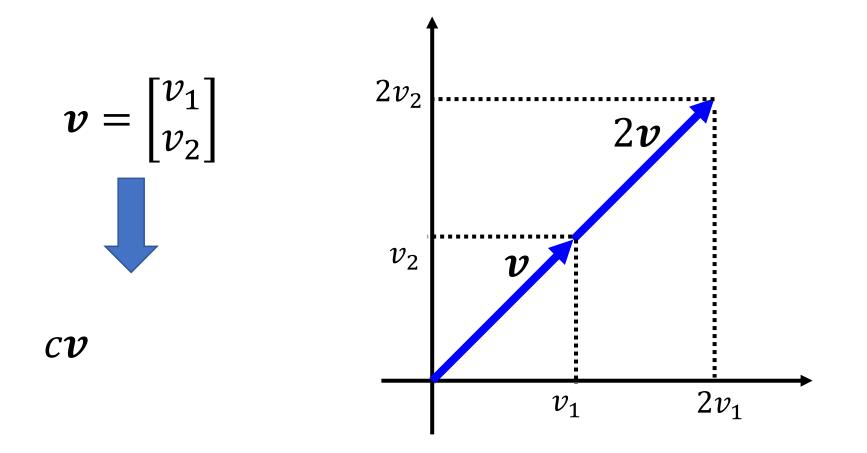


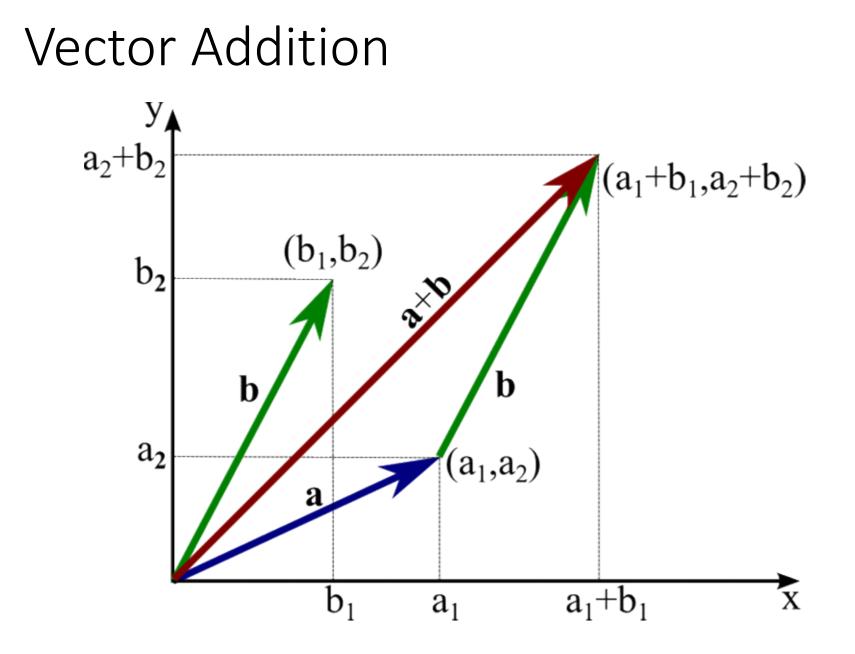
### Vector Set

•  $\Re^n$ : We denote the set of all vectors with *n* entries by  $\Re^n$ 



## Scalar Multiplication





# **Properties of Vector**

**Objects having the** following 8 properties are "vectors".

= - u

For any vectors **u**, **v** and **w** in  $\mathcal{R}^n$ , and any scalars a and b

• u + v = v + u

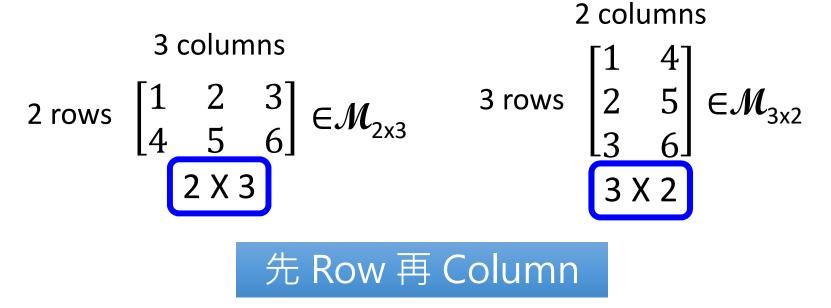
• 
$$(u + v) + w = u + (v + w)$$

- There is an element **0** in  $\mathcal{R}^n$  such that **0** + **u** = **u**
- There is an element  $\mathbf{u}'$  in  $\mathcal{R}^n$  such that  $\mathbf{u}' + \mathbf{u} = \mathbf{0}$
- 1**u** = **u**
- (ab)u = a(bu)
- $\mathbf{0} = \begin{bmatrix} \mathbf{0} \\ \vdots \\ \mathbf{0} \end{bmatrix}$  zero vector • a(**u**+**v**) = a**u** + a**v**
- (a+b)**u** = a**u** + b**u**

In Chapter 7, the above will be generalized to "vector space"

Matrix 
$$A = \begin{bmatrix} 2 & 3 & 5 \\ 3 & 1 & -1 \\ -2 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \end{bmatrix}$$

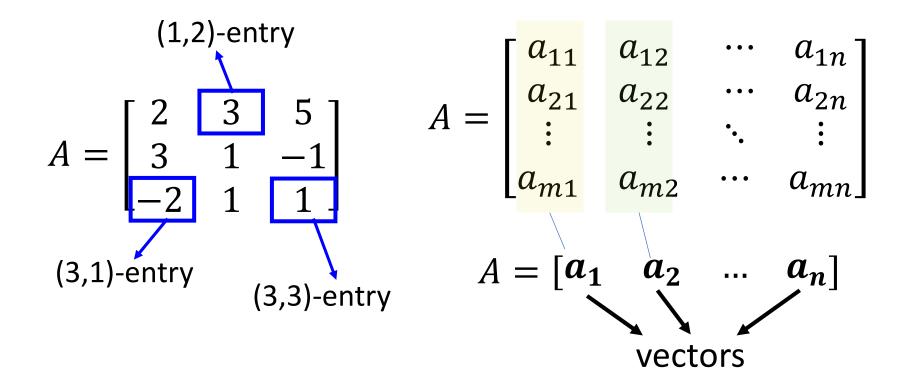
- If the matrix has m rows and n columns, we say the size of the matrix is m by n, written m x n
- We use  $\mathcal{M}_{\rm mxn}$  to denote the set that contains all matrices of size m x n



#### Matrix

#### 先 Row 再 Column

• *Index of component*: the scalar in the i-th row and j-th column is called (i,j)-entry of the matrix



#### Matrix

- Two matrices with the same size can add or subtract.
- Matrix can be multiplied by a scalar

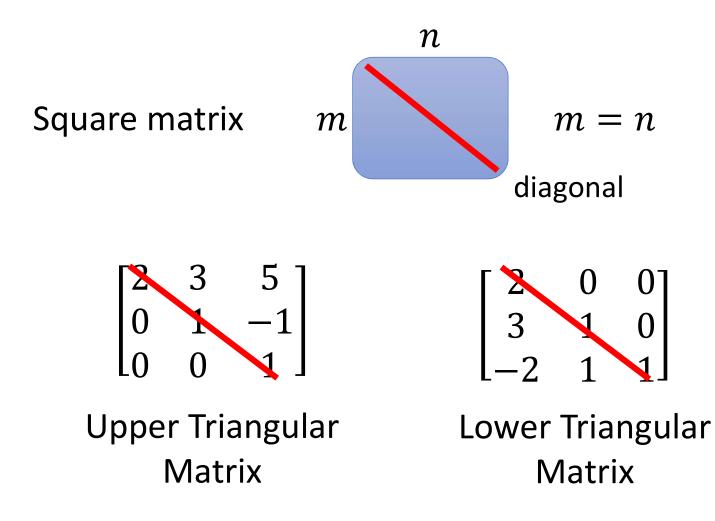
$$A = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix} \qquad B = \begin{bmatrix} 6 & 9 \\ 8 & 0 \\ 9 & 2 \end{bmatrix} \qquad 9B$$

A + B A - B

#### Properties

- A, B, C are mxn matrices, and s and t are scalars
  - A + B = B + A
  - (A + B) + C = A + (B + C)
  - (st)A = s(tA)
  - s(A + B) = sA + sB
  - (s+t)A = sA + tA

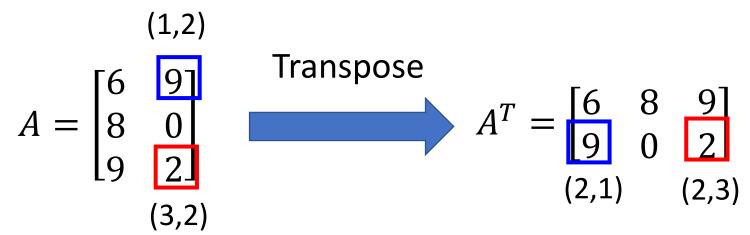
# 有名有姓的 Matrix



Is I3 a diagonal matrix?有名有姓的 MatrixYES• Diagonal Matrix
$$0 \quad 0 \quad 0 \\ 0 \quad 0 \quad 0 \quad 0 \quad 0 \quad 0 \quad 0$$
• Diagonal Matrix $0 \quad 0 \quad 0$ • Identity Matrix $1 \quad 0 \quad 0$ I a =  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ Denoted by I (any size) or  $I_n$ • Zero Matrix $O_{2\times3} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ Denoted by O (any size) or  $O_{m \times n}$ 

### Transpose

• If A is an mxn matrix,  $A^T$  (transpose of A) is an nxm matrix whose (i,j)-entry is the (j-i)-entry of A



Column 變成 Row; Row 變成 Column

Why do we care about the transpose of a matrix? Will explain later!

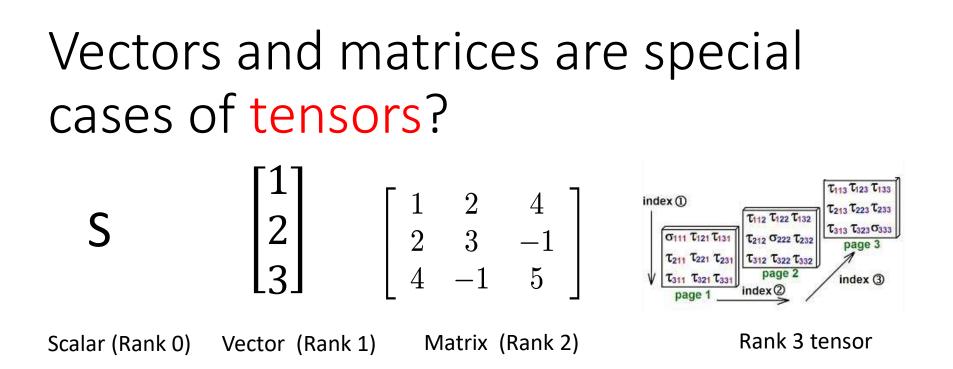
#### Iranspose

- A and B are mxn matrices, and s is a scalar
- $(A^T)^T = A$

•  $(sA)^T = sA^T$ •  $(A+B)^T = A^T + B^T$  This is a linear system  $\textcircled{\odot}$ 

Symmetric Matrix  $A^T = A$ 

$$A = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 3 & -1 \\ 4 & -1 & 5 \end{bmatrix} = A^T \quad B = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \neq B^T$$



The above is an over-simplified definition of a tensor. So what is a tensor?

 Informal Definition: An object that is invariant under a change of coordinates, and has components that change in a special, predictable way under a change of coordinates.



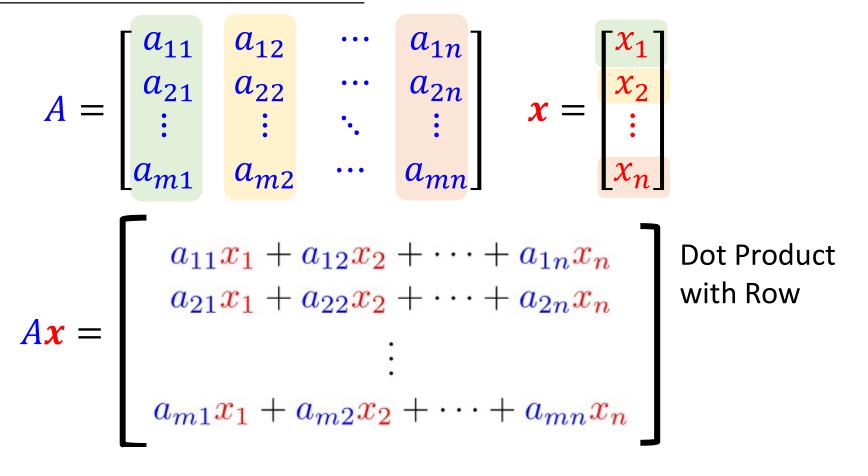
# Matrix-Vector Product (Chapter 1.2)

#### **Matrix-Vector Product**

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}_{n \times 1}$$
$$A\mathbf{x} = \begin{bmatrix} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \end{bmatrix} Dot Product with Row$$
$$\mathbf{x} = \begin{bmatrix} a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_1 \\ x_2 \\ x_2 \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_2 \\ x_2 \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_1 \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_2 \end{bmatrix} \mathbf{x} = \begin{bmatrix} x_1 \\ x_$$

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \qquad \mathbf{x} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \qquad \mathbf{A}\mathbf{x} = \begin{bmatrix} \\ \\ \end{bmatrix}$$

