

# General Vector Space (3A)

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# Vector Space

$V$ : non-empty set of objects

defined operations:

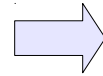
addition

$$\mathbf{u} + \mathbf{v}$$

scalar multiplication

$$k \mathbf{u}$$

if the following axioms are satisfied  
for all object  $\mathbf{u}$ ,  $\mathbf{v}$ ,  $\mathbf{w}$  and all scalar  $k$ ,  $m$



$V$ : vector space

objects in  $V$ : vectors

1. if  $\mathbf{u}$  and  $\mathbf{v}$  are objects in  $V$ , then  $\mathbf{u} + \mathbf{v}$  is in  $V$
2.  $\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$
3.  $\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}$
4.  $\mathbf{0} + \mathbf{u} = \mathbf{u} + \mathbf{0} = \mathbf{u}$  (zero vector)
5.  $\mathbf{u} + (-\mathbf{u}) = (-\mathbf{u}) + (\mathbf{u}) = \mathbf{0}$
6. if  $k$  is any scalar and  $\mathbf{u}$  is objects in  $V$ , then  $k\mathbf{u}$  is in  $V$
7.  $k(\mathbf{u} + \mathbf{v}) = k\mathbf{u} + k\mathbf{v}$
8.  $(k + m)\mathbf{u} = k\mathbf{u} + m\mathbf{u}$
9.  $k(m\mathbf{u}) = (km)\mathbf{u}$
10.  $1(\mathbf{u}) = \mathbf{u}$

# Test for a Vector Space

1. Identify the set  $V$  of objects
2. Identify the addition and scalar multiplication on  $V$
3. Verify  $u + v$  is in  $V$  and  $ku$  is in  $V$   
**closure** under **addition** and **scalar multiplication**
4. Confirm other axioms.

1. if  $u$  and  $v$  are objects in  $V$ , then  $u + v$  is in  $V$
2.  $u + v = v + u$
3.  $u + (v + w) = (u + v) + w$
4.  $0 + u = u + 0 = u$  (zero vector)
5.  $u + (-u) = (-u) + (u) = 0$
6. if  $k$  is any scalar and  $u$  is objects in  $V$ , then  $ku$  is in  $V$
7.  $k(u + v) = ku + kv$
8.  $(k + m)u = ku + mu$
9.  $k(mu) = (km)u$
10.  $1(u) = u$

# Subspace

a subset  $W$  of a vector space  $V$

If the subset  $W$  is itself a vector space  $\Rightarrow$  the subset  $W$  is a **subspace** of  $V$

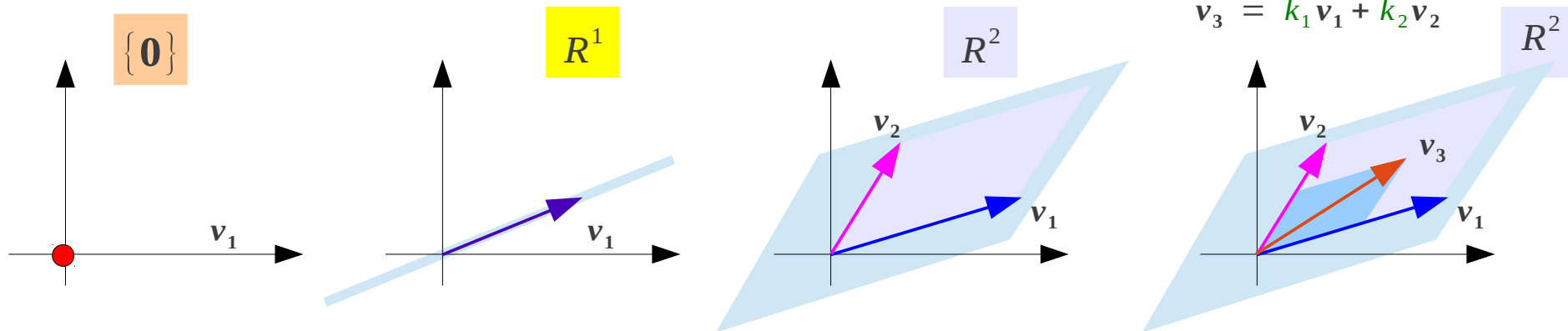
1. if  $u$  and  $v$  are objects in  $W$ , then  $u + v$  is in  $W$
2.  $u + v = v + u$
3.  $u + (v + w) = (u + v) + w$
4.  $0 + u = u + 0 = u$  (zero vector)
5.  $u + (-u) = (-u) + (u) = 0$
6. if  $k$  is any scalar and  $u$  is objects in  $W$ , then  $ku$  is in  $W$
7.  $k(u + v) = ku + kv$
8.  $(k + m)u = ku + mu$
9.  $k(mu) = (km)u$
10.  $1(u) = u$

# Subspace Example (1)

In vector space  $R^2$

any <b>one</b> vector	(linearly indep.)	spans $R^1$	line <u>through 0</u>
any <b>two</b> non-collinear vectors	(linearly indep.)	spans $R^2$	plane
any <b>three or more</b> vectors	(linearly dep.)	spans $R^2$	plane

Subspaces of  $R^2$



# Subspace Example (2)

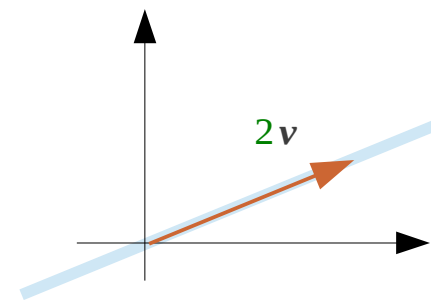
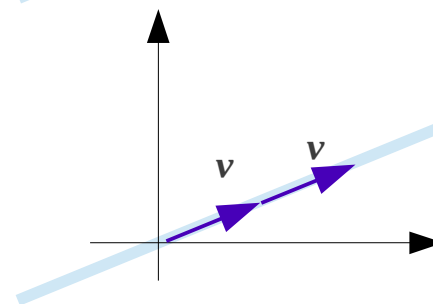
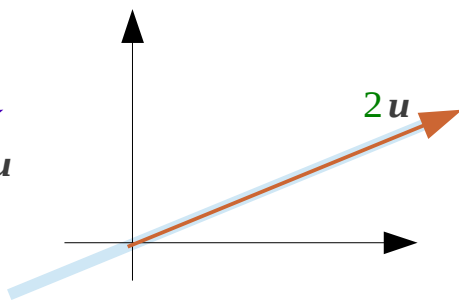
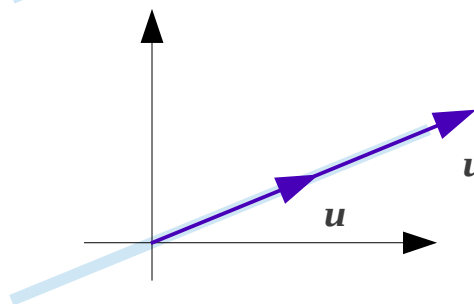
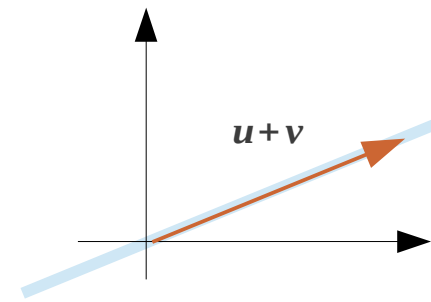
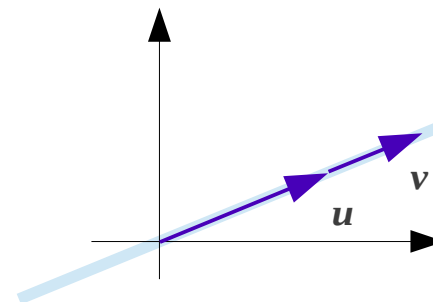
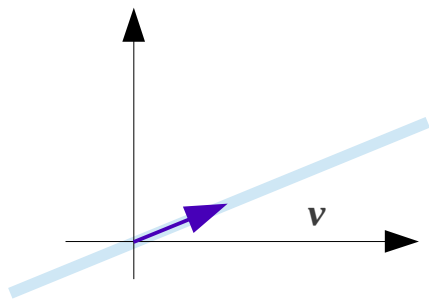
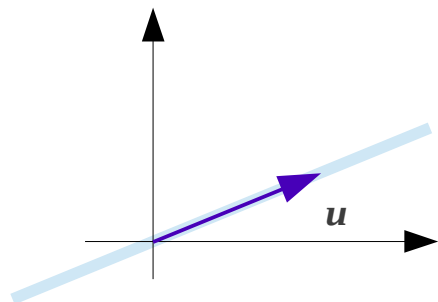
In vector space  $\mathbb{R}^2$

