Linear Algebra

Wronskian.
The Vandermonde determinant.

Lecture 15:

**MATH 304** 

The Vandermonde determinant.

Basis of a vector space.

# Linear independence

*Definition.* Let V be a vector space. Vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k \in V$  are called **linearly dependent** if they satisfy a relation

$$r_1\mathbf{v}_1+r_2\mathbf{v}_2+\cdots+r_k\mathbf{v}_k=\mathbf{0},$$

where the coefficients  $r_1, \ldots, r_k \in \mathbb{R}$  are not all equal to zero. Otherwise the vectors  $\mathbf{v}_1, \mathbf{v}_2, \ldots, \mathbf{v}_k$  are called **linearly** independent. That is, if

$$r_1\mathbf{v}_1+r_2\mathbf{v}_2+\cdots+r_k\mathbf{v}_k=\mathbf{0} \implies r_1=\cdots=r_k=0.$$

A set  $S \subset V$  is **linearly dependent** if one can find some distinct linearly dependent vectors  $\mathbf{v}_1, \ldots, \mathbf{v}_k$  in S. Otherwise S is **linearly independent**.

Remark. If a set S (finite or infinite) is linearly independent then any subset of S is also linearly independent.

**Theorem** Vectors  $\mathbf{v}_1, \dots, \mathbf{v}_k \in V$  are linearly dependent if and only if one of them is a linear combination of the other k-1 vectors.

Examples of linear independence.

- Vectors  $\mathbf{e}_1 = (1,0,0)$ ,  $\mathbf{e}_2 = (0,1,0)$ , and  $\mathbf{e}_3 = (0,0,1)$  in  $\mathbb{R}^3$ .
- $\bullet \ \ \mathsf{Matrices} \ \ E_{11} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} \text{, } E_{12} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \text{,}$

$$E_{21}=\begin{pmatrix}0&0\\1&0\end{pmatrix}$$
, and  $E_{22}=\begin{pmatrix}0&0\\0&1\end{pmatrix}$ .

• Polynomials  $1, x, x^2, \dots, x^n, \dots$ 

**Problem.** Show that functions  $e^x$ ,  $e^{2x}$ , and  $e^{3x}$  are linearly independent in  $C^{\infty}(\mathbb{R})$ .

Suppose that  $ae^x + be^{2x} + ce^{3x} = 0$  for all  $x \in \mathbb{R}$ , where a, b, c are constants. We have to show that a = b = c = 0.

 $ae^{x} + be^{2x} + ce^{3x} = 0$ 

Differentiate this identity twice:

$$ae^{x} + 2be^{2x} + 3ce^{3x} = 0,$$
  
 $ae^{x} + 4be^{2x} + 9ce^{3x} = 0.$ 

It follows that  $A(x)\mathbf{v} = \mathbf{0}$ , where

$$A(x) = \begin{pmatrix} e^{x} & e^{2x} & e^{3x} \\ e^{x} & 2e^{2x} & 3e^{3x} \\ e^{x} & 4e^{2x} & 9e^{3x} \end{pmatrix}, \quad \mathbf{v} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}.$$

$$= e^{x}e^{2x}e^{3x} \begin{vmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 4 & 9 \end{vmatrix} = e^{6x} \begin{vmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 4 & 9 \end{vmatrix} = e^{6x} \begin{vmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 1 & 4 & 9 \end{vmatrix}$$
$$= e^{6x} \begin{vmatrix} 1 & 1 & 1 \\ 0 & 1 & 2 \\ 0 & 3 & 8 \end{vmatrix} = e^{6x} \begin{vmatrix} 1 & 2 \\ 3 & 8 \end{vmatrix} = 2e^{6x} \neq 0.$$

 $\det A(x) = e^{x} \begin{vmatrix} 1 & e^{2x} & e^{3x} \\ 1 & 2e^{2x} & 3e^{3x} \\ 1 & 4e^{2x} & 9e^{3x} \end{vmatrix} = e^{x}e^{2x} \begin{vmatrix} 1 & 1 & e^{3x} \\ 1 & 2 & 3e^{3x} \\ 1 & 4 & 9e^{3x} \end{vmatrix}$ 