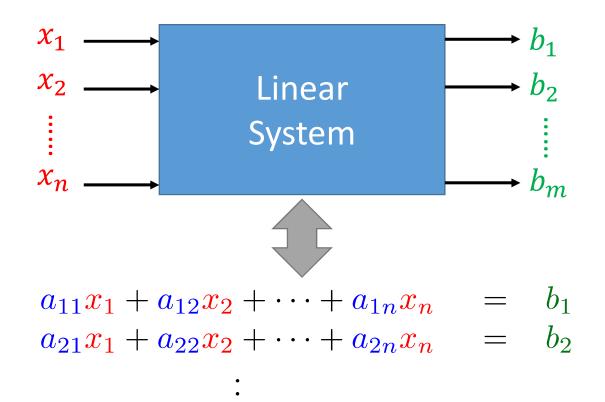
#### **Matrix-Vector Product**

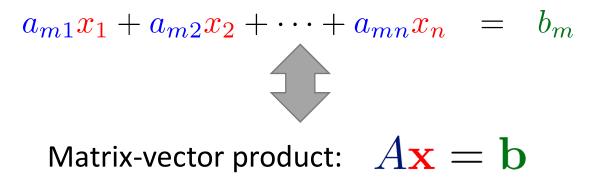


#### Weighted sum of Columns

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \vdots \\ \mathbf{x}_n \end{bmatrix}$$

$$\begin{array}{c} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \mathbf{x}_2 \\ \mathbf{x}_2 \\ \mathbf{x}_1 \\ \mathbf{x}_2 \\ \mathbf{x}_2 \\ \mathbf{x}_1 \\ \mathbf{x}_1 \\ \mathbf{x}_2 \\ \mathbf{x}_1 \\$$





# Properties of Matrix-Vector Product (Chapter 1.2)

#### Matrix-vector Product

• The sizes of matrix and vector should match.

$$A = \begin{bmatrix} 2 & 3 & 5 \\ 3 & 1 & -1 \\ -2 & 1 & 1 \end{bmatrix} \xrightarrow{} x = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$
$$A' = \begin{bmatrix} 1 & -1 \\ 2 & 3 \\ 1 & 4 \end{bmatrix} \qquad A'' = \begin{bmatrix} 2 & 1 \\ 3 & 2 \\ 0 & -1 \\ 1 & -3 \end{bmatrix}$$

# Properties of Matrix-vector Product

- A and B are mxn matrices, **u** and **v** are vectors in  $\Re^n$ , and c is a scalar.
- $A(\boldsymbol{u} + \boldsymbol{v}) = A\boldsymbol{u} + A\boldsymbol{v}$
- $A(c\boldsymbol{u}) = c(A\boldsymbol{u}) = (cA)\boldsymbol{u}$
- (A+B)u = Au + Bu
- A0 is the mx1 zero vector
- Ov is also the mx1 zero vector

• 
$$I_n \boldsymbol{v} = \boldsymbol{v}$$

 $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}$ 

$$=\begin{bmatrix} v_1\\v_2\\v_3\end{bmatrix}$$

## Properties of Matrix-vector Product

• A and B are mxn matrices. If Aw = Bw for all w in  $\mathcal{R}^n$ . Is it true that A = B?

 $Ae_{i} = a_{i}$ , where  $e_{i}$  is the j-th standard vector in  $\mathcal{R}^{n}$ 

$$e_{1} = \begin{bmatrix} 1 \\ \vdots \\ 0 \end{bmatrix} \quad Ae_{1} = \begin{bmatrix} a_{1} & \cdots & a_{n} \end{bmatrix} \begin{bmatrix} 1 \\ \vdots \\ 0 \end{bmatrix} = 1 \cdot a_{1} + 0 \cdot a_{2} + \cdots + 0 \cdot a_{n}$$
$$= a_{1}$$
$$= a_{1}$$
$$= a_{1}$$
$$= a_{1}$$
$$= b_{1} \quad a_{2} = b_{2} \quad a_{n} = b_{n}$$

Linear Combination (Chapter 1.2)

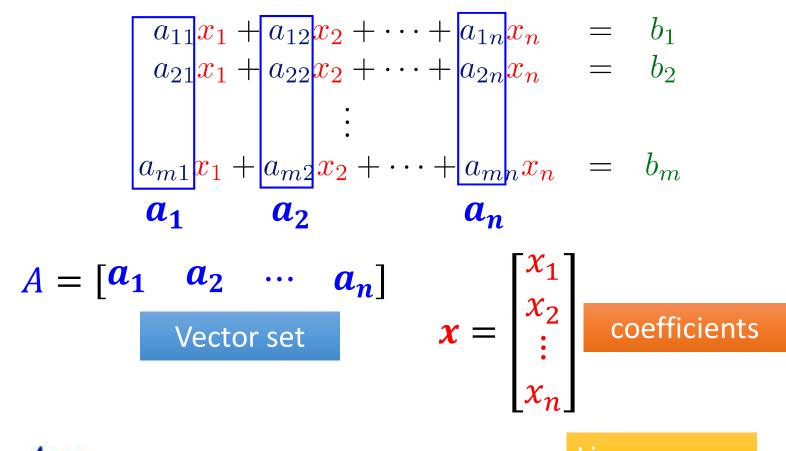
## Linear Combination

- Given a vector set  $\{u_1, u_2, \cdots, u_k\}$
- The linear combination of the vectors in the set
  - $\boldsymbol{v} = c_1 \boldsymbol{u_1} + c_2 \boldsymbol{u_2} + \dots + c_k \boldsymbol{u_k}$
  - $c_1, c_2, \dots, c_k$  are scalars (coefficients of linear combination)

vector set:  $\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$   $-3 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 4 \begin{bmatrix} 1 \\ 3 \end{bmatrix} + \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ coefficients:  $\{-3, 4, 1\}$   $= \begin{bmatrix} 2 \\ 8 \end{bmatrix}$ 

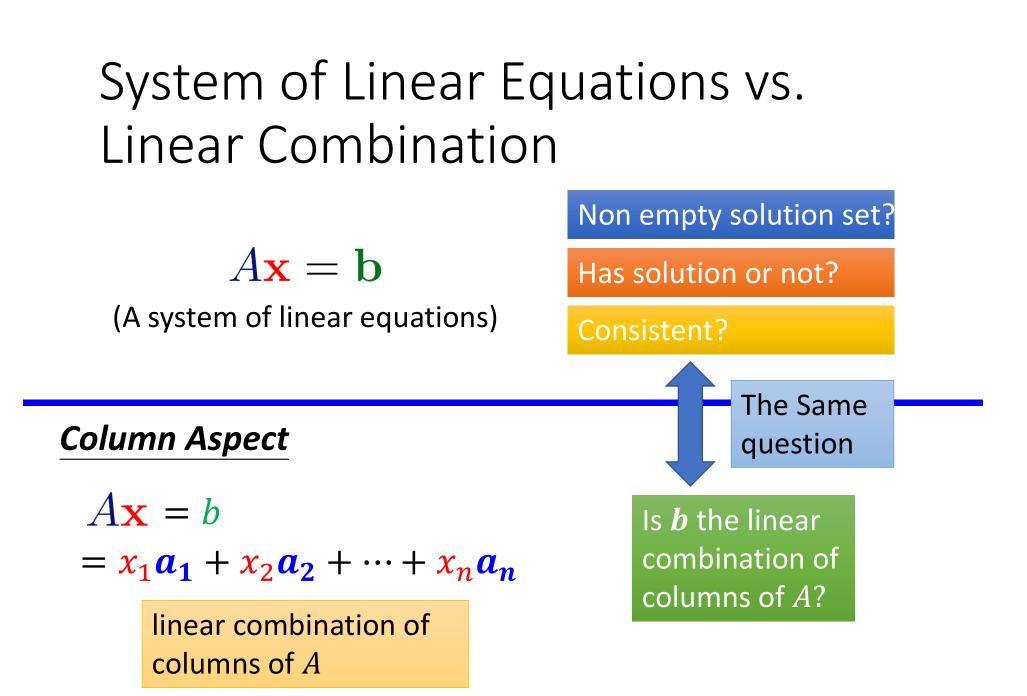
其實就是 weighted sum 啦 ☺

### Column Aspect



 $A\mathbf{x} = x_1 \mathbf{a_1} + x_2 \mathbf{a_2} + \dots + x_n \mathbf{a_n}$  Linear

Linear Combination



$$3x_1 + 6x_2 = 3$$
$$2x_1 + 4x_2 = 4$$

Has solution or not?

$$A = \begin{bmatrix} 3 & 6 \\ 2 & 4 \end{bmatrix}$$
$$A = \begin{bmatrix} 3 & 6 \\ 2 & 4 \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad b = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$$

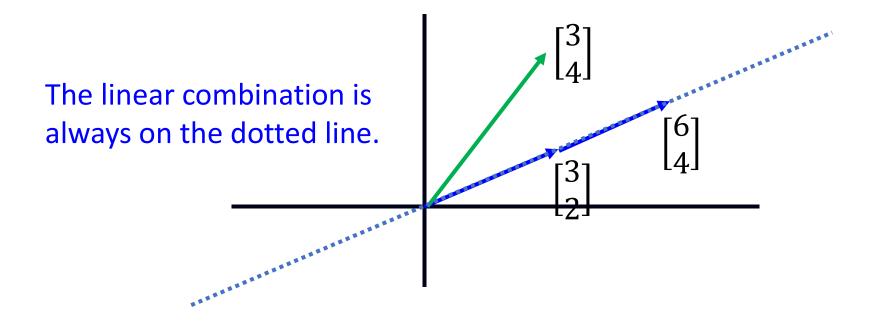
Is *b* the linear combination of columns of *A*?  $\begin{bmatrix}3\\4\end{bmatrix} \qquad \qquad \begin{bmatrix}3\\2\end{bmatrix}, \begin{bmatrix}6\\4\end{bmatrix}$ 

# Example 1

 $3x_1 + 6x_2 = 3$  $2x_1 + 4x_2 = 4$ 

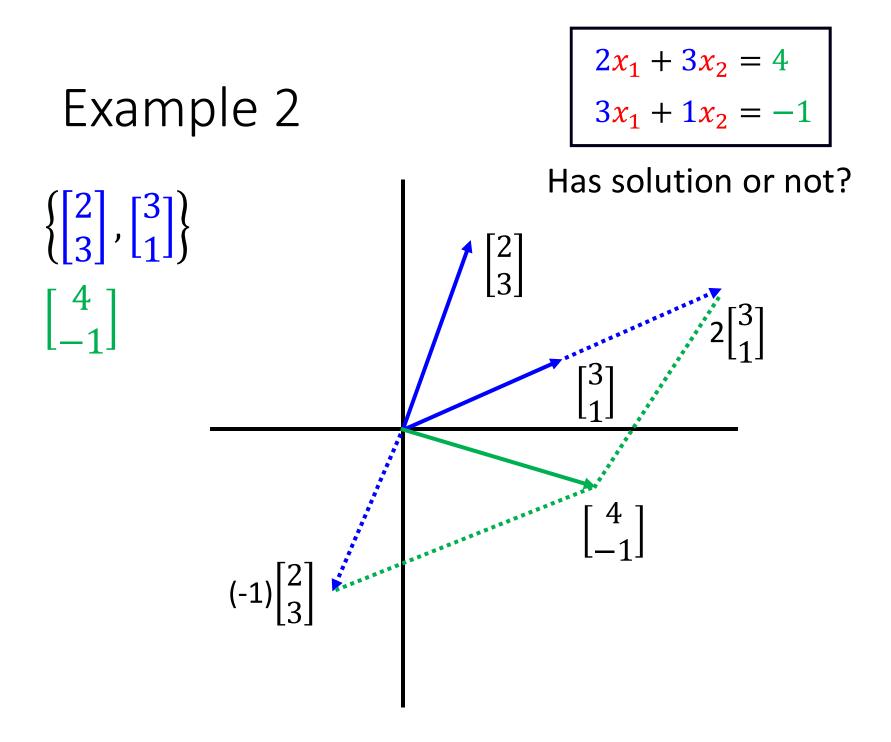
Has solution or not?

- Vector set:  $\left\{ \begin{bmatrix} 3 \\ 2 \end{bmatrix}, \begin{bmatrix} 6 \\ 4 \end{bmatrix} \right\}$
- Is  $\begin{bmatrix} 3 \\ 4 \end{bmatrix}$  a linear combination of  $\left\{ \begin{bmatrix} 3 \\ 2 \end{bmatrix}, \begin{bmatrix} 6 \\ 4 \end{bmatrix} \right\}$ ? **NO**



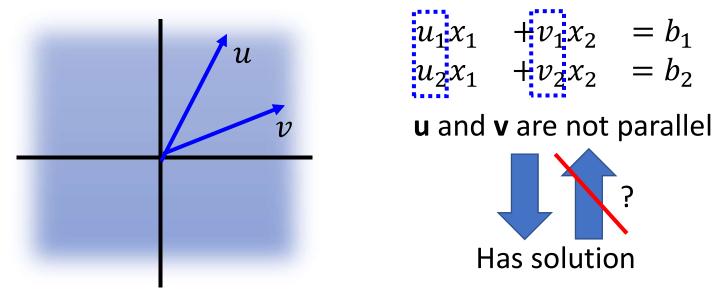
$$\begin{array}{c}
2x_1 + 3x_2 = 4 \\
3x_1 + 1x_2 = -1 \\
\text{Has solution or not?}
\end{array}
\begin{array}{c}
A\mathbf{x} = \mathbf{b} \\
A = \begin{bmatrix} 2 & 3 \\ 3 & 1 \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad b = \begin{bmatrix} 4 \\ -1 \end{bmatrix}
\end{array}$$

Is *b* a linear combination of columns of *A*?  $\begin{bmatrix} 4 \\ -1 \end{bmatrix} \qquad \qquad \begin{bmatrix} 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 3 \\ 1 \end{bmatrix}$ 



## Example 2

- If **u** and **v** are any nonparallel vectors in  $\mathscr{R}^2$ , then every vector in  $\mathscr{R}^2$  is a linear combination of **u** and **v** 
  - Nonparallel: **u** and **v** are nonzero vectors, and  $\mathbf{u} \neq c\mathbf{v}$ .