any one vector

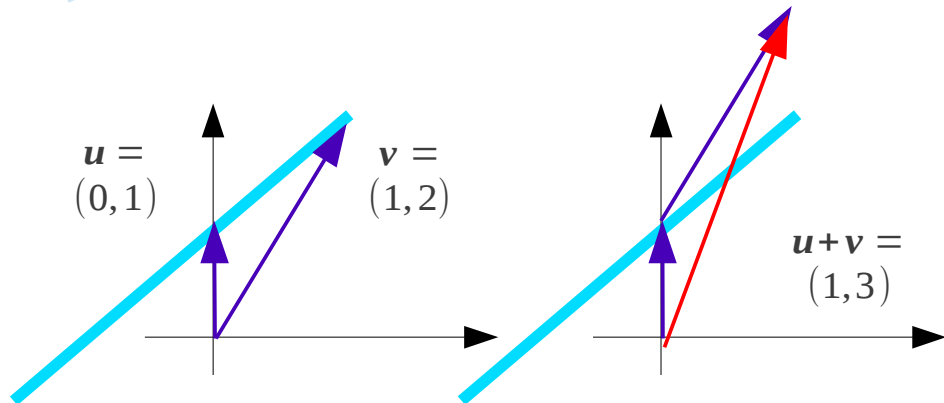
(linearly indep.)

spans  $\mathbb{R}^1$

line through 0



$u = (0,1)$        $v = (1,2)$



~~vector space~~

# Subspace Example (3)

In vector space  $R^3$

any <b>one</b> vector	(linearly indep.)	<b>spans</b>	$R^1$	line <u>through 0</u>
any <b>two</b> non-collinear vectors	(linearly indep.)	<b>spans</b>	$R^2$	plane <u>through 0</u>
any <b>three</b> vectors non-collinear, non-coplanar	(linearly indep.)	<b>spans</b>	$R^3$	3-dim space
any <b>four or more</b> vectors	(linearly dep.)	<b>spans</b>	$R^3$	3-dim space

**Subspaces of**  $R^2$

$\{0\}$

$R^1$

$R^2$

$R^3$

line through 0

plane through 0

3-dim space



# Row & Column Spaces

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}$$

$$\begin{aligned} \mathbf{r}_1 &= \left[ a_{11} \ a_{12} \ \cdots \ a_{1n} \right] \\ \mathbf{r}_2 &= \left[ a_{21} \ a_{22} \ \cdots \ a_{2n} \right] \\ &\quad \vdots \\ \mathbf{r}_m &= \left[ a_{m1} \ a_{m2} \ \cdots \ a_{mn} \right] \end{aligned}$$

$\leftarrow \hspace{10em} \rightarrow$   
 $n$

$\mathbf{r}_i \in R^n$

**ROW Space**      subspace of  $R^n$

$$= \text{span} \{ \mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_m \}$$

**COLUMN Space**      subspace of  $R^m$

$$= \text{span} \{ \mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_n \}$$

$\mathbf{c}_1 \quad \mathbf{c}_2 \quad \cdots \quad \mathbf{c}_n \quad \mathbf{c}_i \in R^m$

$$\begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} \quad \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} \quad \cdots \quad \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix}$$

$\updownarrow$   
 $m$

# Row Space

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \cdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}$$

$$r_i \in R^n$$

$$r_1 = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \end{pmatrix}$$

$$r_2 = \begin{pmatrix} a_{21} & a_{22} & \cdots & a_{2n} \end{pmatrix}$$

$$r_m = \begin{pmatrix} \vdots & \vdots & \cdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}$$



$n$

**ROW Space**      **subspace of**  $R^n$

$$= \text{span}\{r_1, r_2, \dots, r_m\}$$

$$k_1 r_1 + k_2 r_2 + \cdots + k_m r_m$$

$$= k_1 \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \end{pmatrix}$$

$$+ k_2 \begin{pmatrix} a_{21} & a_{22} & \cdots & a_{2n} \end{pmatrix}$$

$$+ k_m \begin{pmatrix} \vdots & \vdots & \cdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}$$

# Column Spaces

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}$$

**COLUMN Space**    subspace of  $R^m$   
 $= \text{span}\{\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_n\}$

$c_i \in R^m$      $\mathbf{c}_1$      $\mathbf{c}_2$      $\mathbf{c}_n$

$$\begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} \quad \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} \quad \cdots \quad \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix}$$

$$k_1 \mathbf{c}_1 + k_2 \mathbf{c}_2 + \cdots + k_n \mathbf{c}_n$$

$$= k_1 \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} + k_2 \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} \quad \cdots \quad + k_n \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix}$$

# Null Space

$$\begin{matrix} & \xleftarrow{n} & & & & & \\ & & & & & & \\ \begin{matrix} \uparrow \\ m \\ \downarrow \end{matrix} & \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} & \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} & \begin{matrix} \uparrow \\ n \\ \downarrow \end{matrix} & = & \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix} & \begin{matrix} \uparrow \\ m \\ \downarrow \end{matrix}
 \end{matrix}$$

