

# Lesson 2

## Chapter 1

### Linear equations in Linear Algebra

► Systems of Linear Equations

► Row Reduction and Echelon Forms

► Vector Equations

► The Matrix Equation  $Ax = b$

► Solution Sets of Linear Systems

► Linear Independence

## System of linear equations:

$$\begin{aligned}a_{11} \cdot x_1 + a_{12} \cdot x_2 + a_{13} \cdot x_3 + \cdots + a_{1n} \cdot x_n &= b_1 \\a_{21} \cdot x_1 + a_{22} \cdot x_2 + a_{23} \cdot x_3 + \cdots + a_{2n} \cdot x_n &= b_2 \\a_{31} \cdot x_1 + a_{32} \cdot x_2 + a_{33} \cdot x_3 + \cdots + a_{3n} \cdot x_n &= b_3 \\&\vdots \\&\vdots \\a_{m1} \cdot x_1 + a_{m2} \cdot x_2 + a_{m3} \cdot x_3 + \cdots + a_{mn} \cdot x_n &= b_m\end{aligned}$$

Augmented matrix:

$$\left[ \begin{array}{cccc|c} a_{11} & a_{12} & \cdots & a_{1n} & b_1 \\ a_{21} & a_{22} & \cdots & a_{2n} & b_2 \\ \vdots & \vdots & \cdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} & b_m \end{array} \right]$$

Row reduction

Row equivalent

Reduced echelon form:

$$\left[ \begin{array}{ccccc} \boxed{1} & 0 & * & 0 & 0 \\ 0 & \boxed{1} & * & 0 & 0 \\ 0 & 0 & 0 & \boxed{1} & 0 \\ 0 & 0 & 0 & 0 & \boxed{1} \end{array} \right]$$

Pivot / Pivot columns

No / One /  $\infty$  solutions

## Vectors

A vector equation:  $c_1 \cdot \mathbf{v}_1 + c_2 \cdot \mathbf{v}_2 + \cdots + c_p \cdot \mathbf{v}_p = \mathbf{b}$

→ have the same solution as the linear system with augmented matrix

$$[\mathbf{v}_1 \quad \mathbf{v}_2 \quad \cdots \quad \mathbf{v}_p \quad \mathbf{b}]$$

$$\text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \cdots, \mathbf{v}_p\} = c_1 \cdot \mathbf{v}_1 + c_2 \cdot \mathbf{v}_2 + \cdots + c_p \cdot \mathbf{v}_p$$

← Linear  
Combination

## 1.4 The Matrix Equation $A\mathbf{x} = \mathbf{b}$

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

Ex 1

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

## Definition

If  $A$  is a  $m \times n$  matrix, with columns  $\mathbf{a}_1, \dots, \mathbf{a}_n$ , and if  $\mathbf{x}$  is in  $\mathbb{R}^n$ , then the **product of  $A$  and  $\mathbf{x}$**  denoted by  $A\mathbf{x}$  is the **linear combination of the columns of  $A$  using the corresponding entries in  $\mathbf{x}$  as weights**; that is,

$$A\mathbf{x} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \dots & \mathbf{a}_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \dots + x_n\mathbf{a}_n \quad \leftarrow \text{Vector in } \mathbb{R}^m$$

Note:

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & & a_{2n} \\ \vdots & & & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \dots & \mathbf{a}_n \end{bmatrix} \quad \text{and} \quad A\mathbf{x} = \begin{bmatrix} a_{11} \cdot x_1 + a_{12} \cdot x_2 + a_{13} \cdot x_3 + \dots + a_{1n} \cdot x_n \\ a_{21} \cdot x_1 + a_{22} \cdot x_2 + a_{23} \cdot x_3 + \dots + a_{2n} \cdot x_n \\ \vdots \\ a_{m1} \cdot x_1 + a_{m2} \cdot x_2 + a_{m3} \cdot x_3 + \dots + a_{mn} \cdot x_n \end{bmatrix} \in \mathbb{R}^m$$