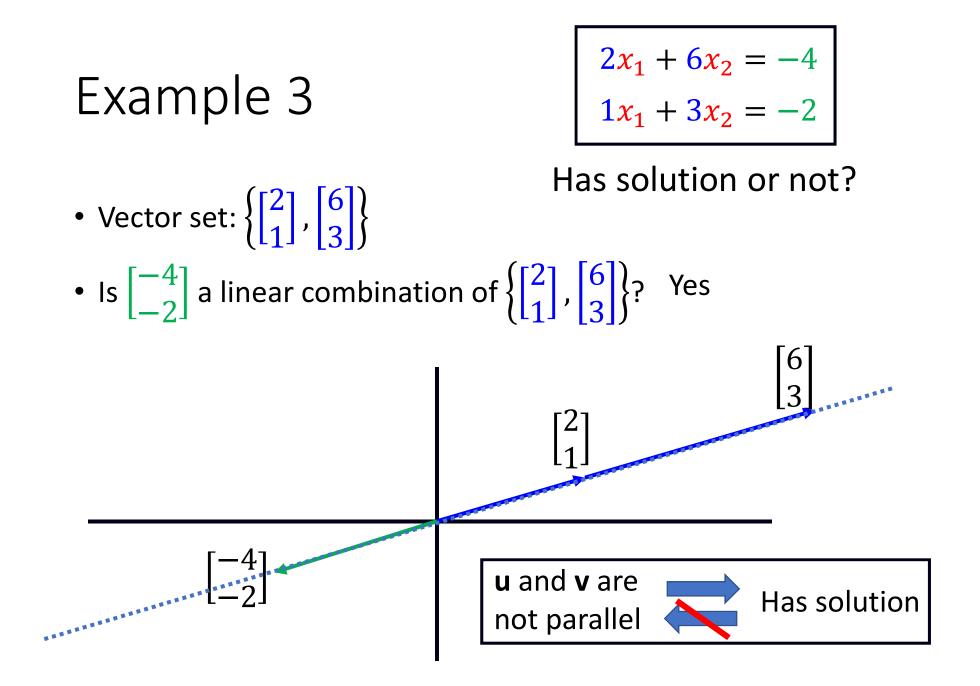


• If **u**, **v** and **w** are any nonparallel vectors in  $\mathcal{R}^3$ , then every vector in  $\mathcal{R}^3$  is a linear combination of **u**, **v** and **w**? **NO** 

$$\begin{array}{c}
2x_1 + 6x_2 = -4 \\
1x_1 + 3x_2 = -2
\end{array}$$

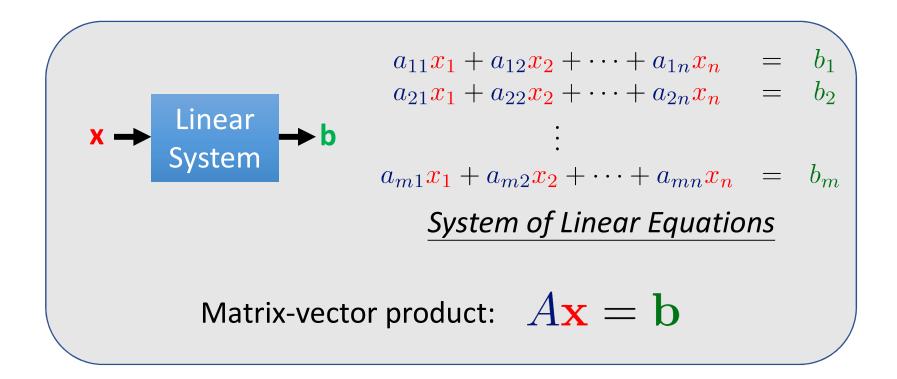
$$\begin{array}{c}
A\mathbf{x} = \mathbf{b} \\
A = \begin{bmatrix} 2 & 6 \\ 1 & 3 \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad b = \begin{bmatrix} -4 \\ -2 \end{bmatrix}$$
Has solution or not?

Is *b* the linear combination of columns of *A*?  $\begin{bmatrix} -4 \\ -2 \end{bmatrix} \qquad \left\{ \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 6 \\ 3 \end{bmatrix} \right\}$ 



# Having Solutions or Not (Chapter 1.3)

#### Review

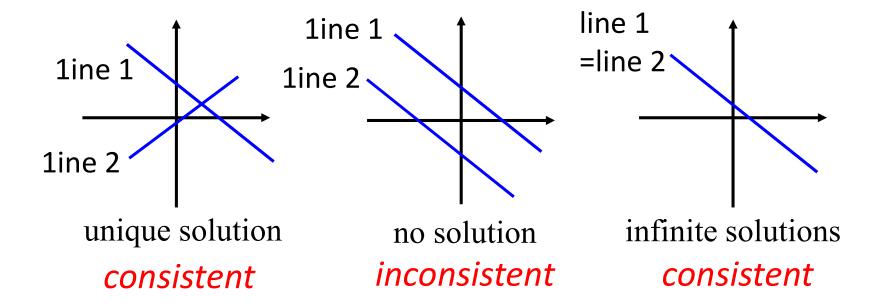


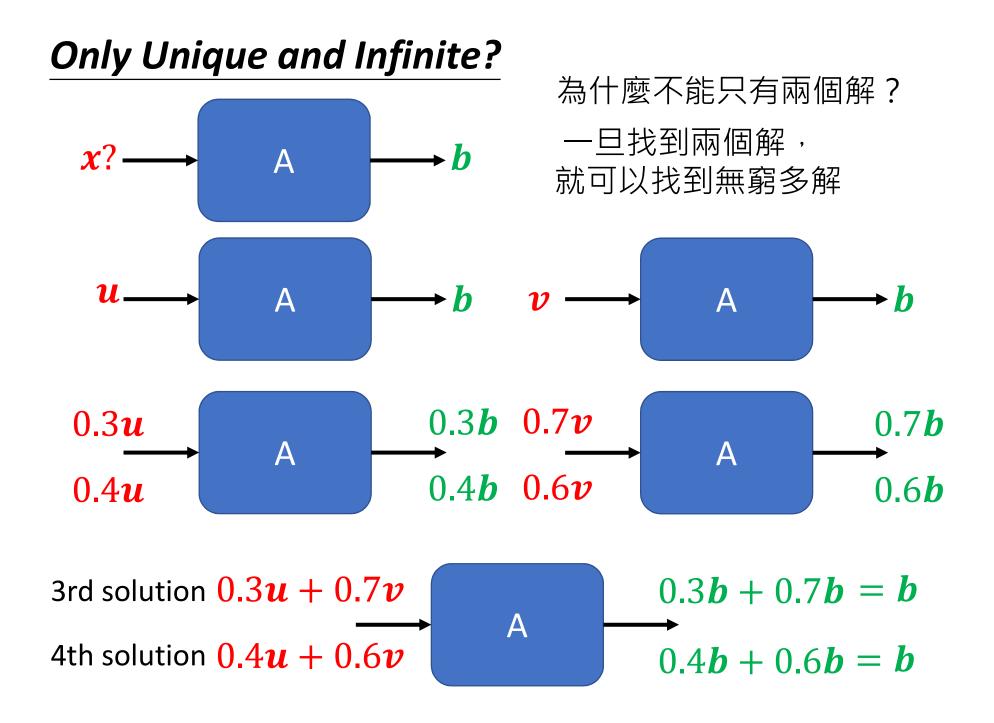
Given A and b, let's find x

$$a_{11}x_1 + a_{12}x_2 = b_1$$
 ..... line 1  
 $a_{21}x_1 + a_{22}x_2 = b_2$  ..... line 2

More

Variables?





Solving System of Linear Equations (Chapter 1.4)

### Equivalent

• Two systems of linear equations are **equivalent** if they have exactly **the same solution set**.

$$\begin{cases} 3x_1 + x_2 = 10 \\ x_1 - 3x_2 = 0 \end{cases}$$
Solution set: 
$$\{\begin{bmatrix} 3 \\ 1 \end{bmatrix}\}$$
equivalent
$$\begin{cases} x_1 = 3 \\ x_2 = 1 \end{cases}$$
Solution set: 
$$\{\begin{bmatrix} 3 \\ 1 \end{bmatrix}\}$$

#### Equivalent

- Applying the following three operations on a system of linear equations will produce an equivalent one.
- 1. Interchange

$$\begin{cases} 3x_1 + x_2 = 10 \\ x_1 - 3x_2 = 0 \end{cases} \implies \begin{cases} x_1 - 3x_2 = 0 \\ 3x_1 + x_2 = 10 \end{cases}$$

• 2. Scaling (non zero)

$$\begin{cases} 3x_1 + x_2 = 10 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_1 + 9x_2 = 0 \end{cases} \Rightarrow \begin{cases} 3x_1 + x_2 = 10 \\ -3x_1 + 9x_2 = 0 \\ x_1 - 3x_1 + 9x_2 = 0 \\ x_1 - 3x_1 + 9x_2 = 0 \end{cases}$$

• 3. Row Addition

$$\begin{cases} 3x_1 + x_2 = 10 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \end{cases} \implies \begin{cases} 10x_2 = 10 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \end{cases}$$

# Solving system of linear equation

- Strategy
  - We know how to transform a given system of linear equations into another equivalent one.
  - We do it again and again until the system of linear equation is very simple
  - Finally, we know the answer at a glance.

$$\begin{cases} x_1 & -3x_2 &= 0 \ 3x_1 & +x_2 &= 10 \\ x_1 & x_2 &= 10 \end{cases} \begin{cases} x_1 & -3x_2 &= 0 \\ 10x_2 &= 10 \ 10x_2 &= 10 \ 10x_1/10 \\ x_2 &= 1 \end{cases} \begin{cases} x_1 & -3x_2 &= 0 \\ x_2 &= 1 \end{cases} \begin{cases} x_1 & -3x_2 &= 0 \\ x_2 &= 1 \ 1x_2 &= 1 \end{cases} \end{cases}$$

#### Augmented Matrix

#### • a system of linear equation

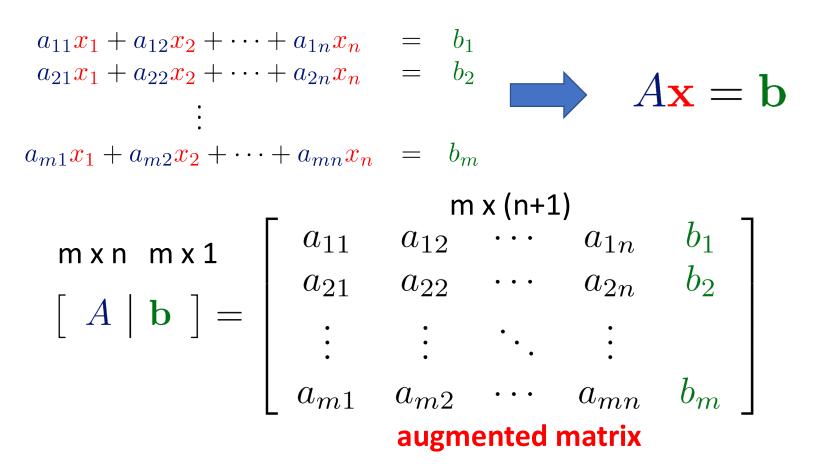
 $\begin{array}{rcl}
a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n &=& b_1 \\
a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n &=& b_2 \\
\vdots & & & & & & & & & & & \\
\end{array}$ 

$$a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m$$

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \quad x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$
coefficient matrix

#### Augmented Matrix

#### a system of linear equation



#### Back to Equivalent

• 1. Interchange

$$\begin{cases} 3x_1 + x_2 = 10 \\ x_1 - 3x_2 = 0 \end{cases} \implies \begin{cases} x_1 - 3x_2 = 0 \\ 3x_1 + x_2 = 10 \end{cases}$$

• 2. Scaling (non zero)

$$\begin{cases} 3x_1 + x_2 = 10 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \end{cases} \implies \begin{cases} 3x_1 + x_2 = 10 \\ -3x_1 + 9x_2 = 0 \\ x_1 - 3x_1 + 9x_2 = 0 \end{cases}$$

• 3. Row Addition

$$\begin{cases} 3x_1 + x_2 = 10 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \end{cases} \implies \begin{cases} 10x_2 = 10 \\ x_1 - 3x_2 = 0 \\ x_1 - 3x_2 = 0 \end{cases}$$

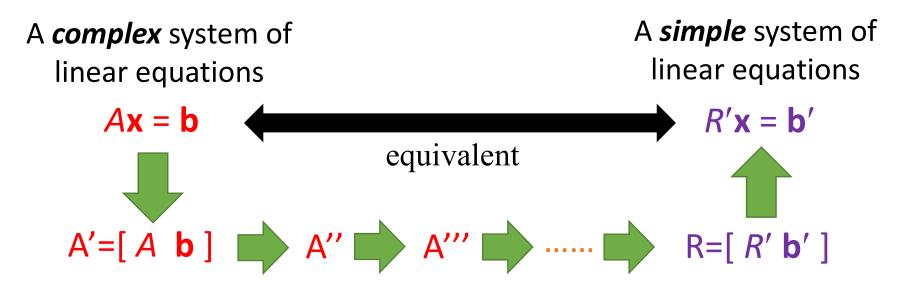
#### Back to Equivalentelementary row operations

• 1. Interchange Interchange any two rows of the matrix

$$\begin{bmatrix} 3 & 1 & 10 \\ 1 & -3 & 0 \end{bmatrix} \longrightarrow \begin{bmatrix} 1 & -3 & 0 \\ 3 & 1 & 10 \end{bmatrix}$$

• 2. Scaling (non zero) Multiply every entry of some row by  
the same nonzero scalar
$$\begin{bmatrix}3 & 1 & 10\\ 1 & -3 & 0\end{bmatrix} \times (-3) \longrightarrow \begin{bmatrix}3 & 1 & 10\\ -3 & 9 & 0\end{bmatrix}$$
• 3. Row Addition Add a multiple of one row of the  
matrix to another row
$$\begin{bmatrix}3 & 1 & 10\\ 1 & -3 & 0\end{bmatrix} \times (-3) \longrightarrow \begin{bmatrix}0 & 10 & 10\\ 1 & -3 & 0\end{bmatrix}$$