**NULL Space**

subspace of  $R^n$

solution space

$$= \begin{pmatrix} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \end{pmatrix} = x_1 \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} + x_2 \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} + \cdots + x_n \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix}$$

$$Ax = x_1 c_1 + x_2 c_2 + \cdots + x_n c_n = 0$$

$$Ax = 0$$

$$Ax = x_1 c_1 + x_2 c_2 + \cdots + x_n c_n = b$$

$$Ax = b$$



# Solution Space of $\mathbf{Ax}=\mathbf{b}$ (1)

$$\left( \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right)$$

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

$$0 = 1$$

$$\left( \begin{array}{ccc|c} 1 & 0 & 3 & -1 \\ 0 & 1 & -4 & 2 \\ 0 & 0 & 0 & 0 \end{array} \right)$$

$$1 \cdot x_1 + 3 \cdot x_3 = -1$$

$$1 \cdot x_2 - 4 \cdot x_3 = 2$$

$$\left( \begin{array}{ccc|c} 1 & -5 & 1 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right)$$

$$1 \cdot x_1 - 5 \cdot x_2 + 1 \cdot x_3 = 4$$

Solve for a leading variable

$$x_1 = -1 - 3 \cdot x_3$$

$$x_1 = 4 + 5 \cdot x_2 - 1 \cdot x_3$$

$$x_2 = 2 + 4 \cdot x_3$$

Treat a free variable as a parameter

$$x_3 = t$$

$$x_2 = s \quad x_3 = t$$

$$\begin{cases} x_1 = -1 - 3t \\ x_2 = 2 + 4t \\ x_3 = t \end{cases}$$

$$\begin{cases} x_1 = 4 + 5s - 1t \\ x_2 = s \\ x_3 = t \end{cases}$$

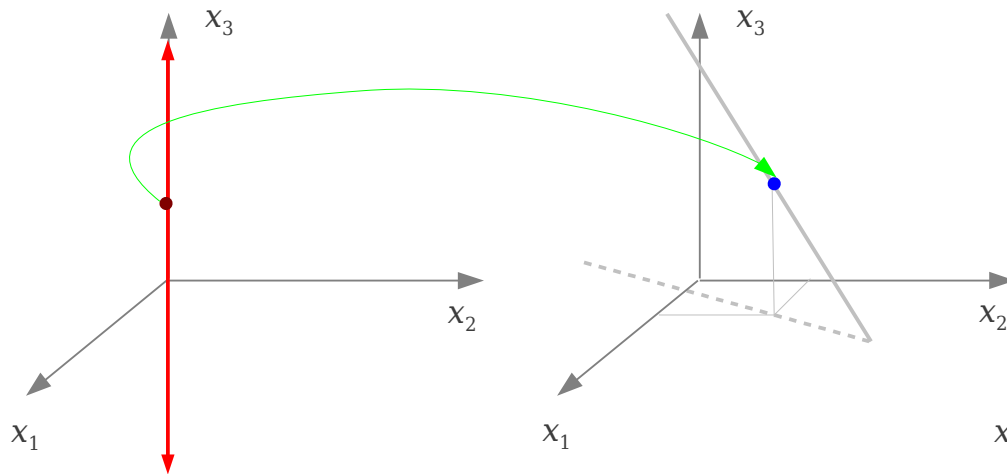
# Solution Space of $\mathbf{Ax}=\mathbf{b}$ (2)

$$\begin{cases} x_1 = -1 - 3t \\ x_2 = 2 + 4t \\ x_3 = t \end{cases} \quad \leftarrow \text{free variable}$$

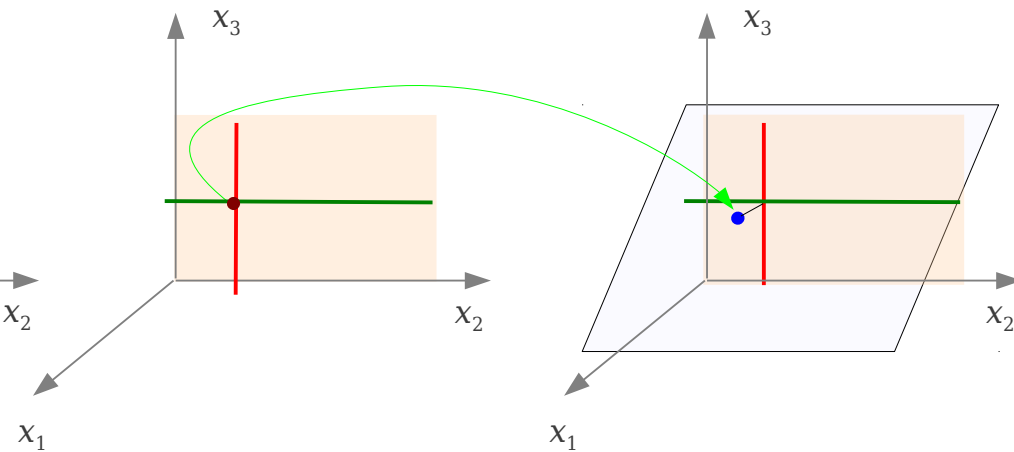
$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix} + t \begin{bmatrix} -3 \\ 4 \\ 1 \end{bmatrix}$$

$$\begin{cases} x_1 = 4 + 5s - 1t \\ x_2 = s \\ x_3 = t \end{cases} \quad \leftarrow \text{free variable}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 0 \\ 0 \end{bmatrix} + s \begin{bmatrix} 5 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$



*infinitely many solutions*



*infinitely many solutions*

# Solution Space of $\mathbf{Ax}=\mathbf{b}$ (3)

$$\left( \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right)$$

$$\left( \begin{array}{ccc|c} 1 & 0 & 3 & -1 \\ 0 & 1 & -4 & 2 \\ 0 & 0 & 0 & 0 \end{array} \right)$$

$$\left( \begin{array}{ccc|c} 1 & -5 & 1 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right)$$

$$\begin{cases} x_1 = -1 - 3t \\ x_2 = 2 + 4t \\ x_3 = t \end{cases}$$

$$\begin{cases} x_1 = 4 + 5s - 1t \\ x_2 = s \\ x_3 = t \end{cases}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix} + t \begin{bmatrix} -3 \\ 4 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 0 \\ 0 \end{bmatrix} + s \begin{bmatrix} 5 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

General  
Solution of  
 $\mathbf{Ax} = \mathbf{b}$



Particular  
Solution of  
 $\mathbf{Ax} = \mathbf{b}$

General  
Solution of  
 $\mathbf{Ax} = \mathbf{0}$

Particular  
Solution of  
 $\mathbf{Ax} = \mathbf{b}$

General  
Solution of  
 $\mathbf{Ax} = \mathbf{0}$



# Linear System & Inner Product (1)

## Linear Equations

### Corresponding Homogeneous Equation

$$a_1 x_1 + a_2 x_2 + \cdots + a_n x_n = 0$$

$$\mathbf{a} = (a_1, a_2, \cdots, a_n)$$

$$\mathbf{x} = (x_1, x_2, \cdots, x_n)$$

normal vector

$$\mathbf{a} \cdot \mathbf{x} = b$$

$$\mathbf{a} \cdot \mathbf{x} = 0$$

each **solution** vector  $\mathbf{x}$  of a **homogeneous** equation  
**orthogonal** to the coefficient vector  $\mathbf{a}$

## Homogeneous Linear System

$$a_{11} x_1 + a_{12} x_2 + \cdots + a_{1n} x_n = 0$$

$$a_{21} x_1 + a_{22} x_2 + \cdots + a_{2n} x_n = 0$$

... ..