Matrix equation:  $A\mathbf{x} = \mathbf{b}$

Let a matrix and vector be given by

$$A = \begin{bmatrix} 2 & 8 \\ 1 & 3 \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$$

Solve the matrix equation:  $A\mathbf{x} = \mathbf{b}$

That is, find  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  so  $A\mathbf{x} = \mathbf{b}$

Ex 2      $A = \begin{bmatrix} 2 & 8 \\ 1 & 3 \end{bmatrix}; \quad \boldsymbol{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}; \quad \boldsymbol{b} = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$



## Theorem 1.3

If  $A$  is an  $m \times n$  matrix with columns  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ , and if  $\mathbf{b}$  is in  $\mathbb{R}^m$  the matrix equation

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

has the same solution set as the vector equation

$$x_1 \cdot \mathbf{a}_1 + x_2 \cdot \mathbf{a}_2 + \dots + x_n \cdot \mathbf{a}_n = \mathbf{b}$$

which, in turn, has the same solution set as the system of linear equations

$$\left\{ \begin{array}{l} a_{11} \cdot x_1 + a_{12} \cdot x_2 + a_{13} \cdot x_3 + \dots + a_{1n} \cdot x_n = b_1 \\ a_{21} \cdot x_1 + a_{22} \cdot x_2 + a_{23} \cdot x_3 + \dots + a_{2n} \cdot x_n = b_2 \\ \vdots \\ a_{m1} \cdot x_1 + a_{m2} \cdot x_2 + a_{m3} \cdot x_3 + \dots + a_{mn} \cdot x_n = b_m \end{array} \right\}$$

whose augmented matrix is

$$[\mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_n \quad \mathbf{b}]$$

**OBS:**

#Columns in  $A$  = #Rows in  $\mathbf{x}$

#Rows in  $A$  = #Rows in  $\mathbf{b}$

Note:

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

→  $\mathbf{b}$  is a linear combination of the columns of  $A$

→  $\mathbf{b} \in \text{Span}\{\mathbf{a}_1, \mathbf{a}_2, \cdots, \mathbf{a}_n\}$

→  $A\mathbf{x} = \mathbf{b}$  has solutions if and only if  $\mathbf{b} \in \text{Span}\{\mathbf{a}_1, \mathbf{a}_2, \cdots, \mathbf{a}_n\}$

Ex 3

$$A = \begin{bmatrix} 1 & 3 & 4 \\ -4 & 2 & -6 \\ -3 & -2 & -7 \end{bmatrix} = [\mathbf{a}_1 \ \mathbf{a}_2 \ \mathbf{a}_3];$$

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

Har  $A\mathbf{x} = \mathbf{b}$  løsninger for alle  $\mathbf{b}$ ?

## Theorem 1.4

Let  $A$  be an  $m \times n$  matrix. Then the following statements are logically equivalent. That is, for a particular  $A$ , either they are all true statements or they are all false.

1. For each  $\mathbf{b}$  in  $\mathbb{R}^m$ , the equation  $A\mathbf{x} = \mathbf{b}$  has a solution.
2. Each  $\mathbf{b}$  in  $\mathbb{R}^m$  is a linear combination of the columns of  $A$ .
3. The columns in  $A$  span  $\mathbb{R}^m$ .
4.  $A$  has a pivot position in every row.