# Solving system of linear equation

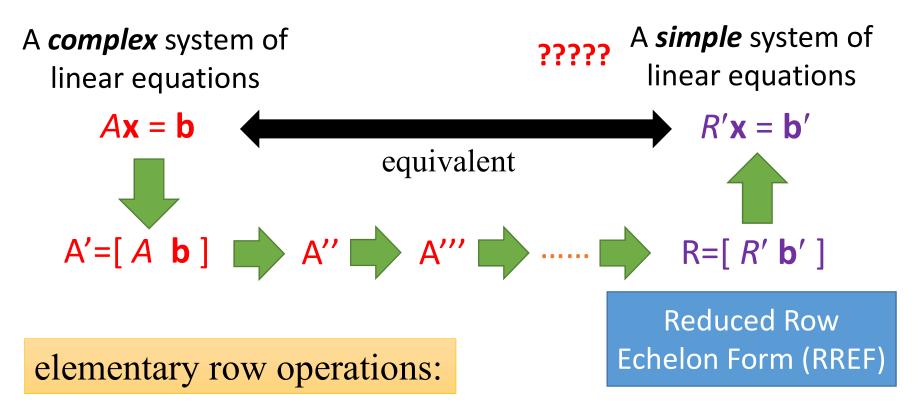


#### elementary row operations

- 1. Interchange any two rows of the matrix
- 2. Multiply every entry of some row by the same nonzero scalar
- 3. Add a multiple of one row of the matrix to another row

$$\begin{cases} x_1 & -3x_2 &= 0 \ X3 \\ 3x_1 & +x_2 &= 10 \end{cases} \longrightarrow \begin{bmatrix} 1 & -3 & 0 \\ 3 & 1 & 10 \end{bmatrix} \times 3 \\ \begin{pmatrix} x_1 & -3x_2 &= 0 \\ 10x_2 &= 10 \ X1/10 \end{pmatrix} \longrightarrow \begin{bmatrix} 1 & -3 & 0 \\ 0 & 10 & 10 \end{bmatrix} \times 1/10 \\ \begin{pmatrix} x_1 & -3x_2 &= 0 \\ x_2 &= 1 \end{pmatrix} \times 3 \\ \begin{pmatrix} x_1 & -3x_2 &= 0 \\ x_2 &= 1 \end{pmatrix} \times 3 \\ \begin{pmatrix} x_1 & -3x_2 &= 0 \\ x_2 &= 1 \end{pmatrix} \times 3 \\ \begin{pmatrix} x_1 & -3x_2 &= 0 \\ x_2 &= 1 \end{pmatrix} \times 3 \\ \begin{pmatrix} 1 & -3 & 0 \\ 0 & 1 & 1 \end{bmatrix} \longrightarrow \times 3 \\ \begin{pmatrix} 1 & -3 & 0 \\ 0 & 1 & 1 \end{bmatrix} \longrightarrow \times 3 \\ \begin{pmatrix} 1 & 0 & 3 \\ 0 & 1 & 1 \end{bmatrix}$$

# Solving system of linear equation

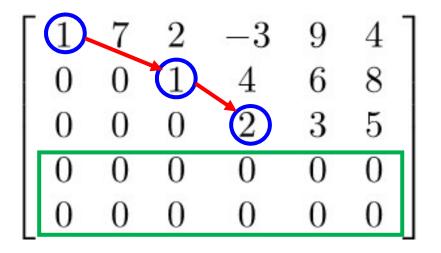


- 1. Interchange any two rows of the matrix
- 2. Multiply every entry of some row by the same nonzero scalar
- 3. Add a multiple of one row of the matrix to another row

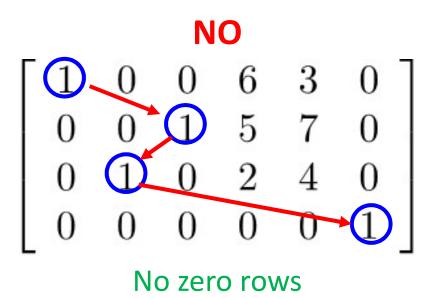
# Reduced Row Echelon Form (RREF) (Chapter 1.4)



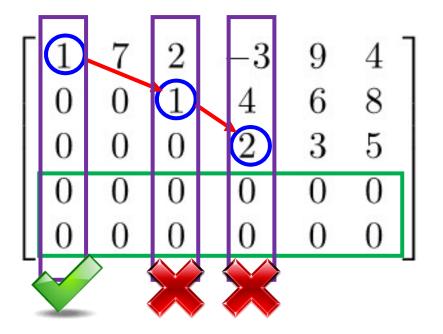
- A system of linear equations is easily solvable if its augmented matrix is in *reduced row echelon form*
- Row Echelon Form (REF)
  - 1. Each nonzero row lies above every zero row
  - 2. The leading entries are in echelon form



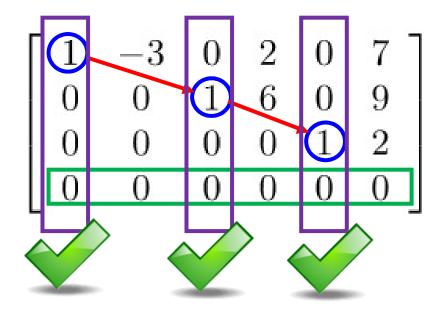
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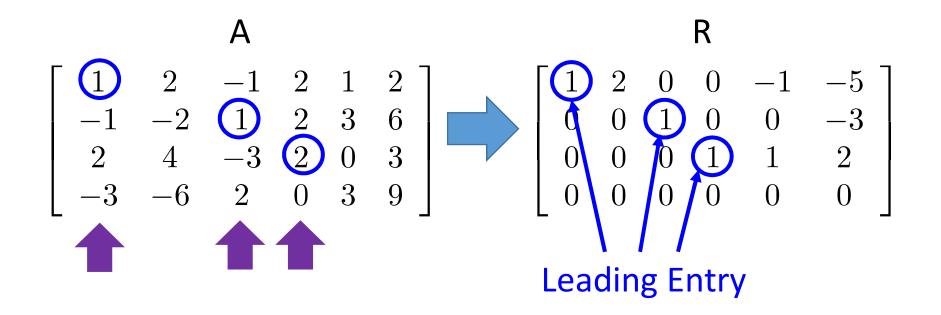


- A system of linear equations is easily solvable if its augmented matrix is in *reduced row echelon form*
- Reduced Row Echelon Form (RREF)
  - 1-2 The matrix is in row echelon form
  - 3. The columns containing the leading entries are standard vectors.



- A system of linear equations is easily solvable if its augmented matrix is in *reduced row echelon form*
- Reduced Row Echelon Form (RREF)
  - 1-2 The matrix is in row echelon form
  - 3. The columns containing the leading entries are standard vectors.

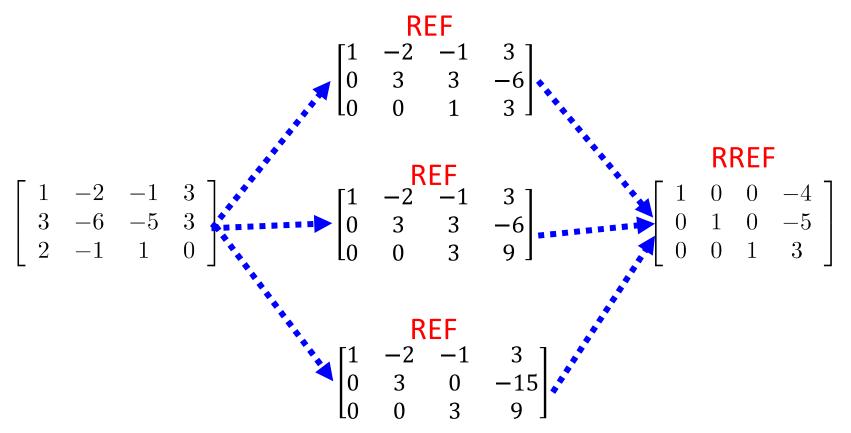




The pivot positions of A are (1,1), (2,3) and (3,4). The pivot columns of A are 1<sup>st</sup>, 3<sup>rd</sup> and 4<sup>th</sup> columns.

## RREF is unique!

• A matrix can be transformed into multiple REFs by row operations, but only one RREF

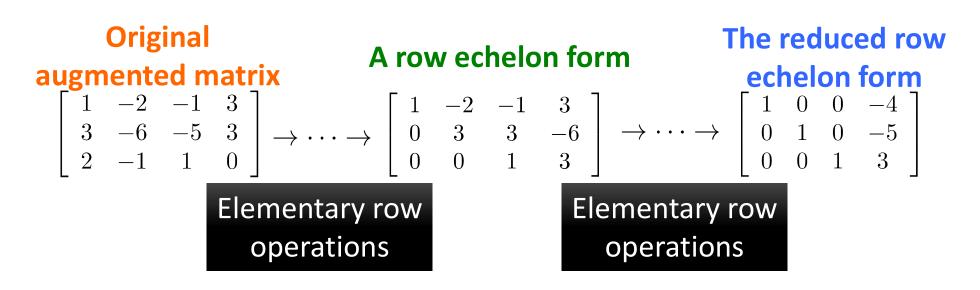


RREF is unique – Proof (by Induction)  $\mathbf{R} = \begin{bmatrix} \mathbf{1} & \mathbf{2} & \mathbf{0} & \mathbf{3} & r_1 \\ \mathbf{0} & \mathbf{0} & \mathbf{1} & \mathbf{4} & r_2 \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & t_1 \end{bmatrix} \quad \mathbf{S} = \begin{bmatrix} \mathbf{1} & \mathbf{2} & \mathbf{0} & \mathbf{3} & s_1 \\ \mathbf{0} & \mathbf{0} & \mathbf{1} & \mathbf{4} & s_2 \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & u_1 \end{bmatrix} \quad \mathbf{R} - \mathbf{S} = \begin{bmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & r_1 - s_1 \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & r_2 - s_2 \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{0} & u_1 \end{bmatrix}$ **R**, **S** are RREF of **A**. Consider  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$  **A** $\mathbf{x} = \mathbf{0} \leftrightarrow \mathbf{R} \mathbf{x} = \mathbf{0}$  and **S** $\mathbf{x} = \mathbf{0}$ ; hence (**R**-**S**) $\mathbf{x} = \mathbf{0} \implies \mathbf{x}_5 = \mathbf{0}$ Claim:  $t_1 \neq 0$ . If  $t_1 = 0$ , consider  $y = \begin{bmatrix} r_1 \\ 0 \\ r_2 \\ 0 \\ 1 \end{bmatrix} \implies \mathbf{R}y = 0 - a \text{ contradiction,}$ since  $y_5 \neq 0$ If  $t_1 \neq 0 \implies t_1 = 1, r_1 = r_2 = 0$ Likewise, we can show that  $u_1 \neq 0 \implies u_1 = 1$ ,  $s_1 = s_2 = 0$  $\mathbf{R} = \mathbf{S}$  $\mathbf{\times}$ 

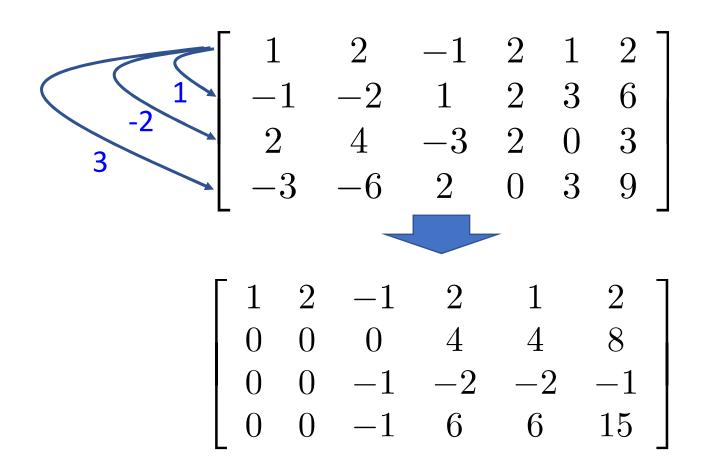
How to find RREF (Chapter 1.4)

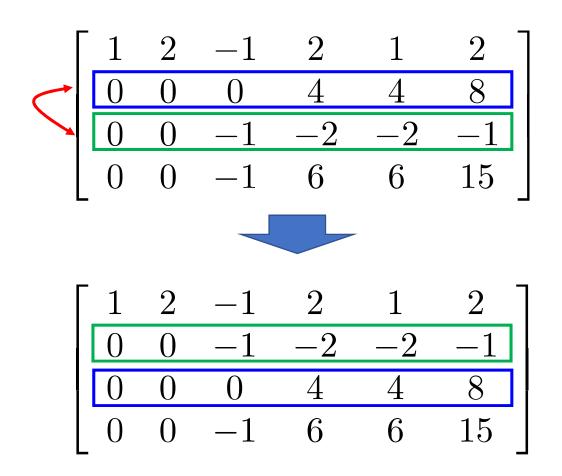
# Reduced Row Echelon Form (RREF)

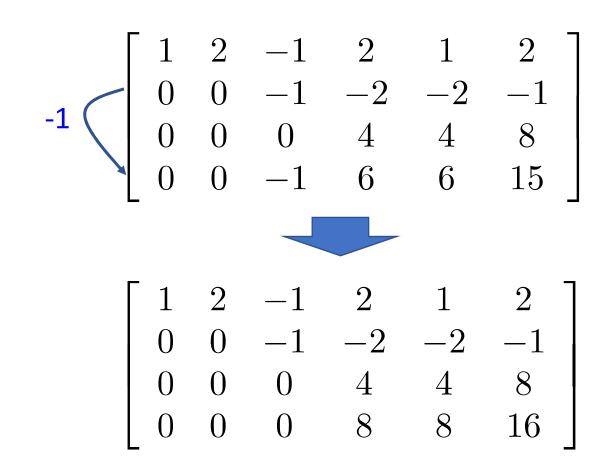
• Gaussian elimination: an algorithm for finding the reduced row echelon form of a matrix.

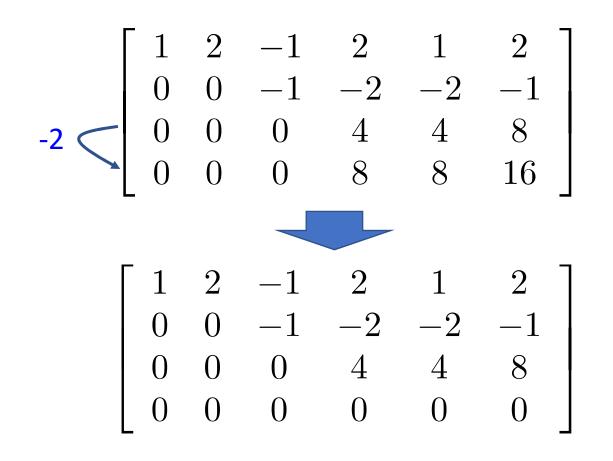


 Because RREF of a matrix is unique, the order of elementary row operations in not important.