$$a_{m1} x_1 + a_{m2} x_2 + \cdots + a_{mn} x_n = 0$$

$$\mathbf{r}_1 \cdot \mathbf{x} = 0$$

$$\mathbf{r}_2 \cdot \mathbf{x} = 0$$

...

$$\mathbf{r}_m \cdot \mathbf{x} = 0$$

# Linear System & Inner Product (2)

## Homogeneous Linear System

$$\begin{array}{rcl} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = 0 & \mathbf{r}_1 \cdot \mathbf{x} = 0 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = 0 & \mathbf{r}_2 \cdot \mathbf{x} = 0 \\ \cdots \quad \cdots \quad \cdots \quad \cdots \quad \cdots & \cdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = 0 & \mathbf{r}_m \cdot \mathbf{x} = 0 \end{array}$$

each **solution** vector  $\mathbf{x}$  of a **homogeneous** equation  
**orthogonal** to the row vector  $\mathbf{r}_i$  of the coefficient matrix

Homogeneous Linear System  $\mathbf{A} \cdot \mathbf{x} = 0$   $\mathbf{A} : m \times n$

**solution set** consists of all vectors in  $R^n$   
that are **orthogonal** to every row vector of  $\mathbf{A}$

# Linear System & Inner Product (3)

Non-Homogeneous Linear System

$$\mathbf{A} \cdot \mathbf{x} = \mathbf{b}$$

$$\mathbf{A} : m \times n$$

Homogeneous Linear System

$$\mathbf{A} \cdot \mathbf{x} = \mathbf{0}$$

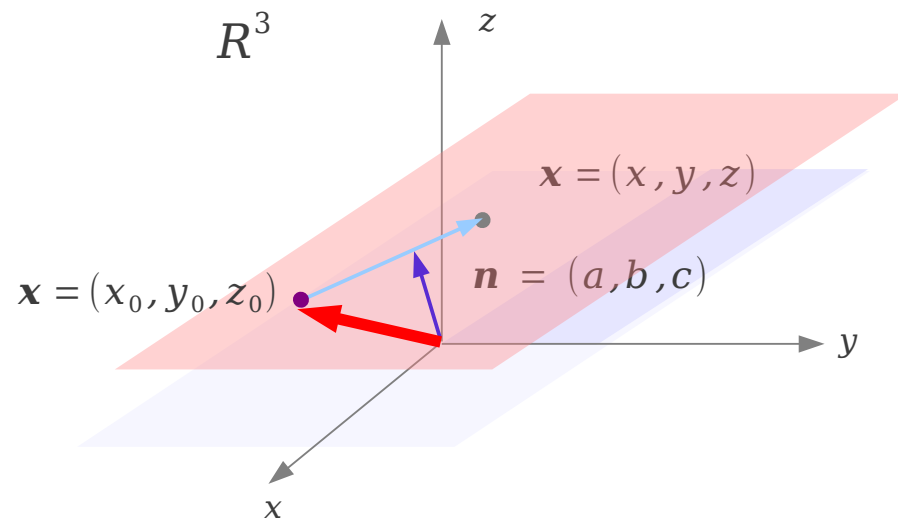
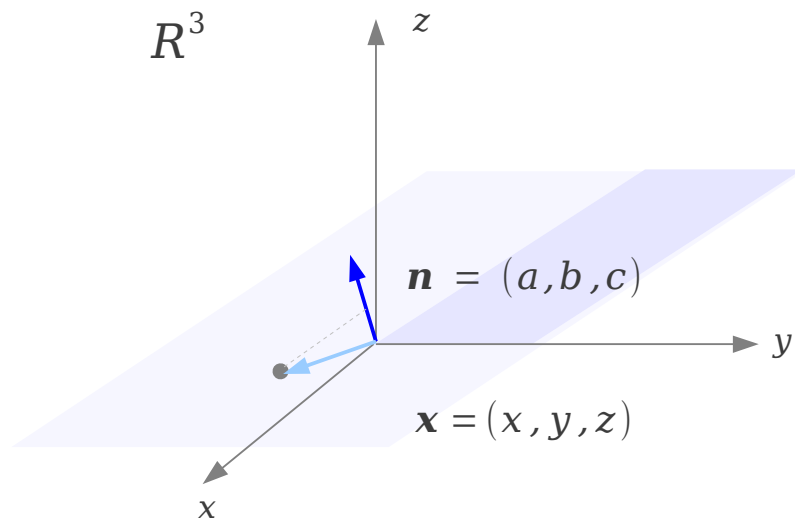
a particular solution

$$\mathbf{A} \cdot \mathbf{x} = \mathbf{b}$$

solution set consists of all vectors in  $R^n$  that are **orthogonal** to every row vector of  $\mathbf{A}$

+

a particular solution  $\mathbf{x}_0$   $\mathbf{A} \cdot \mathbf{x}_0 = \mathbf{b}$



# Linear System & Inner Product (4)

$$\left( \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 1 \end{array} \right)$$

$$\left( \begin{array}{ccc|c} 1 & 0 & 3 & -1 \\ 0 & 1 & -4 & 2 \\ 0 & 0 & 0 & 0 \end{array} \right)$$

$$\left( \begin{array}{ccc|c} 1 & -5 & 1 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right)$$

$$\left. \begin{array}{l} 2 \\ 3 \\ 1 \end{array} \right\} \begin{cases} \mathbf{r}_1 \cdot \mathbf{x} = 0 \\ \mathbf{r}_2 \cdot \mathbf{x} = 0 \\ \text{a line through the origin } R^1 \end{cases}$$

$$\left. \begin{array}{l} 1 \\ 3 \\ 2 \end{array} \right\} \begin{cases} \mathbf{r}_1 \cdot \mathbf{x} = 0 \\ \text{a plane through the origin } R^2 \end{cases}$$