## Theorem 1.5

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

$$A(c\mathbf{u}) = c(A\mathbf{u})$$

Ex.

$$\begin{bmatrix} 2 & -3 \\ -4 & 1 \end{bmatrix} \left( \begin{bmatrix} 2 \\ -1 \end{bmatrix} + \begin{bmatrix} -3 \\ 0 \end{bmatrix} \right) = \begin{bmatrix} 2 & -3 \\ -4 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \end{bmatrix} + \begin{bmatrix} 2 & -3 \\ -4 & 1 \end{bmatrix} \begin{bmatrix} -3 \\ 0 \end{bmatrix} = \begin{bmatrix} 7 \\ -9 \end{bmatrix} + \begin{bmatrix} -6 \\ 12 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

$$\begin{bmatrix} 5 & 3 \\ 0 & 1 \end{bmatrix} \left( -3 \begin{bmatrix} 4 \\ 1 \end{bmatrix} \right) = -3 \begin{bmatrix} 5 & 3 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 4 \\ 1 \end{bmatrix} = -3 \begin{bmatrix} 23 \\ 1 \end{bmatrix} = \begin{bmatrix} -69 \\ -3 \end{bmatrix}$$

# 1.5 Solution Sets of Linear Systems

Homogeneous Linear System:  $A\mathbf{x} = \mathbf{0}$

Inhomogeneous Linear System:  $A\mathbf{x} = \mathbf{b} \neq \mathbf{0}$

## Theorem 1.2: Existence and Uniqueness Theorem

A linear system is **consistent** if and only if **the rightmost column of the augmented matrix is not a pivot column** – that is, if and only if an echelon form of the augmented matrix has not rows of the form:

$$[0 \quad \dots \quad 0 \quad b] \text{ with } b \neq 0$$

If a linear system is consistent, then the solution set contains either:

- i. **A unique solution**, when there are no free variables – that is, **all except the last column are pivot columns**.
- ii. **Infinitely many solutions**, when there is at least one free variable – that is, **at least one column besides the last one is not a pivot column**.

Homogeneous matrix equation:  $A\mathbf{x} = \mathbf{0} = \begin{bmatrix} \mathbf{0} \\ \vdots \\ \mathbf{0} \end{bmatrix}$

→ Always the *trivial* solution:  $\mathbf{x} = \mathbf{0}$

→ *Nontrivial* solution ( $\mathbf{x} \neq \mathbf{0}$ ) if and only if at least one free variable

→  $\infty$  many solutions



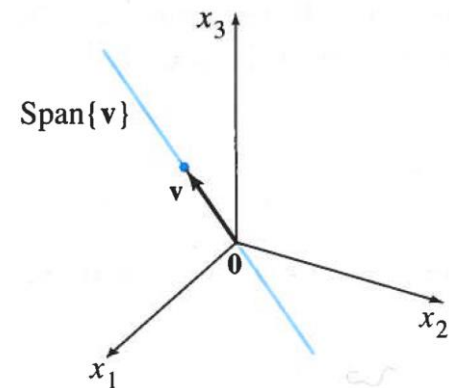
## Ex 4: Homogeneous equation $A\mathbf{x} = \mathbf{0}$

$$\left\{ \begin{array}{l} 3x_1 + 5x_2 - 4x_3 = 0 \\ -3x_1 - 2x_2 + 4x_3 = 0 \\ 6x_1 + x_2 - 8x_3 = 0 \end{array} \right\} \Leftrightarrow \begin{bmatrix} 3 & 5 & -4 \\ -3 & -2 & 4 \\ 6 & 1 & -8 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = \mathbf{0}$$

$$\begin{bmatrix} 3 & 5 & -4 & 0 \\ -3 & -2 & 4 & 0 \\ 6 & 1 & -8 & 0 \end{bmatrix} \sim \begin{bmatrix} 3 & 5 & -4 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & -9 & 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 3 & 0 & -4 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & -4/3 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\Rightarrow x_1 - \frac{4}{3}x_3 = 0; \quad x_2 = 0; \quad x_3 = x_3$$

$$\rightarrow \mathbf{x} = \begin{bmatrix} \frac{4}{3}x_3 \\ 0 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} \frac{4}{3} \\ 0 \\ 1 \end{bmatrix} = x_3 \mathbf{v} \rightarrow \text{Straight line in } \mathbb{R}^3$$