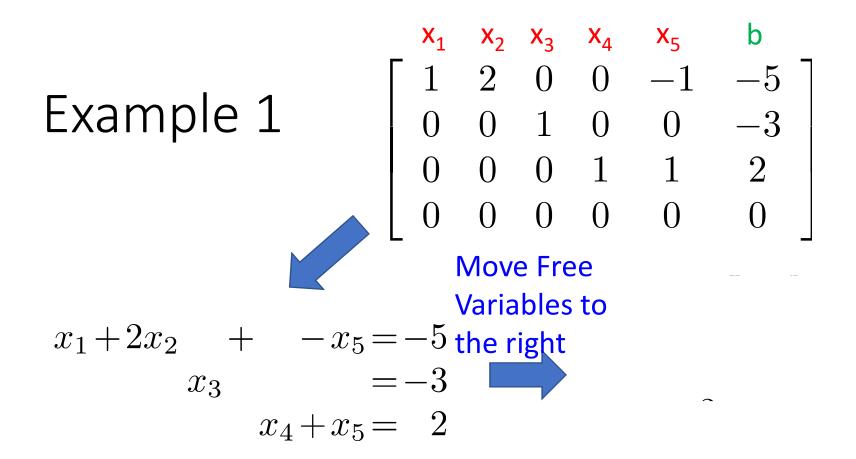


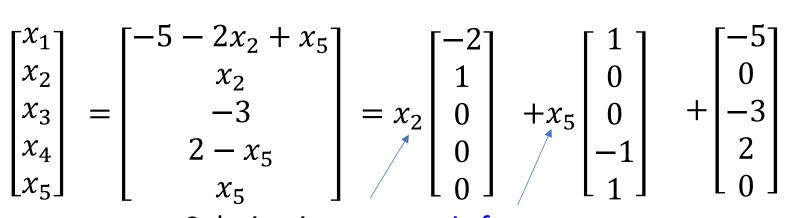




$$\begin{bmatrix} 1 & 2 & -1 & 2 & 1 & 2 \\ 0 & 0 & -1 & -2 & -2 & -1 \\ 0 & 0 & 0 & 4 & 1 & 4 & 1 & 8 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \xrightarrow{2}$$

# $\begin{bmatrix} 1 & 2 & -1 & 0 & -1 & -2 \\ 0 & 0 & -1 & 1 & 0 & 0 & -3 & 3 \\ 0 & 0 & 0 & 1 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}^{1}$





Solution in parametric form

$$\begin{bmatrix} x_1\\x_2\\x_3\\x_4\\x_5 \end{bmatrix} = \begin{bmatrix} -5-2x_2+x_5\\x_2 & 1\\ & -3\\2-x_5 & 3\\x_5 & -1 \end{bmatrix}^{-8} = \begin{bmatrix} 1\\x_2 \begin{bmatrix} 1\\0\\0\\0\\0\end{bmatrix} + \begin{bmatrix} 1\\0\\0\\0\\0\end{bmatrix} + \begin{bmatrix} 1\\0\\0\\0\\0\end{bmatrix} + \begin{bmatrix} -5\\0\\-3\\2\\0\end{bmatrix}$$

$$x_1 free$$

$$x_2 = -\frac{5}{2} - \frac{1}{2}x_1 + \frac{1}{2}x_5$$

$$x_1 = -5 - 2x_2 + x_5$$

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$$x_2 = -\frac{5}{2} - \frac{1}{2}x_1 + \frac{1}{2}x_5$$

$$x_1 = -5 - 2x_2 + x_5$$

$$x_2 = -3$$

$$x_4 + x_5 = 2$$

$$x_4 + x_5 = 2$$

$$x_5 = -3$$

$$x_4 = 2 - x_5$$

$$x_5 = -3$$

$$x_5$$

Span (Chapter 1.6)

# Span

• A vector set 
$$S = \{\boldsymbol{u}_1, \boldsymbol{u}_2, \cdots, \boldsymbol{u}_k\}$$

Span S is the vector set of all linear combinations of u<sub>1</sub>, u<sub>2</sub>, …, u<sub>k</sub>

$$Span S = \{c_1 \boldsymbol{u}_1 + c_2 \boldsymbol{u}_2 + \dots + c_k \boldsymbol{u}_k | for all c_1, c_2, \dots, c_k\}$$

- Vector set V = Span S
  - "S is a generating set for V" or "S generates V"
  - One way to describe a vector set with infinite elements

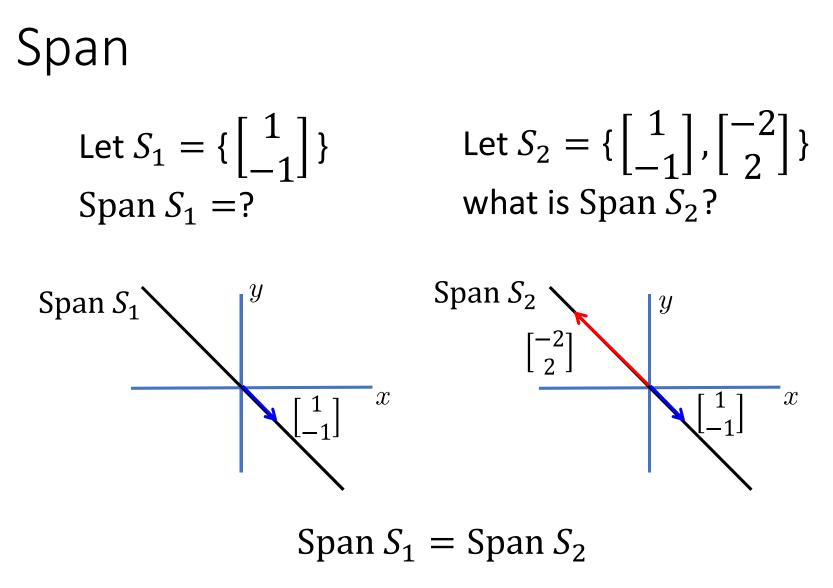
Span 
$$c_1 \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

• Let 
$$S_0 = \{ \begin{bmatrix} 0 \\ 0 \end{bmatrix} \}$$
, what is Span  $S_0$ ?  
• Ans:  $\{ \begin{bmatrix} 0 \\ 0 \end{bmatrix} \}$  (only one member)  
• Let  $S_1 = \{ \begin{bmatrix} 1 \\ -1 \end{bmatrix} \}$ , what is Span  $S_1$ ?

• Let 
$$S_1 = \{ \begin{bmatrix} 1 \\ -1 \end{bmatrix} \}$$
, what is Span  $S_1$ ?

• Span 
$$S_1 = \{ \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ -2 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \dots \end{bmatrix} \}$$

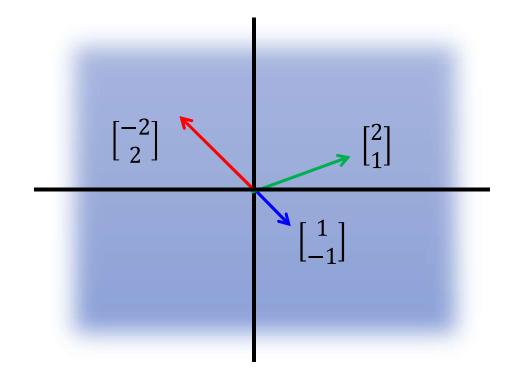
• If S contains a non zero vector, then Span S has infinitely many vectors



(Different number of vectors can generate the same space.)

## Span

• Let 
$$S_3 = \{ \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} -2 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix} \}$$
, what is Span  $S_3$ ?

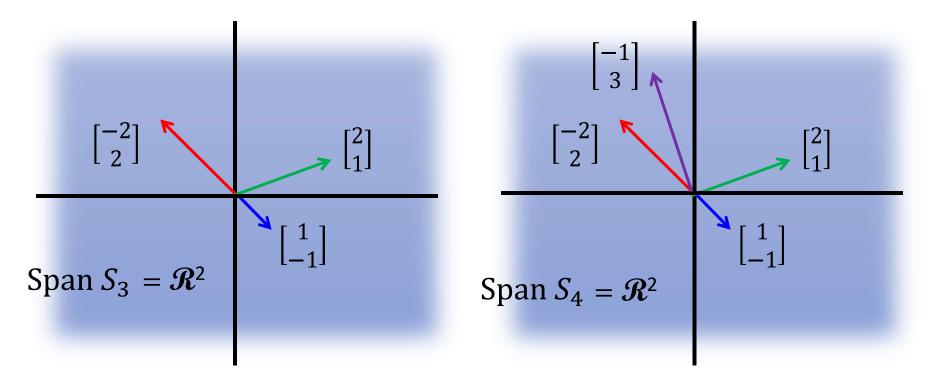


Every vector in  $\mathscr{R}^2$ is their linear combination

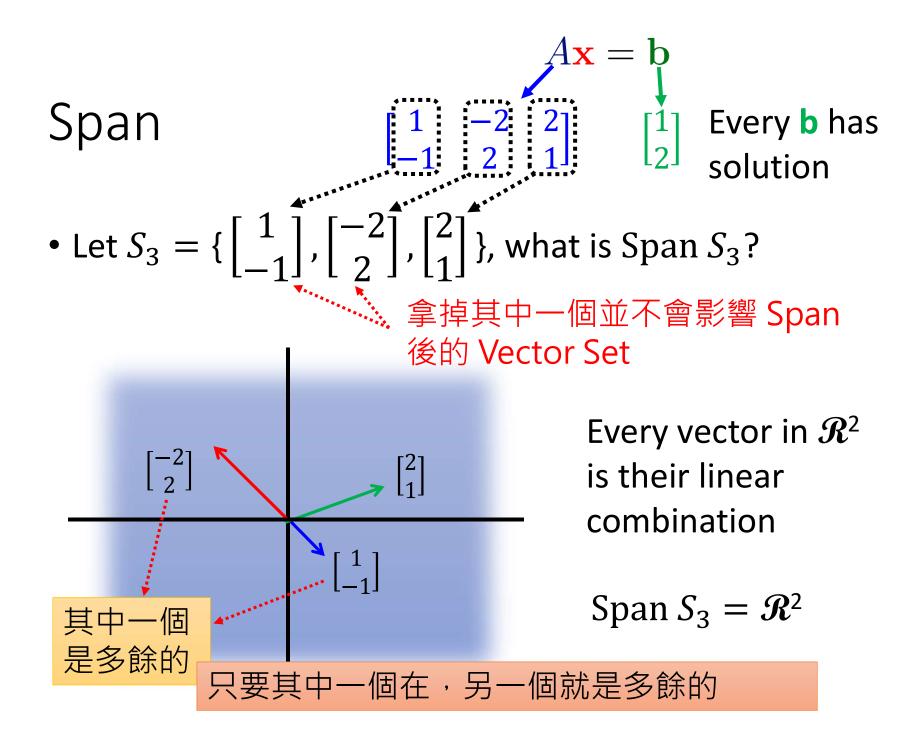
Span  $S_3 = \mathcal{R}^2$ 

## Span

Let  $S_3 = \{ \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} -2 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix} \}$ what is Span  $S_3 = ?$  Let  $S_4 = \{ \begin{bmatrix} 1 \\ -1 \end{bmatrix}, \begin{bmatrix} -2 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 3 \end{bmatrix} \}$ what is Span  $S_4 = ?$ 



# Useless Vector in Span (Chapter 1.6)

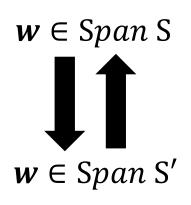


多餘Vector 的特徵Given vector set  $S = \{u_1, u_2 \cdots u_k, v\}$ Given vector set  $S' = \{u_1, u_2 \cdots u_k\}$ v 是多餘:v 是 S 其餘成員的<br/>linear combination

 $\boldsymbol{v} = \boldsymbol{b}_1 \boldsymbol{u}_1 + \boldsymbol{b}_2 \boldsymbol{u}_2 + \dots + \boldsymbol{b}_k \boldsymbol{u}_k$ 

Target

Span S = Span S'

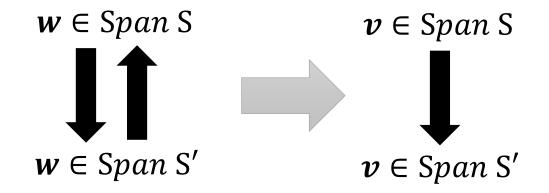


 $w \in \text{Span S}$   $w = c_1 u_1 + c_2 u_2 + \dots + c_k u_k + c v \in \text{Span S'}$   $= (c_1 + cb_1)u_1 + (c_2 + cb_2)u_2 \dots + (c_k + cb_k)u_k$   $w \in \text{Span S'}$   $w = c_1 u_1 + c_2 u_2 + \dots + c_k u_k \quad c = 0$   $= c_1 u_1 + c_2 u_2 + \dots + c_k u_k + c v \in \text{Span S}$ 

 $(\boldsymbol{v} \in \text{Span S}')$ 

多餘Vector 的特徵Given vector set 
$$S = \{u_1, u_2 \cdots u_k, v\}$$
  
Given vector set  $S' = \{u_1, u_2 \cdots u_k\}$  $v$  是多餘:  
Span S = Span S' $v \in S$  其餘成員的  
linear combination  
 $v \in Span S'$ )

 $\boldsymbol{\nu} = 0\boldsymbol{u}_1 + 0\boldsymbol{u}_2 + \dots + 0\boldsymbol{u}_k + 1\boldsymbol{\nu}$ 



$$a_{11}x_{1} + a_{12}x_{2} + \dots + a_{1n}x_{n} = b_{1}$$

$$a_{21}x_{1} + a_{22}x_{2} + \dots + a_{2n}x_{n} = b_{2}$$

$$a_{m1}x_{1} + a_{m2}x_{2} + \dots + a_{mn}x_{n} = b_{m}$$

$$a_{m1}x_{1} + a_{m2}x_{2} + \dots + a_{mn}x_{n} = b_{m}$$

$$x_{1} \begin{bmatrix} a_{11} \\ \vdots \\ a_{m1} \end{bmatrix} + x_{2} \begin{bmatrix} a_{12} \\ \vdots \\ a_{m2} \end{bmatrix} + \dots + x_{n} \begin{bmatrix} a_{1n} \\ \vdots \\ a_{mn} \end{bmatrix} = \begin{bmatrix} b_{1} \\ \vdots \\ b_{m} \end{bmatrix}$$

$$b_{1} = \begin{bmatrix} b_{1} \\ \vdots \\ b_{m} \end{bmatrix}$$

$$b_{2} = b$$

$$b_{3} = b$$

$$b_{4} = b$$

$$b_{4} = b$$

$$b_{2} = b$$

$$b_{3} = b$$

$$b_{4} = b$$

$$constants$$

$$cons$$

Dependent and Independent (Chapter 1.7)