 $A(x) = \begin{pmatrix} e^x & e^{2x} & e^{3x} \\ e^x & 2e^{2x} & 3e^{3x} \\ e^x & Ae^{2x} & Qe^{3x} \end{pmatrix}, \quad \mathbf{v} = \begin{pmatrix} a \\ b \\ c \end{pmatrix}.$ 

Since the matrix A(x) is invertible, we obtain  $A(x)\mathbf{v} = \mathbf{0} \implies \mathbf{v} = \mathbf{0} \implies a = b = c = 0$ 

#### Wronskian

Let  $f_1, f_2, ..., f_n$  be smooth functions on an interval [a, b]. The **Wronskian**  $W[f_1, f_2, ..., f_n]$  is a function on [a, b] defined by

$$W[f_1, f_2, \ldots, f_n](x) = \begin{vmatrix} f_1(x) & f_2(x) & \cdots & f_n(x) \\ f'_1(x) & f'_2(x) & \cdots & f'_n(x) \\ \vdots & \vdots & \ddots & \vdots \\ f_1^{(n-1)}(x) & f_2^{(n-1)}(x) & \cdots & f_n^{(n-1)}(x) \end{vmatrix}.$$

**Theorem** If  $W[f_1, f_2, ..., f_n](x_0) \neq 0$  for some  $x_0 \in [a, b]$  then the functions  $f_1, f_2, ..., f_n$  are linearly independent in C[a, b].

**Theorem** Let  $\lambda_1, \lambda_2, \ldots, \lambda_k$  be distinct real numbers. Then the functions  $e^{\lambda_1 x}, e^{\lambda_2 x}, \ldots, e^{\lambda_k x}$  are linearly independent.

$$W[e^{\lambda_1 x}, e^{\lambda_2 x}, \dots, e^{\lambda_k x}](x) = \begin{vmatrix} e^{\lambda_1 x} & e^{\lambda_2 x} & \cdots & e^{\lambda_k x} \\ \lambda_1 e^{\lambda_1 x} & \lambda_2 e^{\lambda_2 x} & \cdots & \lambda_k e^{\lambda_k x} \\ \vdots & \vdots & \ddots & \vdots \\ \lambda_1^{k-1} e^{\lambda_1 x} & \lambda_2^{k-1} e^{\lambda_2 x} & \cdots & \lambda_k^{k-1} e^{\lambda_k x} \end{vmatrix}$$

$$=e^{(\lambda_1+\lambda_2+\cdots+\lambda_k)x}\left|egin{array}{cccc} 1&1&\cdots&1\ \lambda_1&\lambda_2&\cdots&\lambda_k\ dots&dots&\ddots&dots\ \lambda_1^{k-1}&\lambda_2^{k-1}&\cdots&\lambda_k^{k-1} \end{array}
ight|.$$

## The Vandermonde determinant

Definition. The **Vandermonde determinant** is the determinant of the following matrix

$$V = \begin{pmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ 1 & x_3 & x_3^2 & \cdots & x_3^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{pmatrix},$$

where  $x_1, x_2, \ldots, x_n \in \mathbb{R}$ . Equivalently,  $V = (a_{ij})_{1 \leq i,j \leq n}$ , where  $a_{ij} = x_i^{j-1}$ .

## Examples.

$$\bullet \quad \begin{vmatrix} 1 & x_1 \\ 1 & x_2 \end{vmatrix} = x_2 - x_1.$$

$$\bullet \quad \begin{vmatrix} 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \\ 1 & x_3 & x_3^2 \end{vmatrix} = \begin{vmatrix} 1 & x_1 & 0 \\ 1 & x_2 & x_2^2 - x_1 x_2 \\ 1 & x_3 & x_3^2 - x_1 x_3 \end{vmatrix}$$

$$= \begin{vmatrix} 1 & 0 & 0 \\ 1 & x_2 - x_1 & x_2^2 - x_1 x_2 \\ 1 & x_3 - x_1 & x_3^2 - x_1 x_3 \end{vmatrix} = \begin{vmatrix} x_2 - x_1 & x_2^2 - x_1 x_2 \\ x_3 - x_1 & x_3^2 - x_1 x_3 \end{vmatrix}$$

$$= \begin{vmatrix} 1 & x_2 - x_1 & x_2^2 - x_1 x_2 \\ 1 & x_3 - x_1 & x_3^2 - x_1 x_3 \end{vmatrix} = \begin{vmatrix} x_2 - x_1 & x_2^2 - x_1 x_2 \\ x_3 - x_1 & x_3^2 - x_1 x_3 \end{vmatrix}$$