## Ex 5: Inhomogeneous equation $A\mathbf{x} = \mathbf{b}$

$$\left\{ \begin{array}{l} 3x_1 + 5x_2 - 4x_3 = 7 \\ -3x_1 - 2x_2 + 4x_3 = -1 \\ 6x_1 + x_2 - 8x_3 = -4 \end{array} \right\} \Leftrightarrow \begin{bmatrix} 3 & 5 & -4 \\ -3 & -2 & 4 \\ 6 & 1 & -8 \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 7 \\ -1 \\ -4 \end{bmatrix}$$

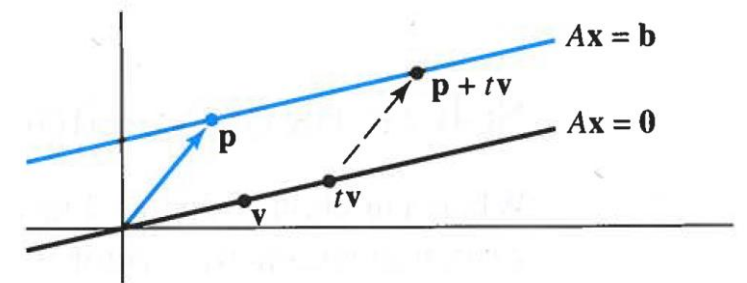
$$\begin{bmatrix} 3 & 5 & -4 & 7 \\ -3 & -2 & 4 & -1 \\ 6 & 1 & -8 & -4 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & -4/3 & -1 \\ 0 & 1 & 0 & 2 \\ 0 & 0 & 0 & 0 \end{bmatrix} \Rightarrow \begin{cases} x_1 - \frac{4}{3}x_3 = -1 \\ x_2 = 2 \\ x_3 = x_3 = t \end{cases} \Rightarrow \begin{cases} x_1 = -1 + \frac{4}{3} \cdot t \\ x_2 = 2 + 0 \cdot t \\ x_3 = 0 + 1 \cdot t \end{cases}$$

Parametric vector equation

$$\Rightarrow \mathbf{x} = \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix} + t \begin{bmatrix} \frac{4}{3} \\ 0 \\ 1 \end{bmatrix} = \mathbf{p} + \mathbf{v}_h \rightarrow \text{Translated straight line in } \mathbb{R}^3$$

Inhomogeneous  
solution

Homogeneous  
solution

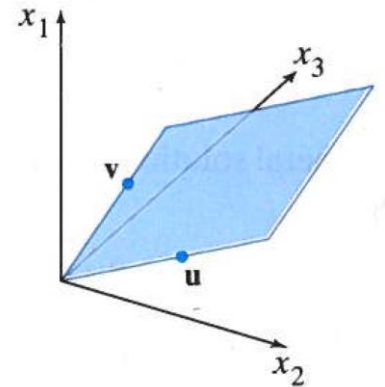


Ex 6: Homogeneous equation  $A\mathbf{x} = \mathbf{0}$

$$\{10x_1 - 3x_2 - 2x_3 = 0\} \Leftrightarrow [10 \quad -3 \quad -2] \cdot \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} = \mathbf{0}$$

$$\Rightarrow 10x_1 = 3x_2 + 2x_3 \Rightarrow x_1 = \frac{3}{10}x_2 + \frac{2}{10}x_3; \quad x_2, x_3 \text{ free}$$

$$\Rightarrow \mathbf{x} = \begin{bmatrix} \frac{3}{10}x_2 + \frac{2}{10}x_3 \\ x_2 \\ x_3 \end{bmatrix} = x_2 \begin{bmatrix} \frac{3}{10} \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} \frac{2}{10} \\ 0 \\ 1 \end{bmatrix} = x_2 \mathbf{u} + x_3 \mathbf{v} \rightarrow \text{Plane in } \mathbb{R}^3$$