**Ax=0** 

Definition

$$x_1 \begin{bmatrix} a_{11} \\ \vdots \\ a_{m1} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ \vdots \\ a_{m2} \end{bmatrix} + \dots + x_n \begin{bmatrix} a_{1n} \\ \vdots \\ a_{mn} \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

- A set of n vectors {a<sub>1</sub>, a<sub>2</sub>, ..., a<sub>n</sub>} is linearly
   *dependent* Find one
   Obtain many
  - If there exist scalars  $\underline{x_1, x_2, \cdots, x_n}$ , **not all zero**, such that

$$x_1a_1 + x_2a_2 + \dots + x_na_n = 0$$

• A set of n vectors  $\{a_1, a_2, \cdots, a_n\}$  is linearly *independent* 

$$x_1 a_1 + x_2 a_2 + \dots + x_n a_n = 0$$
  
Only if  $x_1 = x_2 = \dots = x_n = 0$  unique

- A set of n vectors  $\{a_1, a_2, \cdots, a_n\}$  is linearly *dependent* 
  - If there exist scalars  $x_1, x_2, \dots, x_n$ , **not all zero**, such that  $x_1a_1 + x_2a_2 + \dots + x_na_n = \mathbf{0}$
- A set of n vectors  $\{a_1, a_2, \cdots, a_n\}$  is linearly *independent*

$$x_1a_1 + x_2a_2 + \dots + x_na_n = 0$$
  
Only if  $x_1 = x_2 = \dots = x_n = 0$ 

$$\begin{cases} \begin{bmatrix} -4\\12\\6 \end{bmatrix}, \begin{bmatrix} -10\\30\\15 \end{bmatrix} \end{cases}$$
 Dependent or Independent?  
dependent  
$$x_1 \begin{bmatrix} -4\\12\\6 \end{bmatrix} + x_2 \begin{bmatrix} -10\\30\\15 \end{bmatrix} = \mathbf{0}$$

- A set of n vectors  $\{a_1, a_2, \cdots, a_n\}$  is linearly *dependent* 
  - If there exist scalars  $x_1, x_2, \dots, x_n$ , **not all zero**, such that  $x_1a_1 + x_2a_2 + \dots + x_na_n = \mathbf{0}$
- A set of n vectors  $\{a_1, a_2, \cdots, a_n\}$  is linearly *independent*

 $x_1a_1 + x_2a_2 + \dots + x_na_n = 0$ Only if  $x_1 = x_2 = \dots = x_n = 0$ 

 $\left\{ \begin{bmatrix} 6\\3\\3\end{bmatrix}, \begin{bmatrix} 1\\8\\3\end{bmatrix}, \begin{bmatrix} 7\\11\\6\end{bmatrix} \right\}$  Dependent or Independent? dependent

$$\begin{array}{c} x_1 \begin{bmatrix} 6\\3\\3 \end{bmatrix} + x_2 \begin{bmatrix} 1\\8\\3 \end{bmatrix} + x_3 \begin{bmatrix} 7\\11\\6 \end{bmatrix} = \mathbf{0} \\ \begin{array}{c} \mathbf{0} \\ \mathbf{1} \end{bmatrix}$$

- A set of n vectors  $\{a_1, a_2, \cdots, a_n\}$  is linearly *dependent* 
  - If there exist scalars  $x_1, x_2, \dots, x_n$ , **not all zero**, such that  $x_1a_1 + x_2a_2 + \dots + x_na_n = \mathbf{0}$
- A set of n vectors  $\{a_1, a_2, \cdots, a_n\}$  is linearly *independent*

 $x_1a_1 + x_2a_2 + \dots + x_na_n = 0$ Only if  $x_1 = x_2 = \dots = x_n = 0$ 

 $\left\{ \begin{bmatrix} 3\\-1\\7 \end{bmatrix}, \begin{bmatrix} 0\\0\\0 \end{bmatrix}, \begin{bmatrix} -2\\5\\1 \end{bmatrix} \right\}$  Dependent or Independent? dependent

$$x_{1} \begin{bmatrix} 3 \\ -1 \\ 7 \end{bmatrix} + x_{2} \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + x_{3} \begin{bmatrix} -2 \\ 5 \\ 1 \end{bmatrix} = \mathbf{0}$$

Any set containing zero vector would be linearly dependent

### Linearly Dependent

(for  $n \ge 2$ )

Given a vector set  $\{\mathbf{a}_1, \mathbf{a}_2, ..., \mathbf{a}_n\}$ , there exists scalars  $x_1, x_2, ..., x_n$ , that are **not all zero**, such that  $x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \cdots + x_n\mathbf{a}_n = \mathbf{0}$ .

$$x_{1}a_{1} + x_{2}a_{2} \dots + x_{i}a_{i} + \dots + x_{n}a_{n} = 0$$
  
永遠可以找到某一項有非 0係數  $x_{i} \neq 0$   
 $x_{1}a_{1} + x_{2}a_{2} \dots + x_{n}a_{n} = -x_{i}a_{i}$   
 $-\left(\frac{x_{n}}{x_{1}}\right)a_{1} - \left(\frac{x_{n}}{x_{2}}\right)a_{2} \dots - \left(\frac{x_{n}}{x_{i}}\right)a_{n} = a_{i}$ 

Given a vector set  $\{\mathbf{a}_1, \mathbf{a}_2, ..., \mathbf{a}_n\}$ , if there exists any  $\mathbf{a}_i$  that is a linear combination of other vectors

### Linearly Dependent

(for  $n \ge 2$ )

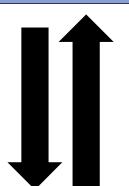
Given a vector set  $\{\mathbf{a}_1, \mathbf{a}_2, ..., \mathbf{a}_n\}$ , there exists scalars  $x_1, x_2, ..., x_n$ , that are **not all zero**, such that  $x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \cdots + x_n\mathbf{a}_n = \mathbf{0}$ .

$$a_i = c_1 a_1 + c_2 a_2 ... + c_n a_n$$
  
- $c_1 a_1 - c_2 a_2 ... + a_i ... - c_n a_n = 0$   
至少這項有非 0係數  
 $x_i \neq 0$ 

Given a vector set  $\{\mathbf{a}_1, \mathbf{a}_2, ..., \mathbf{a}_n\}$ , if there exists any  $\mathbf{a}_i$  that is a linear combination of other vectors

### Linearly Dependent = Vector Set 中有多餘的

Given a vector set  $\{\mathbf{a}_1, \mathbf{a}_2, ..., \mathbf{a}_n\}$ , there exists scalars  $x_1, x_2, ..., x_n$ , that are **not all zero**, such that  $x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \cdots + x_n\mathbf{a}_n = \mathbf{0}$ .



(for 
$$n \ge 2$$
)

Given a vector set  $\{\mathbf{a}_1, \mathbf{a}_2, ..., \mathbf{a}_n\}$ , if there exists any  $\mathbf{a}_i$  that is a linear combination of other vectors

Linearly Independent

Ax=0 has infinite solutions

Columns of A are dependent → If Ax=b has a solution, it will have Infinite solutions

We can find non-zero solution **u** such that Au = 0

There exists **v** such that Av = b

 $A(\boldsymbol{u}+\boldsymbol{v})=\mathbf{b}$ 

u + v is another solution different to v

If Ax=b has Infinite solutions → Columns of A are dependent
 Ax=0 has infinite solutions

# Column Correspondence Theorem (Chapter 1.7)

# Column Correspondence Theorem

RREF  

$$A = \begin{bmatrix} a_1 & \cdots & a_n \end{bmatrix} \longrightarrow R = \begin{bmatrix} r_1 & \cdots & r_n \end{bmatrix}$$
If  $a_j$  is a linear combination of the corresponding columns of R with other columns of A  
 $a_5 = -a_1 + a_4$   
 $r_5 = -r_1 + r_4$ 

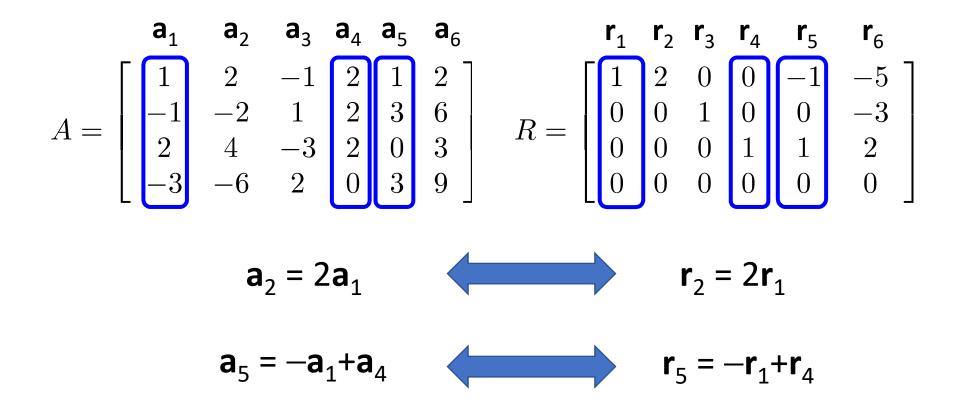
 $a_j$  is a linear combination of the corresponding columns of A with the same coefficients

$$a_3 = 3a_1 - 2a_2$$

If *r<sub>j</sub>* is a linear
combination of
other columns of R

 $r_3 = 3r_1 - 2r_2$ 

# Column Correspondence Theorem - Example





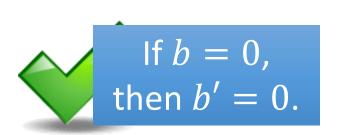
 The RREF of matrix A is R
 Ax = b and Rx = b have the same solution set?



b] is  $\begin{bmatrix} R & b' \end{bmatrix}$ 

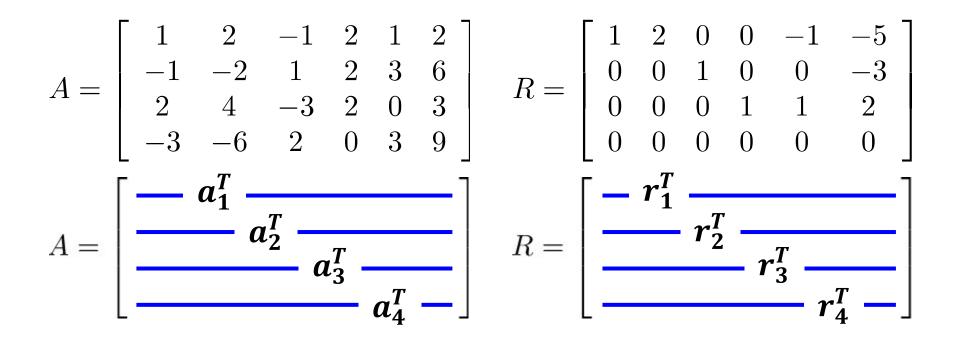
- The RREF of augmented matrix [A Ax = b and Rx = b' have the same solution set
- The RREF of matrix A is R

Ax = 0 and Rx = 0 have the same solution set



### How about Rows?

Are there row correspondence theorem? NO



# Check Independence (Chapter 1.7)

# Checking Independence $S = \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\}$ Linearly independent or not?

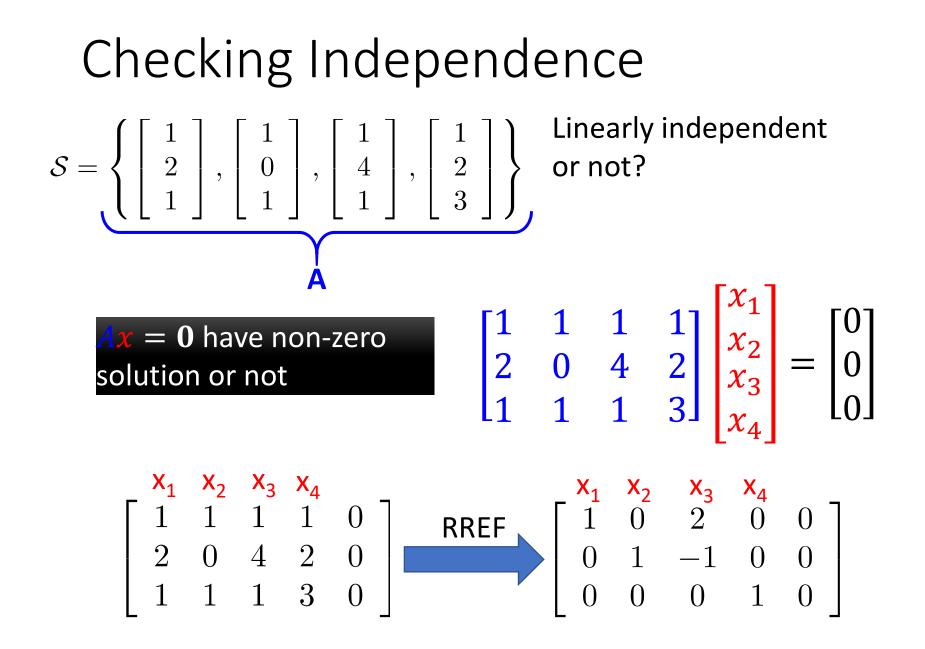
A set of n vectors  $\{\mathbf{a}_1, \mathbf{a}_2, ..., \mathbf{a}_n\}$  is linearly dependent

Given a vector set  $\{a_1, a_2, ..., a_n\}$ , if there exists any  $a_i$  that is a linear combination of other vectors

### matrix A

Given a vector set  $\{\underline{a}_1, \underline{a}_2, \dots, \underline{a}_n\}$ , there exists scalars  $\underline{x}_1, \underline{x}_2, \dots, \underline{x}_n$ , that are **not all zero**, such that  $x_1\underline{a}_1 + x_2\underline{a}_2 + \dots + x_n\underline{a}_n = 0$ . Vector x

Ax = 0 have non-zero solution



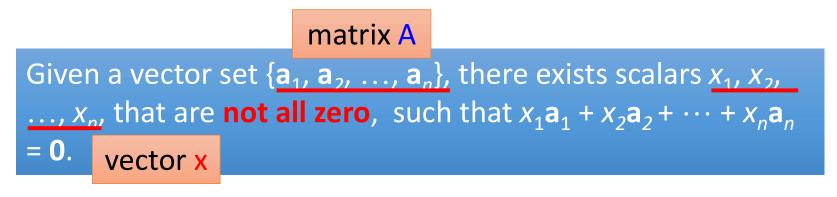
# Checking Independence

$$\begin{bmatrix} x_{1} & x_{2} & x_{3} & x_{4} \\ 1 & 1 & 1 & 1 \\ 2 & 0 & 4 & 2 \\ 1 & 1 & 1 & 3 \\ 0 & 0 & 0 \end{bmatrix} \xrightarrow{\mathsf{RREF}} \begin{bmatrix} x_{1} & x_{2} & x_{3} & x_{4} \\ 1 & 0 & 2 & 0 & 0 \\ 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$
  
dependent  
$$x_{1} + 2x_{3} = 0$$
  
$$x_{2} - x_{3} = 0$$
  
$$x_{4} = 0$$
  
$$\begin{bmatrix} x_{1} \\ x_{2} \\ x_{3} \\ x_{4} \end{bmatrix} = \begin{bmatrix} -2x_{3} \\ x_{3} \\ x_{3} \\ 0 \end{bmatrix} = x_{3} \begin{bmatrix} -2 \\ 1 \\ 1 \\ 0 \end{bmatrix}$$
  
setting  $x_{3} = 1$   
$$\begin{bmatrix} x_{1} \\ x_{2} \\ x_{3} \\ x_{4} \end{bmatrix} = \begin{bmatrix} -2 \\ 1 \\ 1 \\ 0 \end{bmatrix}$$

### Checking Independence $S = \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\}$ Linearly independent or not? 其實這題用看的就 知道答案了!