$$\begin{cases} x_1 = -1 - 3t \\ x_2 = 2 + 4t \\ x_3 = t \end{cases}$$

$$\begin{cases} x_1 = 4 + 5s - 1t \\ x_2 = s \\ x_3 = t \end{cases}$$

# Consistent Linear System $\mathbf{Ax}=\mathbf{b}$

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \end{pmatrix}$$

$\mathbf{Ax} = \mathbf{b}$  consistent  $\longleftrightarrow$

$$x_1 \mathbf{c}_1 + x_2 \mathbf{c}_2 + \cdots + x_n \mathbf{c}_n = \mathbf{b}$$

expressed in linear combination  
of column vectors

$\longleftrightarrow$   $\mathbf{b}$  is in the column space of  $\mathbf{A}$

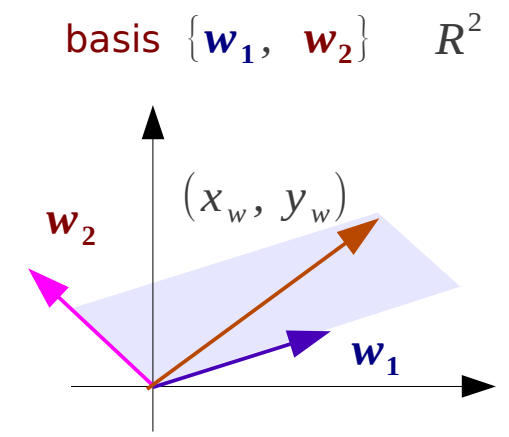
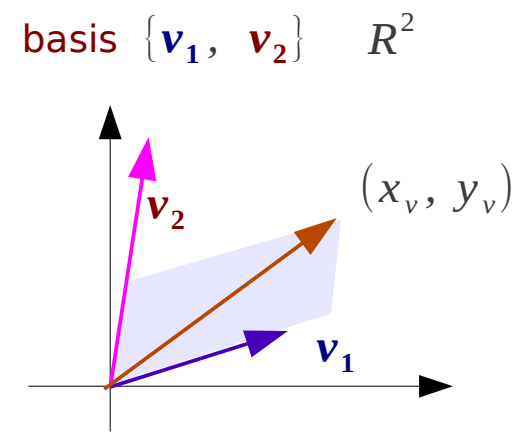
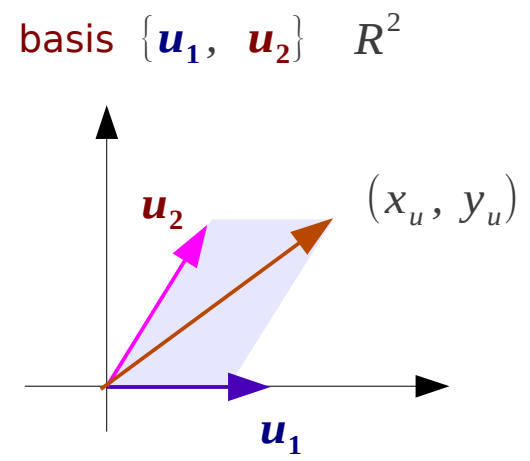
$$= x_1 \begin{pmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{pmatrix} + x_2 \begin{pmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{pmatrix} + \cdots + x_n \begin{pmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{pmatrix}$$

$$\mathbf{Ax} = x_1 \mathbf{c}_1 + x_2 \mathbf{c}_2 + \cdots + x_n \mathbf{c}_n = \mathbf{b}$$

# Dimension

In a **finite-dimensional** vector space  $R^n$   ~~$R^\infty$~~   
all bases  $\rightarrow$  the **same number** of vectors  $n$

many bases but the same number of basis vectors



The **dimension** of a **finite-dimensional** vector space  $V$

$\dim(V)$   $\leftrightarrow$  the **number** of vectors in a **basis**

# Dimension of a Basis (1)

In vector space  $R^2$

any **one** vector (linearly indep.) ~~spans  $R^2$~~  line through 0

basis any **two** non-collinear vectors (linearly indep.) spans  $R^2$  plane ←

any **three or more** vectors ~~(linearly indep.)~~ spans  $R^2$  plane

In vector space  $R^3$

any **one** vector (linearly indep.) ~~spans  $R^3$~~  line through 0

any **two** non-collinear vectors (linearly indep.) ~~spans  $R^3$~~  plane through 0

basis any **three** vectors non-collinear, non-coplanar (linearly indep.) spans  $R^3$  3-dim space ←

any **four or more** vectors ~~(linearly indep.)~~ spans  $R^3$  3-dim space

# Dimension of a Basis (2)

In vector space  $R^n$

any  $n-1$  vectors

(linearly indep.)?

~~spans~~

~~$R^n$~~

line through 0

basis

$n$  vectors of a basis

(linearly indep.)

spans

$R^n$

plane

any  $n+1$  vectors

~~(linearly indep.)~~

spans?

$R^n$

plane

a finite-dimensional vector space  $V$

a basis  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$

- { a set of more than  $n$  vectors  $\rightarrow$  ~~(linearly indep.)~~
- { a set of less than  $n$  vectors  $\rightarrow$  ~~spans  $V$~~

$S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  non-empty finite set of vectors in  $V$

$S$  is a basis



- {  $S$  linearly independent
- {  $S$  spans  $V$



# Basis Test

$S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  non-empty finite set of vectors in  $V$

$S$  is a basis  $\iff$   $\left\{ \begin{array}{l} S \text{ linearly independent} \\ S \text{ spans } V \end{array} \right.$

$V$  an  $n$ -dimensional vector space

$S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  a set of  $n$  vectors in  $V$

$S$  linearly independent  $\implies$   $S$  is a basis

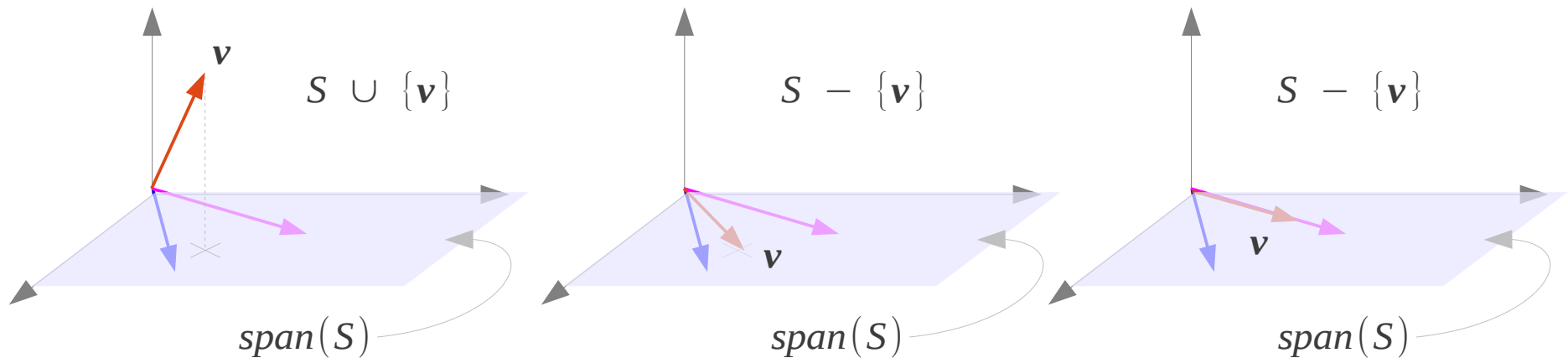
$S$  spans  $V$   $\implies$   $S$  is a basis

# Plus / Minus Theorem

$S$  a nonempty set of vectors in a vector space  $V$

$\left\{ \begin{array}{l} S : \text{linear independent} \\ \mathbf{v} \text{ a vector in } V \text{ but outside of } \text{span}(S) \end{array} \right. \Rightarrow S \cup \{\mathbf{v}\} : \text{linear independent}$

$\left\{ \begin{array}{l} \mathbf{v}, \mathbf{u}_i \in S \text{ linear combination} \\ \mathbf{v} = k_1 \mathbf{u}_1 + k_2 \mathbf{u}_2 + \dots + k_n \mathbf{u}_n \end{array} \right. \Rightarrow \text{span}(S) = \text{span}(S - \{\mathbf{v}\})$