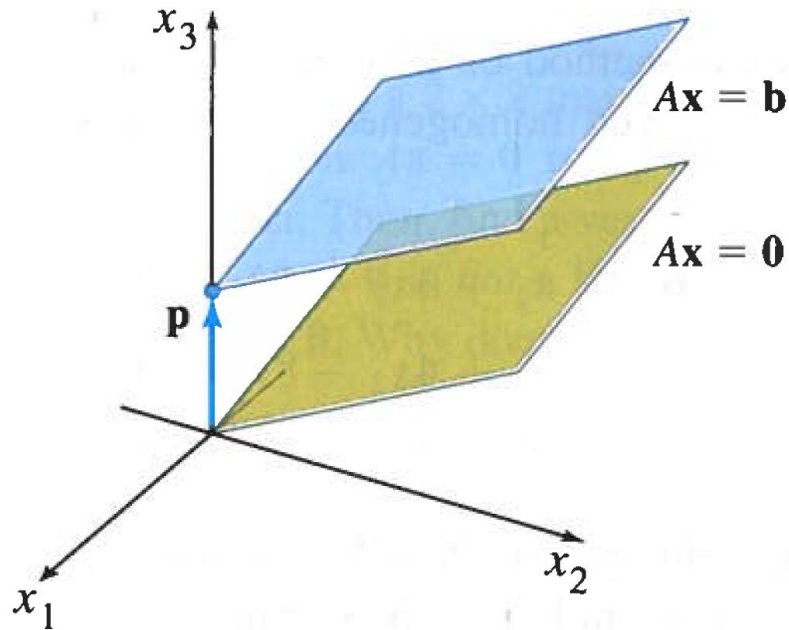


→ Solutions to  $A\mathbf{x} = \mathbf{0}$ :  $\mathbf{x} \in \text{Span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_j\}$  if  $j$  free variables

↖ Null Space

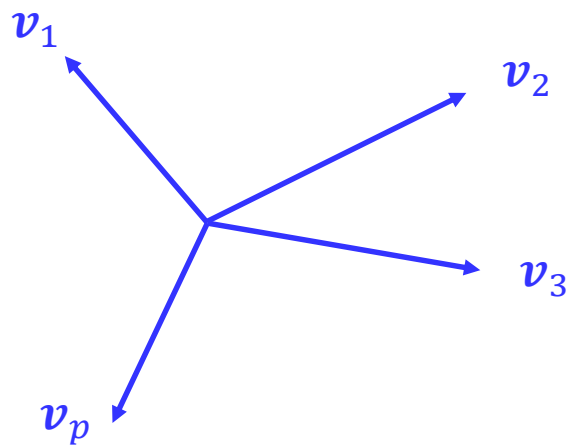
## Theorem 1.6

Let  $\mathbf{p}$  be a solution to the consistent inhomogeneous matrix equation  $A\mathbf{x} = \mathbf{b}$ . Then the set of all solutions to  $A\mathbf{x} = \mathbf{b}$  is the set of vectors  $\mathbf{w} = \mathbf{p} + \mathbf{v}_h$ , where  $\mathbf{v}_h$  is the solution set of the homogeneous equation  $A\mathbf{x} = \mathbf{0}$ .



# 1.7 Linear Independence

$$c_1 \cdot \mathbf{v}_1 + c_2 \cdot \mathbf{v}_2 + \cdots + c_p \cdot \mathbf{v}_p = \mathbf{0}$$



**Homogeneous** matrix equation:  $A\mathbf{x} = \mathbf{0} = \begin{bmatrix} \mathbf{0} \\ \vdots \\ \mathbf{0} \end{bmatrix}$

→ Always the **trivial** solution:  $\mathbf{x} = \mathbf{0}$

→ **Nontrivial** solution ( $\mathbf{x} \neq \mathbf{0}$ ) if and only if at least one free variable  
→  $\infty$  many solutions

When are the trivial solution the only solution?