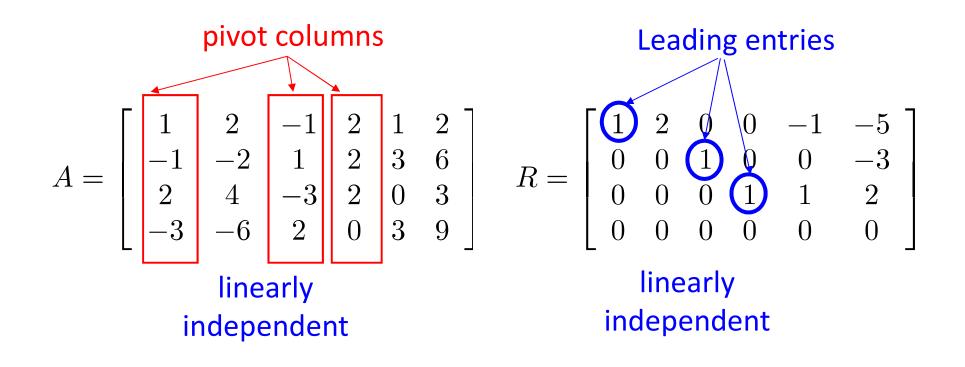
A set of n vectors  $\{a_1, a_2, ..., a_n\}$  is linearly dependent

Given a vector set  $\{a_1, a_2, ..., a_n\}$ , if there exists any  $a_i$  that is a linear combination of other vectors



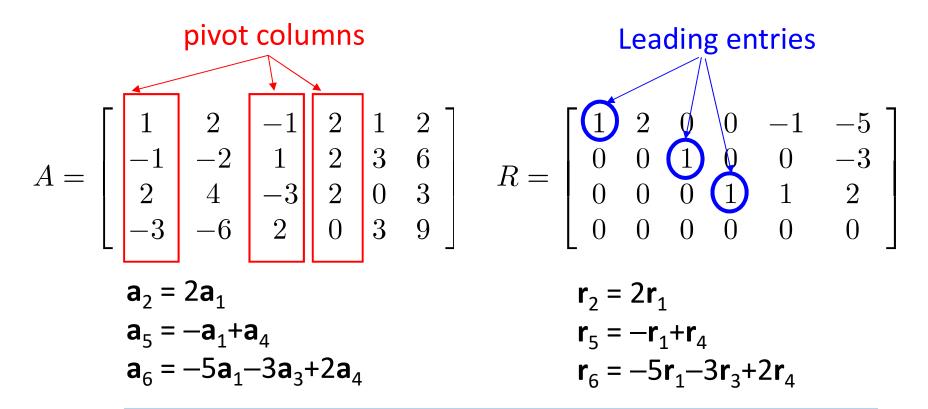
Ax = 0 have non-zero solution

# Column Correspondence Theorem



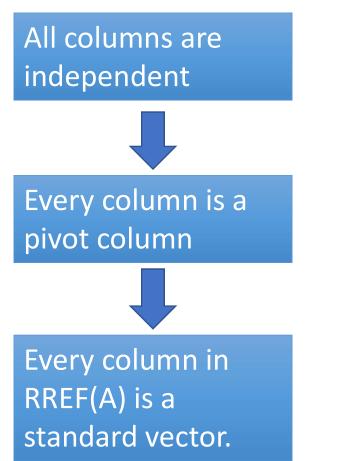
The pivot columns are linearly independent.

# Column Correspondence Theorem



The non-pivot columns are the linear combination of the previous pivot columns.

# Independent

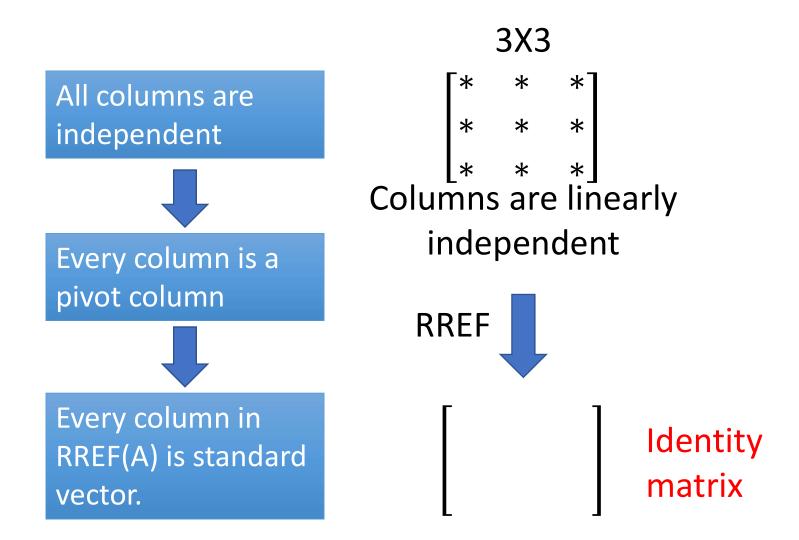


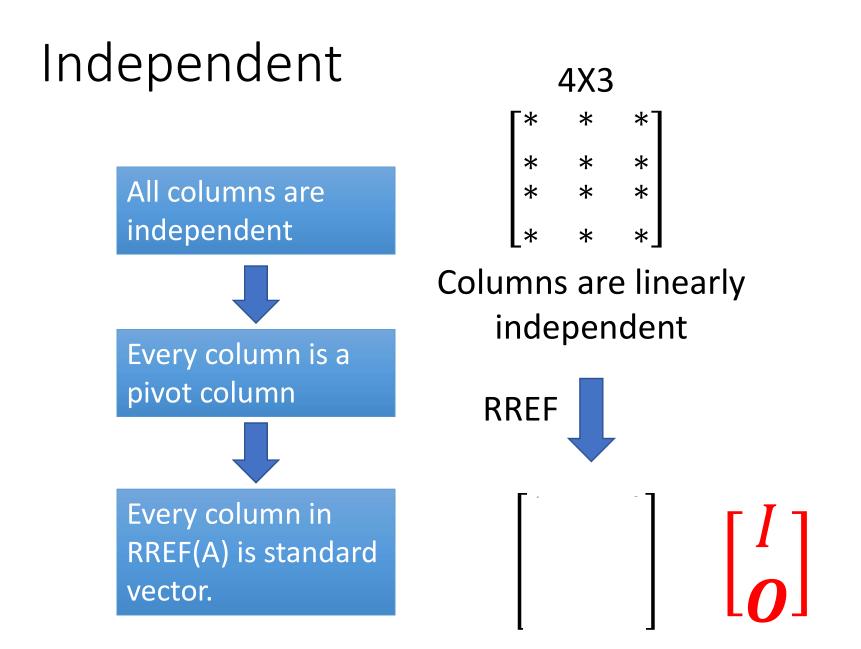
#### Dependent

The column is the linear combination of left pivot column.

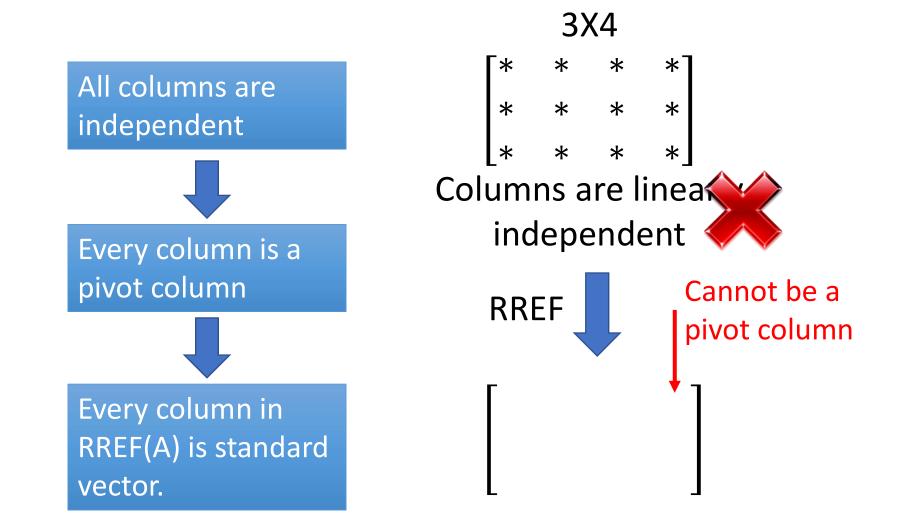
If a column is not pivot

## Independent

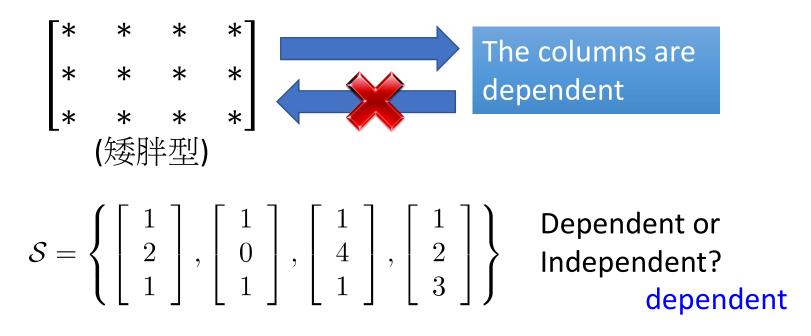




## Independent



 $\begin{bmatrix} 1 & 1 \\ 2 & 2 \\ 3 & 3 \end{bmatrix}$ dependent



More than 3 vectors in R<sup>3</sup> must be dependent.

Independent

More than m vectors in R<sup>m</sup> must be dependent.

Rank of a Matrix (Chapter 1.7)

## Rank

Rank R = Rank A

Maximum number of Independent Columns

 $\|$ 

Number of Pivot

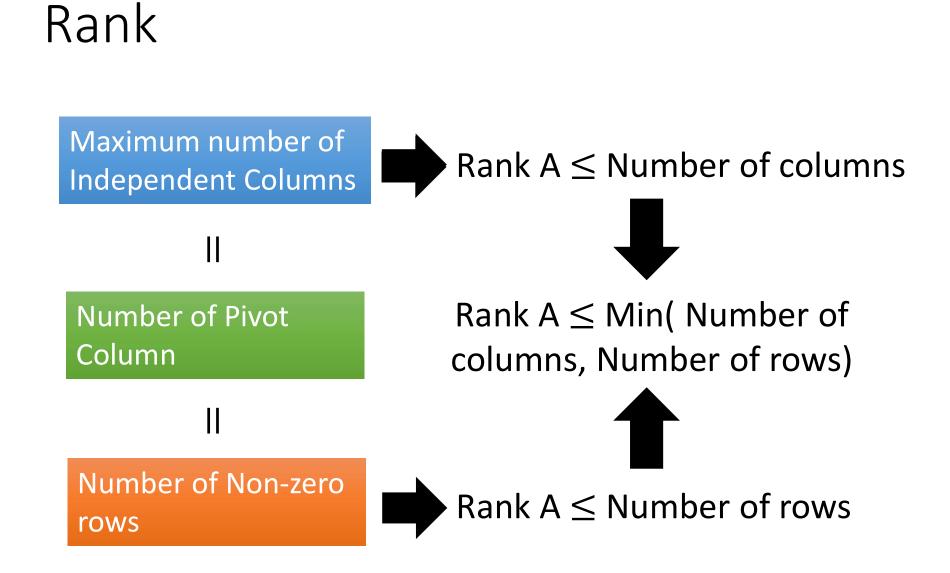
Column

 $A = \begin{bmatrix} 1 & 2 & -1 & 2 & 1 & 2 \\ -1 & -2 & 1 & 2 & 3 & 6 \\ 2 & 4 & -3 & 2 & 0 & 3 \\ -3 & -6 & 2 & 0 & 3 & 9 \end{bmatrix}$ 

Rank = ? 3

$$R = \begin{bmatrix} 1 & 2 & 0 & 0 & -1 & -5 \\ 0 & 0 & 1 & 0 & 0 & -3 \\ 0 & 0 & 0 & 1 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$
  