# Finding a Basis

$S$  a nonempty set of vectors in a vector space  $V$

$\left\{ \begin{array}{l} S : \text{linear independent} \\ \mathbf{v} \text{ a vector in } V \text{ but outside of } \text{span}(S) \end{array} \right. \Rightarrow S \cup \{\mathbf{v}\} : \text{linear independent}$

if  $S$  is a *linearly independent* set that is not already a basis for  $V$ ,  
then  $S$  can be enlarged to a basis for  $V$   
by inserting appropriate vectors into  $S$

$\left\{ \begin{array}{l} \mathbf{v}, \mathbf{u}_i \in S \quad \text{linear combination} \\ \mathbf{v} = k_1 \mathbf{u}_1 + k_2 \mathbf{u}_2 + \cdots + k_n \mathbf{u}_n \end{array} \right. \Rightarrow \text{span}(S) = \text{span}(S - \{\mathbf{v}\})$

if  $S$  spans  $V$  but is not a basis for  $V$ ,  
then  $S$  can be reduced to a basis for  $V$   
by removing appropriate vectors from  $S$

# Vectors in a Vector Space

$S$  a nonempty set of vectors in a vector space  $V$

if  $S$  is a *linearly independent* set that is not already a basis for  $V$ ,  
then  $S$  can be enlarged to a basis for  $V$   
by inserting appropriate vectors into  $S$

Every *linearly independent* set in a subspace is  
either a **basis** for that subspace  
or can be **extended to a basis** for it

if  $S$  spans  $V$  but is not a basis for  $V$ ,  
then  $S$  can be reduced to a basis for  $V$   
by removing appropriate vectors from  $S$

Every *spanning set* for a subspace is  
either a **basis** for that subspace  
or has a **basis as a subset**

# Dimension of a Subspace

$W$  a subspace of a finite-dimensional vector space  $V$

$W$  is *finite-dimensional*

$$\dim(W) \leq \dim(V)$$

$$W = V \iff \dim(W) = \dim(V)$$

# Rank and Nullity

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}$$

**ROW Space**      subspace of  $R^n$   
 $= \text{span}\{\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_m\}$

**COLUMN Space**      subspace of  $R^m$   
 $= \text{span}\{\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_n\}$

**NULL Space**      subspace of  $R^n$

**solution space**       $A\mathbf{x} = \mathbf{0}$

**Invertible A**

$$\mathbf{x} = A^{-1}\mathbf{0} = \mathbf{0}$$

only trivial solution

**Non-invertible A**

zero row(s) in a RREF

free variables

parameters  $s, t, u, \dots$

$$\dim(\text{row space of } A) = \dim(\text{column space of } A) = \text{rank}(A)$$

$$\dim(\text{null space of } A) = \text{nullity}(A)$$

# Solution Space of $\mathbf{Ax}=\mathbf{0}$

$$\left[ \begin{array}{ccc|c} 1 & 0 & 0 & 0 \\ 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$$\left[ \begin{array}{ccc|c} 1 & 0 & 3 & 0 \\ 0 & 1 & -4 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$$\left[ \begin{array}{ccc|c} 1 & -5 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

the same case



$$\begin{cases} x_1 = -1 - 3t \\ x_2 = 2 + 4t \\ x_3 = t \end{cases}$$

$$\begin{cases} x_1 = 4 + 5s - 1t \\ x_2 = s \\ x_3 = t \end{cases}$$

General  
Solution of  
 $\mathbf{Ax} = \mathbf{0}$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} -3 \\ 4 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + s \begin{bmatrix} 5 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

dim(row space of A)  
dim(col space of A)

$\text{rank}(A) = 2$

$\text{rank}(A) = 1$

dim(null space of A)

$\text{nullity}(A) = 1$

$\text{nullity}(A) = 2$

# Elementary Row Operation (1)

**ROW Space**          subspace of  $R^n$

$$= \text{span}\{\mathbf{r}_1, \mathbf{r}_2, \dots, \mathbf{r}_m\}$$

**COLUMN Space**      subspace of  $R^m$

$$= \text{span}\{\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_n\}$$

**NULL Space**          subspace of  $R^n$

**solution space**       $A\mathbf{x} = \mathbf{0}$

free variables          parameters  $s, t, u, \dots$

Elementary row operations do not change the **null space** of a matrix

Elementary row operations do not change the **row space** of a matrix

Elementary row operations do change the **col space** of a matrix

Elementary row operations do not change  
the **linear dependence** and **linear independence** relationship  
among column vectors

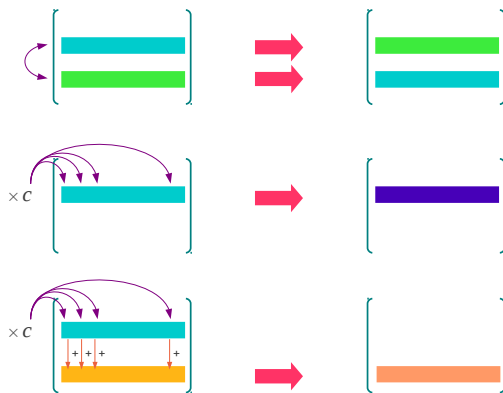
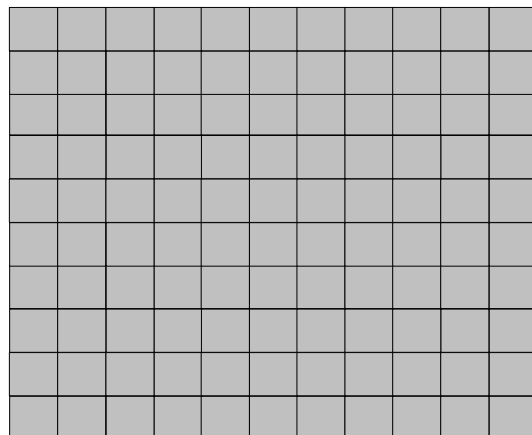


# Elementary Row Operation (2)

Elementary row operations do not change the **null space** of a matrix  
 Elementary row operations do not change the **row space** of a matrix  
 Elementary row operations do not change the **linear dependence** and **linear independence** relationship among column vectors

