# MATH2111 Tutorial 3

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#### 1 The Matrix Equation

1. Matrix-Vector Products. We can multiply an  $m \times n$  matrix A by a vector  $v \in \mathbb{R}^n$ . The result, written Av, belongs to  $\mathbb{R}^m$ . If  $a_1, a_2, \dots, a_n \in \mathbb{R}^m$  are the columns of A and  $v_1, v_2, \dots, v_n \in \mathbb{R}$  are the entries of v, then

$$\mathbf{A}\mathbf{v} = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \cdots & \mathbf{a}_n \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} = v_1\mathbf{a}_1 + v_2\mathbf{a}_2 + \cdots + v_n\mathbf{a}_n$$

- 2. **Property of Matrix-Vector Products**. If  $A \in \mathbb{R}^{m \times n}$ ,  $u, v \in \mathbb{R}^n$  and  $c \in \mathbb{R}$ , then:
  - (a)  $A(u+v) = Au + Av \in \mathbb{R}^m$
  - (b)  $A(cv) = c(Av) \in \mathbb{R}^m$
- 3. **Theorem.** If A is an  $m \times n$  matrix, with columns  $a_1, \ldots, a_n$ , and if b is in  $\mathbb{R}^m$ , the matrix equation

$$Ax = b$$

has the same solution set as the vector equation

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

which, in turn, has the same solution set as the system of linear equations whose augmented matrix is

$$\begin{bmatrix} a_1 & a_2 & \cdots & a_n \mid b \end{bmatrix}$$

Existence of 4. Theorem. Let A be an  $m \times n$  matrix. Then the following statements are logically equivalent:

solutions to linear systems:

- (a) For each b in  $\mathbb{R}^m$ , the equation Ax = b has a solution.
- (b) Each b in  $\mathbb{R}^m$  is a linear combination of the columns of A.
- (c) The columns of A span  $\mathbb{R}^m$ . i.e. Span  $\{a_1, a_2, \dots, a_n\} = \mathbb{R}^m$
- (d) A has a pivot position in every row.

Warning: The above theorem is about a coefficient matrix, not an augmented matrix. If an augmented matrix  $[A \mid b]$  has a pivot position in every row, then the equation Ax = b may or may not be consistent.

**Example 1.1.** Could a set of *n* vectors in  $\mathbb{R}^m$  span all of  $\mathbb{R}^m$  if n < m? Explain. Can't

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 vectors in  $\mathbb{R}^m$  span all of  $\mathbb{R}^m$  if  $n < m$ ? Explain. Can't  $\{V_1, V_2, \dots, V_n\}$ ,  $V_i \in \mathbb{R}^m$ 

$$V = \begin{bmatrix} V_1 & V_2 & \dots & V_n \end{bmatrix} \in \mathbb{R}^{m \times n}$$

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$$V = \begin{bmatrix} V_1 & V_2$$

# 2 Solution Sets of Linear Systems

1. **Homogeneous Linear Systems**. A system of linear equations is said to be **homogeneous** if it can be written in the form

$$Ax = 0$$

where **A** is an  $m \times n$  matrix and **0** is the zero vector in  $\mathbb{R}^m$ .

### Note:

The system Ax = 0 always has at least one solution, namely, x = 0 (the zero vector in  $\mathbb{R}^n$ ), and (a) this zero solution is called the **trivial solution**.

- (b) the other non-zero solution are called the **nontrivial solution**.
- 2. **Theorem**. The homogeneous equation Ax = 0 has a nontrivial solution if and only if the equation has at least one free variable.
- 3. **Theorem**. Suppose A has k free columns, then the homogeneous equation Ax = 0 has k free variables, and the general solution can be written as **parametric vector form**

$$x = s_1x_1 + s_2x_2 + \ldots + s_kx_k$$

In other words, the solution set of the homogeneous system is

Span 
$$\{x_1, \ldots, x_k\}$$
 Since Span  $\{\vec{o}\} = \{\vec{o}\}$ 

Note: If there are no non-pivot columns (i.e. no free variables), the solution set is just {0}. (because, ⅓, o = o ∨ x)

- 4. Non-Homogeneous Linear Systems. Suppose the equation Ax = b is consistent for some given b  $(b \ne 0)$ , and let p be a particular solution. Then the solution set of Ax = b is the set of all vectors of the form  $w = p + v_h$ , where  $v_h$  is any solution of the homogeneous equation Ax = 0.
- 5. Procedures of Writing a Solution Set (of a consistent system) in Parametric Vector Form
  - (a) Row reduce the augmented matrix to reduced row echelon form (RREF).
  - (b) Express each basic variable in terms of any free variables appearing in an equation.
  - (c) Write a typical solution x as a vector whose entries depend on the free variables, if any.
  - (d) Decompose x into a linear combination of vectors (with numeric entries) using the free variables as parameters.