Rank = ? 3

Number of Non-zero rows



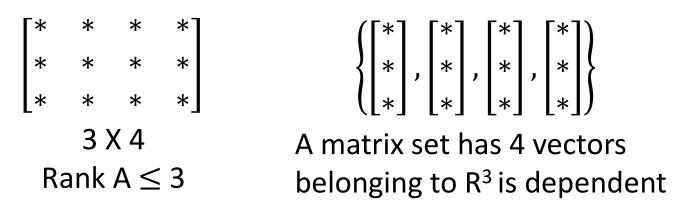
#### Rank

- Given an mxn matrix A:
  - Rank  $A \le min(m, n)$

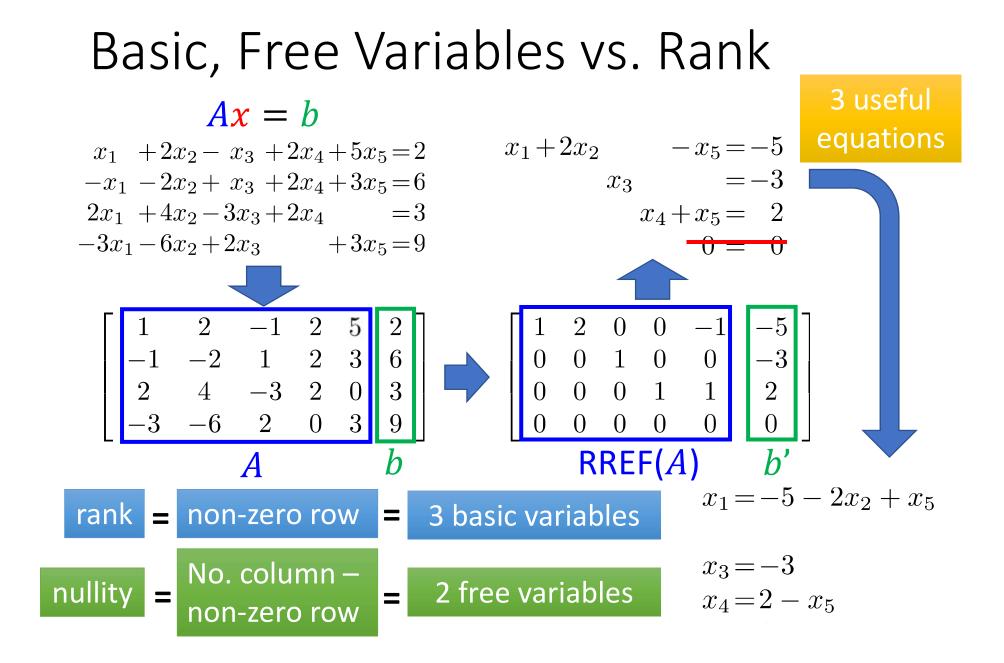
Matrix A is *full rank* if Rank A = min(m,n)

Matrix A is *rank deficient* if Rank A < min(m,n)

- Because "the columns of A are independent" is equivalent to "rank A = n"
  - If m < n, the columns of A are dependent.



In R<sup>m</sup>, you cannot find more than m vectors that are independent.



## Rank

# RankMaximum number of<br/>Independent ColumnsNumber of Pivot<br/>ColumnsNumber of Non-zero<br/>rows of RREFNumber of Basic<br/>Variables

#### Nullity = no. column - rank

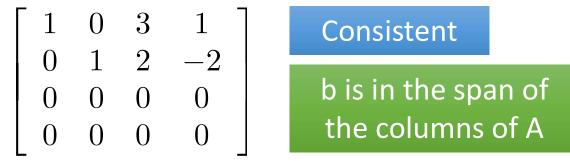


RREF vs. Span (Chapter 1.7)

## Consistent or not

 $0 \cdot x_1 + 0 \cdot x_2 + 0 \cdot x_3 = 1$ 

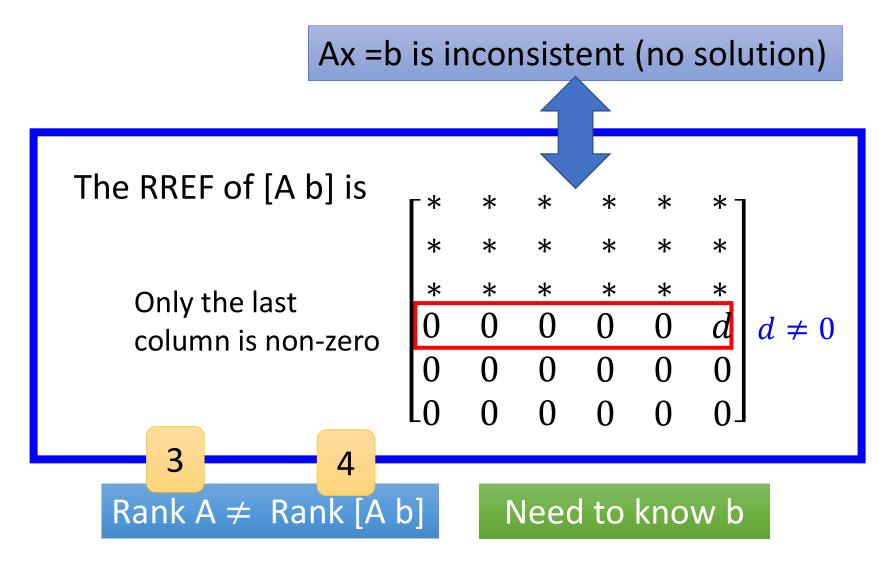
• Given Ax=b, if the reduced row echelon form of [ A b]is

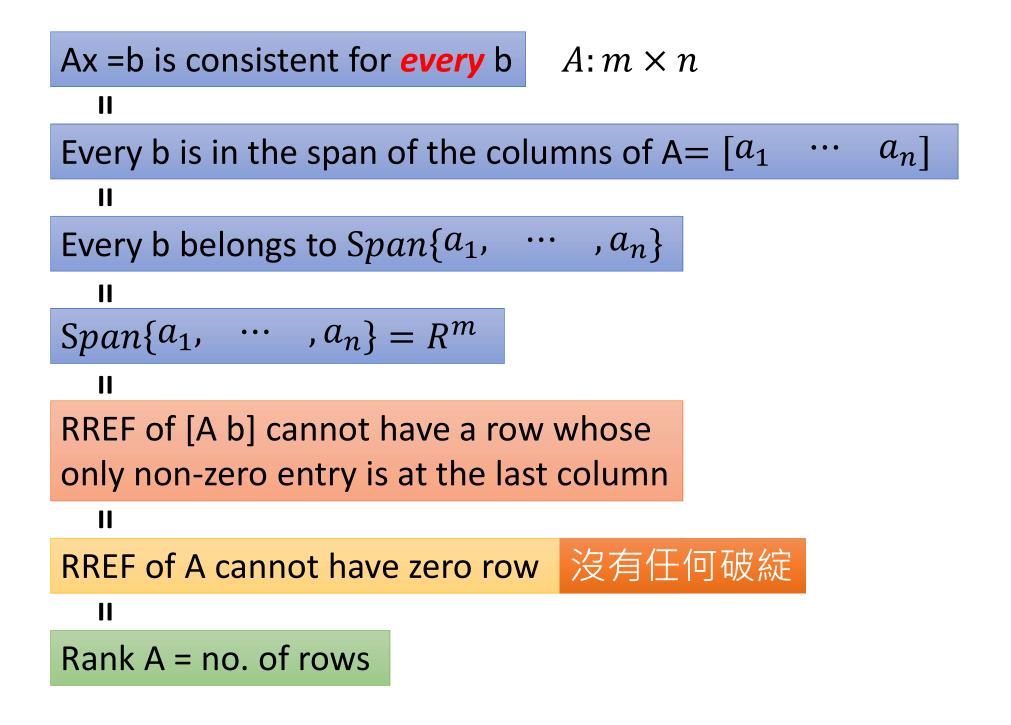


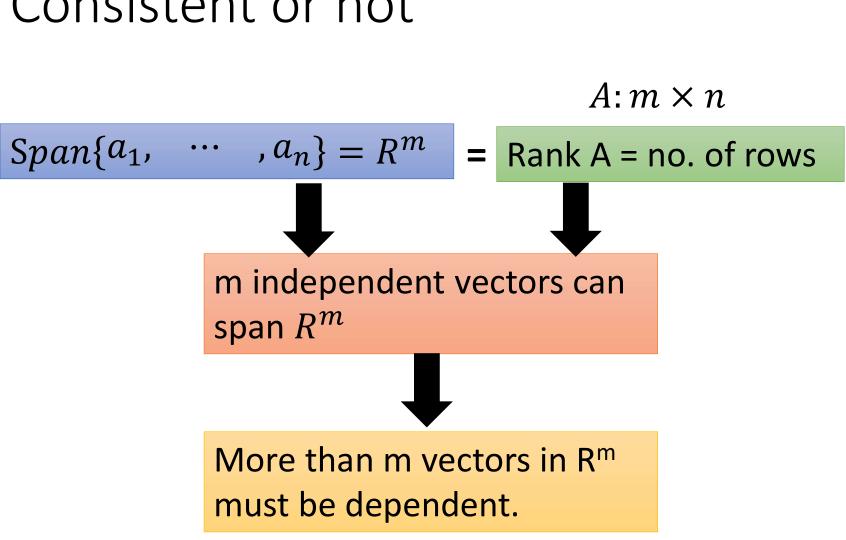
 Given Ax=b, if the reduced row echelon form of [ A b]is inconsistent

b is NOT in the span of the columns of A

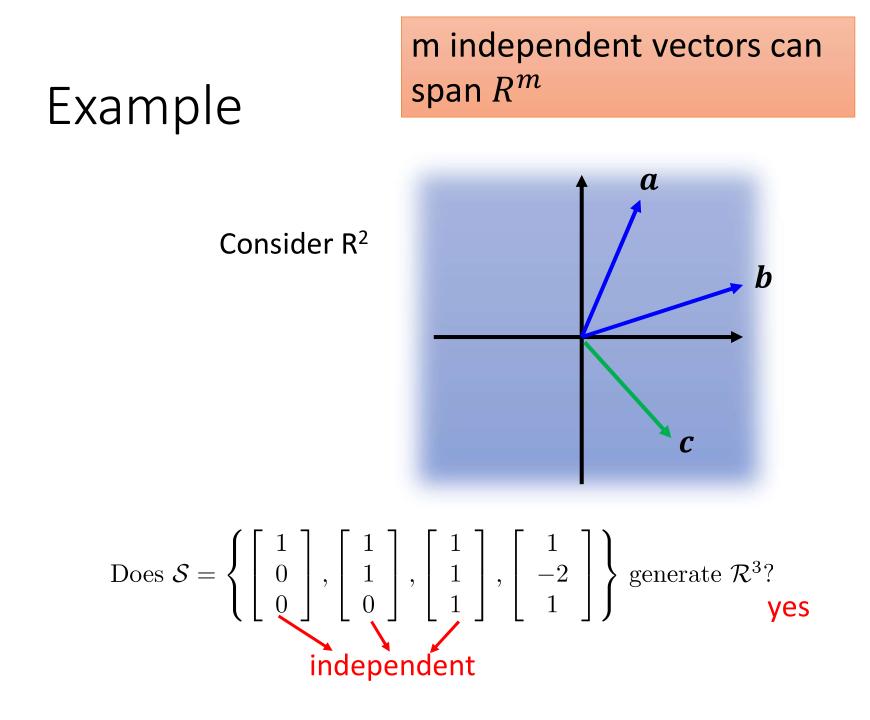
### Consistent or not



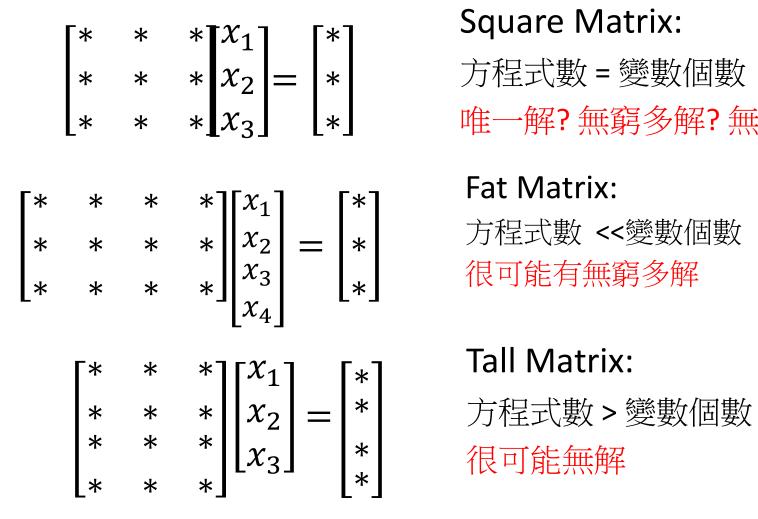




# Consistent or not



#### The Big Picture of Ax = b



Square Matrix: 
 $\begin{bmatrix} * & * & * & X_1 \\ * & * & * & X_2 \\ * & * & * & X_2 \end{bmatrix} = \begin{bmatrix} * \\ * \\ * \end{bmatrix}$   $\overleftarrow{x}$ 
 $\begin{bmatrix} * \\ X_2 \\ Y_2 \end{bmatrix} = \begin{bmatrix} * \\ * \\ * \end{bmatrix}$   $\overleftarrow{x}$ 
 $\begin{bmatrix} * \\ X_2 \end{bmatrix} = \begin{bmatrix} * \\ * \\ * \end{bmatrix}$   $\overleftarrow{x}$ 
 $\begin{bmatrix} * \\ X_2 \end{bmatrix} = \begin{bmatrix} * \\ * \\ * \end{bmatrix}$   $\overleftarrow{x}$ 
 $\begin{bmatrix} * \\ X_2 \end{bmatrix} = \begin{bmatrix} * \\ * \\ * \end{bmatrix}$   $\overleftarrow{x}$ 
 $\begin{bmatrix} * \\ X_2 \end{bmatrix} = \begin{bmatrix} * \\ * \\ * \end{bmatrix}$   $\overleftarrow{x}$ 
 $\begin{bmatrix} * \\ X_2 \end{bmatrix} = \begin{bmatrix} * \\ * \\ * \end{bmatrix}$   $\overleftarrow{x}$ 
 $\overleftarrow{x}$   $\overleftarrow{x}$  <

Fat Matrix:

Tall Matrix:

 $\mathbf{\times}$ 

# Number of Solutions of Ax = b

#### **One Solution**

**No Solution** 

可否找到好的近似解? (Linear regression) Infinite Solutions

可否找到最小的解?

Pseudo-inverse Matrix (based on SVD)



Solutions of Ax = bZero, One, Infinity ...

#### No Solution

#### One Solution

- *b* is NOT a linear combination of column vectors of A
- *b* is a linear combination of column vectors of A
- Ax=0 only has zero solution

#### Infinite Solutions

- *b* is a linear combination of column vectors of A
- Ax=0 has a non-zero solution