Let  $A = [\mathbf{v}_1 \quad \mathbf{v}_2 \quad \cdots \quad \mathbf{v}_p]$ :

$$A\mathbf{x} = x_1 \cdot \mathbf{v}_1 + x_2 \cdot \mathbf{v}_2 + \cdots + x_p \cdot \mathbf{v}_p = \mathbf{0} \stackrel{?}{\Leftrightarrow} x_1 = x_2 = \cdots = x_p = 0$$

## Definition

An indexed set of vectors  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  in  $\mathbb{R}^n$  is said to be **linearly independent** if the vector equation

$$x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \dots + x_p \mathbf{v}_p = \mathbf{0}$$

has only the trivial solution. The set  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  is said to be **linearly dependent** if there exist weights  $c_1, c_2, \dots, c_p$ , not all zero, such that

$$c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + \dots + c_p \mathbf{v}_p = \mathbf{0}$$

**Homogeneous** matrix equation:  $A\mathbf{x} = \mathbf{0} = \begin{bmatrix} \mathbf{0} \\ \vdots \\ \mathbf{0} \end{bmatrix}$

→ Always the **trivial** solution:  $\mathbf{x} = \mathbf{0}$

→ **Nontrivial** solution ( $\mathbf{x} \neq \mathbf{0}$ ) if and only if at least one free variable  
→  $\infty$  many solutions

When are the trivial solution the only solution?

$A = [\mathbf{v}_1 \quad \mathbf{v}_2 \quad \cdots \quad \mathbf{v}_p]$ :

$A\mathbf{x} = x_1 \cdot \mathbf{v}_1 + x_2 \cdot \mathbf{v}_2 + \cdots + x_p \cdot \mathbf{v}_p = \mathbf{0} \rightarrow$  Only the trivial solution

$\Leftrightarrow$  The columns of  $A$  are linearly independent



# Linearly independence

a)  $\mathbf{v} \neq \mathbf{0} \rightarrow$  Linearly independent

b)  $\mathbf{v}_1 \neq c \cdot \mathbf{v}_2 \rightarrow$  Linearly independent

c)  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p \in \mathbb{R}^n; p > n \rightarrow$  Linearly dependent

$[\mathbf{v}_1 \ \mathbf{v}_2 \ \dots \ \mathbf{v}_p \ \mathbf{0}] \rightarrow$  Max.  $n$  pivot (one in each row)  $\rightarrow$  Min. 1 column not a Pivot column

$\rightarrow$  At least one free parameter

$\rightarrow c_1 \cdot \mathbf{v}_1 + c_2 \cdot \mathbf{v}_2 + \dots + c_p \cdot \mathbf{v}_p = \mathbf{0}$  has infinitely many (nontrivial) solutions

## Theorem 1.7

An indexed set  $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  of two or more vectors is linearly dependent if and only if at least one of the vectors in  $S$  is a linear combination of the others. That is the set is linearly dependent if

$$\mathbf{v}_j = c_1 \cdot \mathbf{v}_1 + c_2 \cdot \mathbf{v}_2 + \dots + c_{j-1} \cdot \mathbf{v}_{j-1}$$

for some  $1 < j \leq p$  and  $\mathbf{v}_1 \neq \mathbf{0}$

## Theorem 1.8

If a set contains more vectors than there are entries (rows) in each vector, then the set is linearly dependent. That is, any set  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  in  $\mathbb{R}^n$  is linearly dependent if  $p > n$ .

## Theorem 1.9

If a set  $S = \{\boldsymbol{v}_1, \boldsymbol{v}_2, \dots, \boldsymbol{v}_p\}$  in  $\mathbb{R}^n$  contains the zero vector, then the set is linearly dependent.

# Today's words and concepts

Matrix equation

Homogeneous linear system

Linearly dependent

Multiplication properties

Trivial solution

Non-trivial solution

Inhomogeneous linear system

Linear combination

Linearly independent

Parametric vector equation

Matrix-vector multiplication