Elementary row operations do change the **col space** of a matrix

**A**



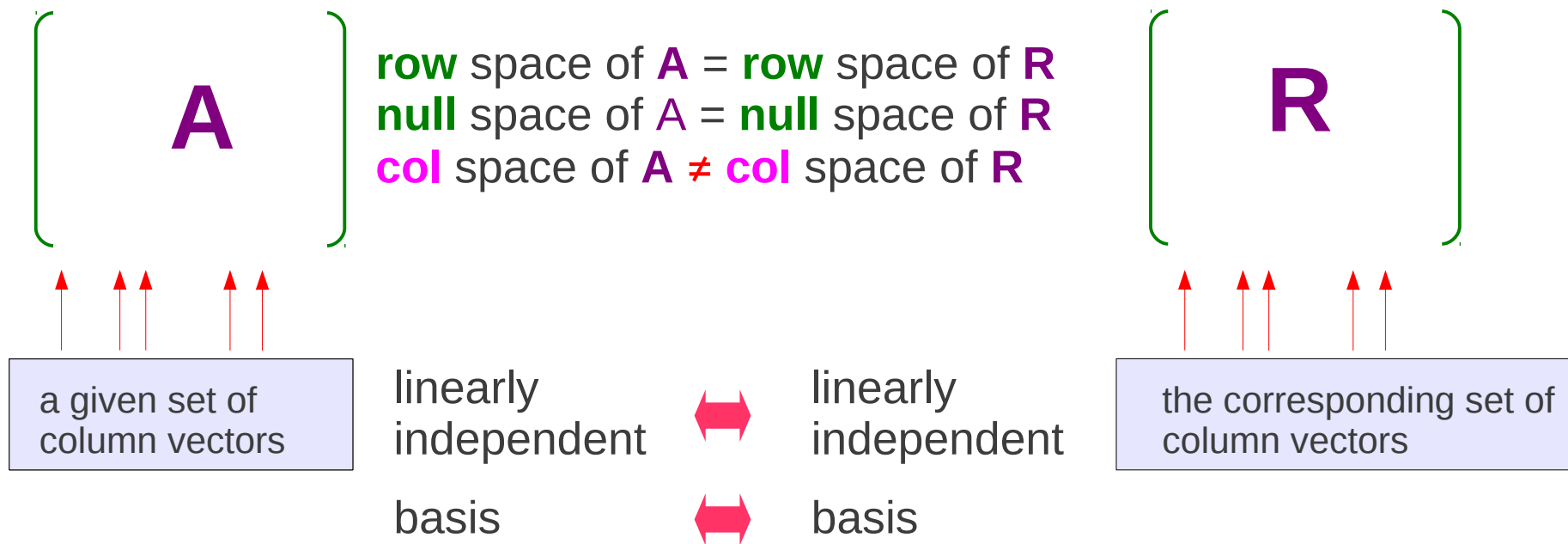
**R**

1	0	0	0	0	0	0	0	0	0
0	1	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	0	0
0	0	0	0	0	1	0	0	0	0
0	0	0	0	0	0	1	0	0	0
0	0	0	0	0	0	0	1	0	0
0	0	0	0	0	0	0	0	0	1
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

# Elementary Row Operation (3)

## Elementary row operations

- do not change the **null space** of a matrix
- do not change the **row space** of a matrix
- do not change the **linear dependence** and **linear independence** relationship among column vectors
- do change the **col space** of a matrix



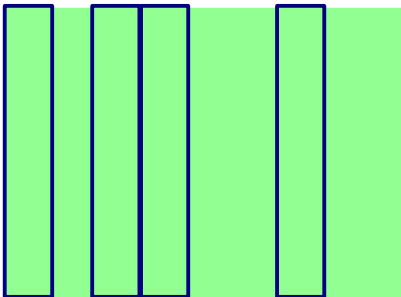
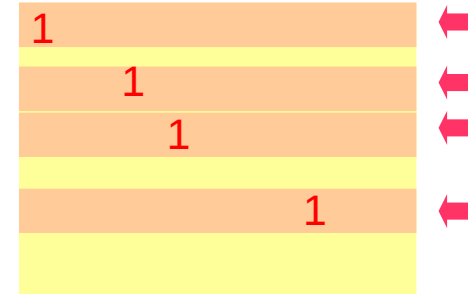
# Bases of Row & Column Spaces (1)



basis of  
row space  
of **A**

=

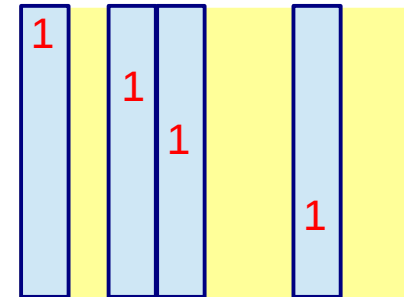
basis of  
row space  
of **R**



basis of  
col space  
of **A**

≠

basis of  
col space  
of **R**



the corresponding set of  
column vectors

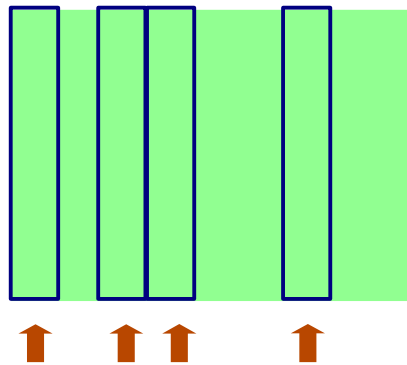


a given set of  
column vectors



$$\dim(\text{row space of } A) = \dim(\text{column space of } A) = \text{rank}(A)$$

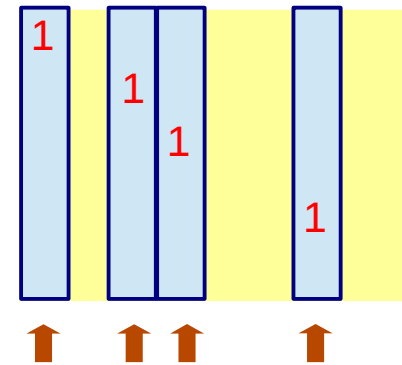
# Bases of Row & Column Spaces (2)



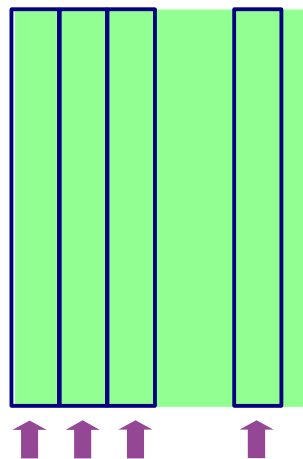
basis of  
col space  
of **A**

$\neq$

basis of  
col space  
of **R**



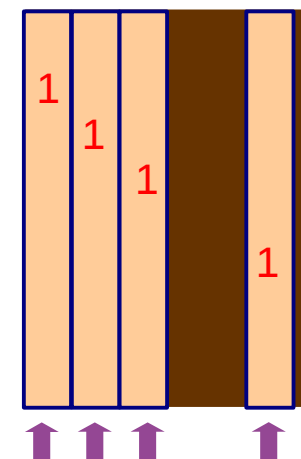
the basis consisting of **columns** of **A**