$$= (x_2 - x_1) \begin{vmatrix} 1 & x_2 \\ x_3 - x_1 & x_2^2 - x_1 x_3 \end{vmatrix} = (x_2 - x_1)(x_3 - x_1) \begin{vmatrix} 1 & x_2 \\ 1 & x_3 \end{vmatrix}$$

$$|x_3 - x_1 |x_3 - x_1 |x_3 - x_1 |$$

$$= (x_2 - x_1)(x_3 - x_1)(x_3 - x_2).$$

#### **Theorem**

$$\begin{vmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-1} \\ 1 & x_3 & x_3^2 & \cdots & x_3^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-1} \end{vmatrix} = \prod_{1 \le i < j \le n} (x_j - x_i).$$

**Corollary** The Vandermonde determinant is not equal to 0 if and only if the numbers  $x_1, x_2, \ldots, x_n$  are distinct.

Let  $x_1, x_2, \ldots, x_n$  be distinct real numbers.

**Theorem** For any  $b_1, b_2, \ldots, b_n \in \mathbb{R}$  there exists a unique polynomial  $p(x) = a_0 + a_1 x + \cdots + a_{n-1} x^{n-1}$  of degree less than n such that  $p(x_i) = b_i$ ,  $1 \le i \le n$ .

$$\begin{cases} a_0 + a_1 x_1 + a_2 x_1^2 + \dots + a_{n-1} x_1^{n-1} = b_1 \\ a_0 + a_1 x_2 + a_2 x_2^2 + \dots + a_{n-1} x_2^{n-1} = b_2 \\ \dots \\ a_0 + a_1 x_n + a_2 x_n^2 + \dots + a_{n-1} x_n^{n-1} = b_n \end{cases}$$

 $a_0, a_1, \ldots, a_{n-1}$  are unknowns. The coefficient matrix is the Vandermonde matrix.

## **Basis**

Definition. Let V be a vector space. Any linearly independent spanning set for V is called a **basis**.

Suppose that a set  $S \subset V$  is a basis for V.

"Spanning set" means that any vector  $\mathbf{v} \in V$  can be represented as a linear combination

$$\mathbf{v} = r_1 \mathbf{v}_1 + r_2 \mathbf{v}_2 + \cdots + r_k \mathbf{v}_k,$$

where  $\mathbf{v}_1, \dots, \mathbf{v}_k$  are distinct vectors from S and  $r_1, \dots, r_k \in \mathbb{R}$ . "Linearly independent" implies that the above representation is unique:

$$\mathbf{v} = r_1 \mathbf{v}_1 + r_2 \mathbf{v}_2 + \dots + r_k \mathbf{v}_k = r'_1 \mathbf{v}_1 + r'_2 \mathbf{v}_2 + \dots + r'_k \mathbf{v}_k$$

$$\implies (r_1 - r'_1) \mathbf{v}_1 + (r_2 - r'_2) \mathbf{v}_2 + \dots + (r_k - r'_k) \mathbf{v}_k = \mathbf{0}$$

$$\implies r_1 - r'_1 = r_2 - r'_2 = \dots = r_k - r'_k = 0$$

Examples. • Standard basis for  $\mathbb{R}^n$ :  $\mathbf{e}_1 = (1, 0, 0, \dots, 0, 0), \ \mathbf{e}_2 = (0, 1, 0, \dots, 0, 0), \dots$ 

$$\mathbf{e}_n = (0, 0, 0, \dots, 0, 1).$$
  
Indeed,  $(x_1, x_2, \dots, x_n) = x_1 \mathbf{e}_1 + x_2 \mathbf{e}_2 + \dots + x_n \mathbf{e}_n.$ 

• Matrices  $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$ 

form a basis for 
$$\mathcal{M}_{2,2}(\mathbb{R})$$
.
$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} = a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + d \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}.$$

- Polynomials  $1, x, x^2, \dots, x^{n-1}$  form a basis for  $\mathcal{P}_n = \{a_0 + a_1x + \dots + a_{n-1}x^{n-1} : a_i \in \mathbb{R}\}.$
- The infinite set  $\{1, x, x^2, \dots, x^n, \dots\}$  is a basis for  $\mathcal{P}$ , the space of all polynomials.