#### 3 **Exercises**

1. Write the matrix equation as a vector equation, or vice versa.

1. Write the matrix equation as a vector equation (a) 
$$\begin{bmatrix} 5 & 1 & -8 & 4 \\ -2 & -7 & 3 & -5 \end{bmatrix} \begin{bmatrix} 5 \\ -1 \\ 3 \\ -2 \end{bmatrix} = \begin{bmatrix} -8 \\ 16 \end{bmatrix}$$

(b) 
$$x_1 \begin{bmatrix} 4 \\ -1 \\ 7 \\ -4 \end{bmatrix} + x_2 \begin{bmatrix} -5 \\ 3 \\ -5 \\ 1 \end{bmatrix} + x_3 \begin{bmatrix} 7 \\ -8 \\ 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 6 \\ -8 \\ 0 \\ -7 \end{bmatrix}$$

2. Suppose A is a  $3 \times 3$  matrix and b is a vector in  $\mathbb{R}^3$  with the property that Ax = b has a unique solution. Explain why the columns of A must span  $\mathbb{R}^3$ .

1. (a) 
$$A\vec{x} = \vec{b}$$
,  $[\vec{a}_1 \ \vec{a}_2 \ \vec{a}_3 \ \vec{a}_4] \begin{bmatrix} x_1 \\ x_2 \\ x_4 \end{bmatrix} = \vec{b}$ 

$$5 \cdot \begin{bmatrix} 5 \\ -2 \end{bmatrix} - 1 \cdot \begin{bmatrix} -1 \\ -7 \end{bmatrix} + 3 \cdot \begin{bmatrix} -8 \\ 3 \end{bmatrix} - 2 \cdot \begin{bmatrix} 4 \\ -5 \end{bmatrix} = \begin{bmatrix} -8 \\ 16 \end{bmatrix}$$

(b) 
$$\begin{bmatrix} 4 & -5 & 7 \\ -1 & 3 & -8 \\ 7 & -5 & 0 \\ -4 & 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b \\ -8 \\ 0 \\ -7 \end{bmatrix}$$

 $A\vec{x} = \vec{b}$  has a unique solution,

Since  $A \in \mathbb{R}^{3 \times 3}$ , A has a pivot position for each row,

By theorem (A), columns of A span Rs.

3. Determine if the columns of the matrix span  $\mathbb{R}^4$ 

3.

$$\begin{bmatrix} 5 & -7 & -4 & 9 \\ 6 & -8 & -7 & 5 \\ 4 & -4 & -9 & -9 \\ -9 & 11 & 16 & 7 \end{bmatrix} \xrightarrow{R_2 - \frac{b}{5}R_1 \to R_2} \xrightarrow{R_2} \begin{bmatrix} 5 & -7 & -4 & 9 \\ 0 & \frac{2}{5} & -\frac{11}{5} & -\frac{29}{5} \\ 0 & \frac{2}{5} & -\frac{11}{5} & -\frac{29}{5} \\ 0 & \frac{8}{5} & -\frac{29}{5} & -\frac{81}{5} \\ 0 & -\frac{8}{5} & \frac{44}{5} & \frac{116}{5} \end{bmatrix}$$

$$\frac{R_{3}-4R_{2} \Rightarrow R_{3}}{R_{4}+4R_{2} \Rightarrow R_{4}} \Rightarrow
\begin{bmatrix}
5 & -7 & -4 & 9 \\
0 & \frac{2}{5} & -\frac{11}{5} & -\frac{29}{5} \\
0 & 0 & 3 & 7 \\
0 & 0 & 0 & 0
\end{bmatrix}$$

A only has 3 pivot columns, by theorem (\$),

colums of A don't span R4.

4. Determine if the system has a nontrivial solution.

(1) 
$$\begin{cases} 2x_1 - 5x_2 + 8x_3 = 0 \\ -2x_1 - 7x_2 + x_3 = 0 \\ 4x_1 + 2x_2 + 7x_3 = 0 \end{cases}$$
(2) 
$$\begin{cases} x_1 - 3x_2 + 7x_3 = 0 \\ -2x_1 + x_2 - 4x_3 = 0 \\ x_1 + 2x_2 + 9x_3 = 0 \end{cases}$$

(1) 
$$\begin{bmatrix} 2 & -5 & 8 & 0 \\ -2 & -7 & 1 & 0 \\ 4 & 2 & 7 & 0 \end{bmatrix}$$
  $R_2 + R_1 \rightarrow R_2$   $R_3 - 2R_1 \rightarrow R_3$   $\begin{bmatrix} 2 & -5 & 8 & 0 \\ 0 & -12 & 8 & 0 \\ 0 & 12 & -9 & 0 \end{bmatrix}$ 

73 is free, (X1, x2 are basic variables), the system has a nontrivial solution.