basis of  
col space  
of **A**

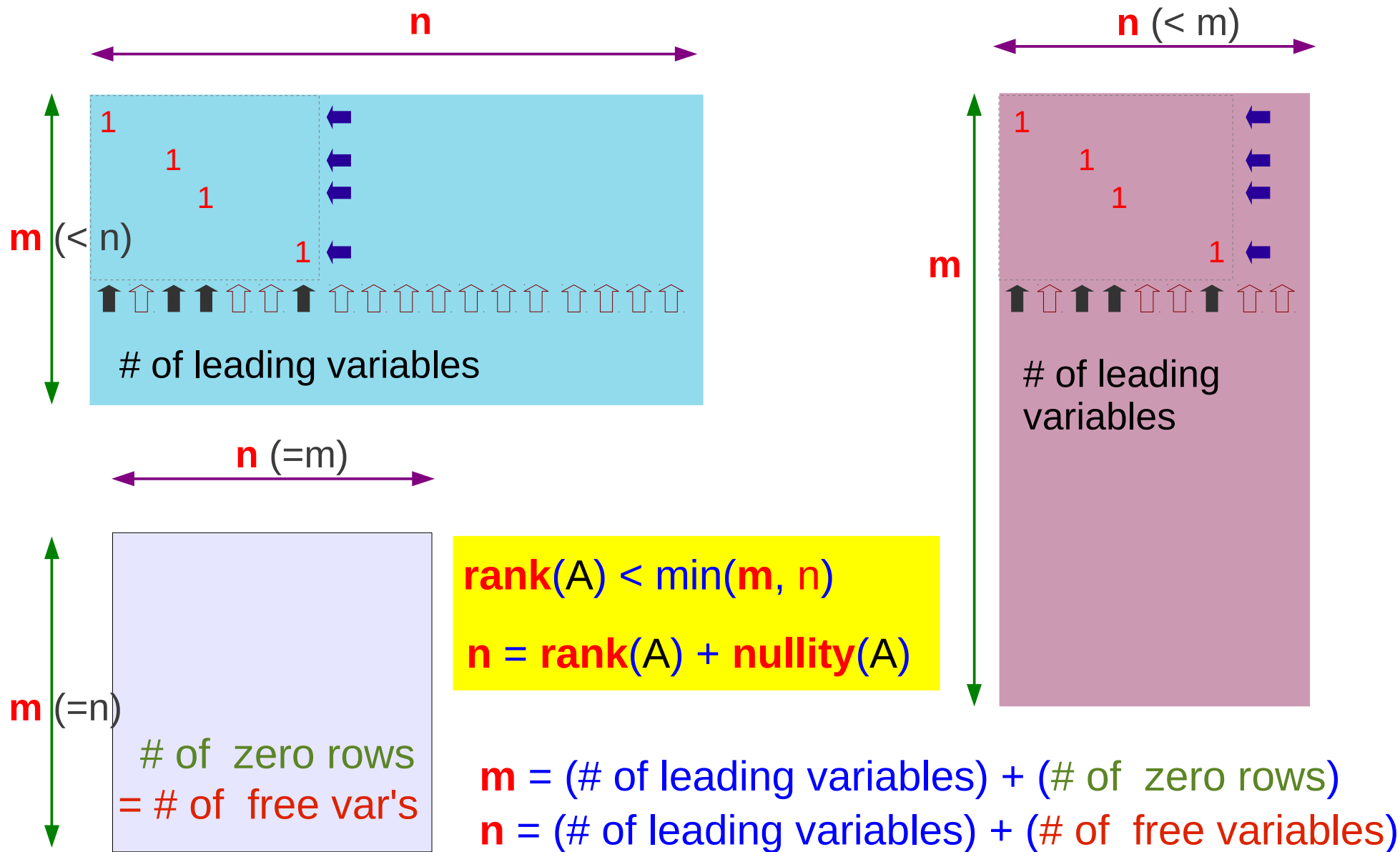
$\neq$

basis of  
col space  
of **R**

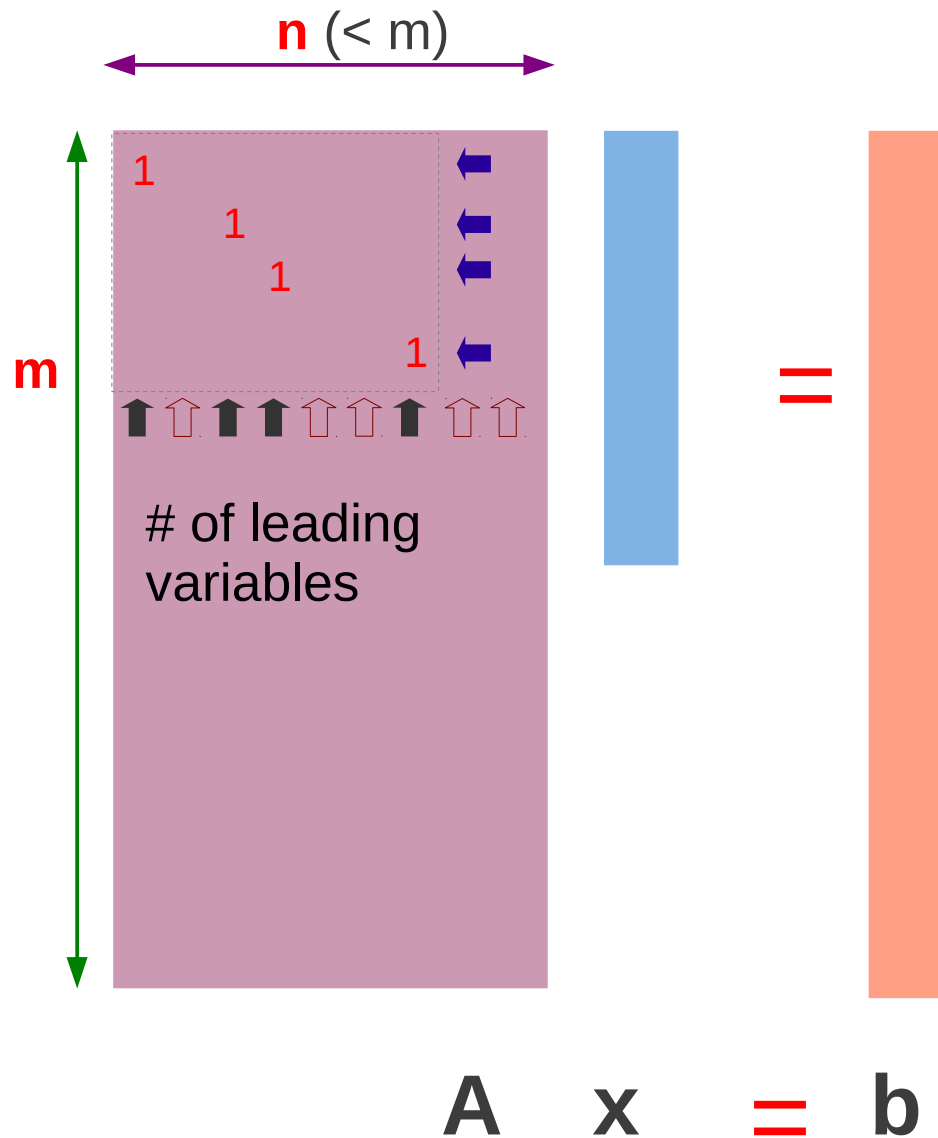


the basis consisting of **rows** of **A**

# Rank and Nullity



# Overdetermined System



$n$  column vectors  
can span at most  $\mathbb{R}^n$

$b$  is in  $\mathbb{R}^m$        $\mathbb{R}^m \supset \mathbb{R}^n$

At least one vector  $b$  in  $\mathbb{R}^m$   
does not lie in column space

For such  $b$  in  $\mathbb{R}^m$   
 $Ab = b$  inconsistent

## References

- [1] <http://en.wikipedia.org/>
- [2] Anton, et al., Elementary Linear Algebra, 10<sup>th</sup> ed, Wiley, 2011
- [3] Anton, et al., Contemporary Linear Algebra,