(2) 
$$\begin{bmatrix} 1 & -3 & 7 & | & 0 \\ -2 & | & -4 & | & 0 \\ | & 2 & 9 & | & 0 \end{bmatrix} \xrightarrow{R_2+2R_1 \to R_2} \begin{bmatrix} 1 & -3 & 7 & | & 0 \\ 0 & -5 & | & | & 0 \\ 0 & 5 & 2 & | & 0 \end{bmatrix} \xrightarrow{R_2+R_2 \to R_3} \begin{bmatrix} 1 & -3 & 7 & | & 0 \\ 0 & -5 & | & | & 0 \\ 0 & 0 & | & 2 & | & 0 \end{bmatrix}$$

No free variables, the system has no nontrivial solution.

Axi =0, only  $x^2 = 0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ 

5. Describe all solutions of Ax = 0 in parametric vector form.

$$\mathbf{A} = \left[ \begin{array}{cccccc} 1 & 5 & 2 & -6 & 9 & 0 \\ 0 & 0 & 1 & -7 & 4 & -8 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

$$\begin{bmatrix}
A \mid \overrightarrow{b} \end{bmatrix} = \begin{bmatrix}
1 & 5 & 2 & -b & 1 & 0 & 0 \\
0 & 0 & 1 & -7 & 4 & -8 & 0 \\
0 & 0 & 0 & 0 & 0 & 0
\end{bmatrix}$$

$$\begin{array}{c}
R_{2} + 8R_{3} \rightarrow R_{1} \\
0 & 0 & 1 & -7 & 4 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}$$

$$\begin{array}{c}
R_{2} + 8R_{3} \rightarrow R_{1} \\
0 & 0 & 1 & -7 & 4 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0
\end{array}$$

$$\begin{array}{c}
R_{1} - 2R_{2} \rightarrow R_{1} \\
\end{array}$$

$$\begin{array}{c}
R_{1} - 2R_{2} \rightarrow R_{1} \\
\end{array}$$

$$\begin{array}{c}
R_{1} - 2R_{2} \rightarrow R_{1} \\
\end{array}$$

$$\begin{array}{c}
R_{2} + 8R_{3} \rightarrow R_{1} \\
\end{array}$$

$$\begin{array}{c}
R_{3} + R_{1} - R_{2} \rightarrow R_{1} \\
\end{array}$$

$$\begin{array}{c}
R_{2} + R_{1} \rightarrow R_{2} \rightarrow R_{1} \\
\end{array}$$

$$\begin{array}{c}
R_{3} + R_{1} - R_{2} \rightarrow R_{1} \\
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R_{2} + R_{1} \rightarrow R_{2} \rightarrow R_{1} \\
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$$\begin{array}{c}
R_{3} + R_{1} \rightarrow R_{2} \rightarrow R_{1} \\
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$$\begin{array}{c}
R_{4} + R_{1} \rightarrow R_{2} \rightarrow R_{1} \\
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$$\begin{array}{c}
R_{4} + R_{1} \rightarrow R_{2} \rightarrow R_{1} \\
\end{array}$$

Thus, 
$$\begin{cases} x_1 + 5x_2 + 8x_4 + x_5 = 0 \\ x_3 - 7x_4 + 4x_5 = 0 \\ x_6 = 0 \\ 0 = 0 \end{cases}$$

$$\Rightarrow \vec{X} = \begin{bmatrix} x_{1} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \\ x_{6} \end{bmatrix} = \begin{bmatrix} -5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{1} \\ x_{2} \\ x_{4} \\ x_{5} \\ x_{6} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \\ x_{6} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \\ x_{6} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \\ x_{6} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \\ x_{6} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \\ x_{6} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \\ x_{6} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{2} \\ x_{3} \\ x_{4} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{2} - 8x_{4} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - 5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{5} \\ x_{5} \\ x_{5} \end{bmatrix} = \begin{bmatrix} -5x_{1} - x_{5} \\ x_{$$

X2, X4, X5 are arbitrary

- 6. (1) Suppose w, p are two solutions of the equation Ax = b and define  $v_h = w p$ . Show that  $v_h$  is a solution of Ax = 0.
- (2) Suppose Ax = b has a solution. Explain why the solution is unique precisely when Ax = 0 has only the trivial solution.

(1) 
$$A\vec{w} = \vec{b}$$
,  $A\vec{p} = \vec{b}$ ,  $\vec{v}_h = \vec{w} - \vec{p}$ 

$$\therefore A\vec{v}_h = A(\vec{w} - \vec{p}) = A\vec{w} - A\vec{p} = \vec{b} - \vec{b} = \vec{0}$$

$$\vec{v}$$
 Vn is a solution of  $A\vec{x} = \vec{0}$ .

(2) Suppose 
$$A\vec{x} = \vec{b}$$
 has two solutions  $\vec{v}$ ,  $\vec{v}$ , so,  $A\vec{u} = \vec{b}$ ,  $A\vec{v} = \vec{b}$ 

Then, by linearity, 
$$A(\vec{x} - \vec{v}) = A\vec{x} - A\vec{v} = \vec{b} - \vec{b} = \vec{0}$$

Since 
$$A\vec{x} = \vec{0}$$
 has only trivial solution, which means  $\vec{u} - \vec{v} = \vec{0}$ , thus,  $\vec{u} = \vec{v}$