# Linear Independence

Ch.11

econMath.namun+2016su@gmail.com

April 21, 2017

## 1 Linear Independence

#### LI: Definition

**Definition 1** (Linear Combinations, Span, Linear (in)dependence).  $\mathcal{L}$  is <u>spanned</u> set generated by <u>linear combination</u> of k vectors  $\mathbf{v_1}, \dots, \mathbf{v_k}$ 

$$\mathcal{L}[\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^{\mathbf{n}}] := \left\{ \sum_{i=1}^{k} r_i \mathbf{v}_i : \forall r_i \in \mathbb{R} \right\}$$

 $\mathbf{v_1}, \cdots, \mathbf{v_k}$  are <u>linearly independent</u> iff:

$$\mathcal{L}[\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^{\mathbf{n}}] \subset \mathbb{R}^k$$

Otherwise,  $v_1, \cdots, v_k$  are <u>linearly dependent</u>

Note: Different vectors can span the same space. Canonical basis  $\mathbf{e_i}$  can be the representative vector (basis) for  $\mathbb{R}^n$  space.

#### LI: Alternative Definition

**Definition 2** (Linear (in)dependence: Alternative definition).  $\mathbf{v_1}, \dots, \mathbf{v_k} \in \mathbb{R}^n$  are <u>linearly</u> dependent iff:

$$\exists c_1, \cdots, c_k \neq 0 \quad s.t. \quad \sum_{i=1}^k c_i \mathbf{v_i} = \mathbf{0}$$

 $\mathbf{v_1}, \cdots, \mathbf{v_k} \in \mathbb{R}^n$  are linearly independent iff:

$$\sum_{i=1}^{k} c_i \mathbf{v_i} = \mathbf{0} \quad \Rightarrow \quad c_1 = \dots = c_k = 0$$

**Theorems** 

**Theorem 1** (11.1).  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^n$  are linearly dependent iff

$$A \begin{pmatrix} c_1 \\ \vdots \\ c_k \end{pmatrix} = \mathbf{0}$$

has nonzero solution  $\mathbf{c}$ , where A is  $n \times k$  matrix whose  $C_i = \mathbf{v_i}$ . i.e.,

$$A = \begin{pmatrix} \mathbf{v_1} & \mathbf{v_2} & \cdots & \mathbf{v_k} \end{pmatrix}$$

**Theorem 2** (11.2).  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_n} \in \mathbb{R}^n$  are linearly independent iff

$$\det \begin{pmatrix} \mathbf{v_1} & \mathbf{v_2} & \cdots & \mathbf{v_n} \end{pmatrix} \neq 0$$

#### Checking LI

Procedure: Checking Linear (in)dependence

- 1. Stack  $\mathbf{v_i}$  to make  $k \times n$  matrix, A.
- 2. Calculate rank(A): this is the dimension of  $\mathcal{L}[\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^n]$
- 3.  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^{\mathbf{n}}$  are linearly independent iff rank(A) = k. Otherwise, they are linearly dependent

Note:  $rank(A^T) = rank(A)$ 

**Theorem 3** (11.3). If k > n, any set of k vectors in  $\mathbb{R}^n$  is linearly dependent.

## 2 Spanning Sets

### 3 Basis and Dimension in $\mathbb{R}^n$

Basis

**Definition 3** (Basis). Let  $V = \mathcal{L}[\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^n]$ . If  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^n$  are linearly independent,  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^n$  is called a <u>basis</u> of V. More generally,  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^n$  forms a basis of V if:

- 1.  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^n \ span \ V$
- 2.  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_k} \in \mathbb{R}^n$  are linearly independent

**Theorem 4** (11.7). Every basis of  $\mathbb{R}^n$  contains n vectors.

#### Linear Independence and Basis

**Theorem 5** (11.8). Let  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_n} \in \mathbb{R}^{\mathbf{n}}$  be a collection of n vectors in  $\mathbb{R}^n$ . And let  $n \times n$  matrix  $A = (\mathbf{v_1} \cdots \mathbf{v_n})$ . Then the following statements are equivalent:

- 1.  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_n} \in \mathbb{R}^n$  are linearly independent.
- 2.  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_n} \in \mathbb{R}^n \ span \ \mathbb{R}^n$
- 3.  $\mathbf{v_1}, \mathbf{v_2}, \cdots, \mathbf{v_n} \in \mathbb{R}^n$  form a basis of  $\mathbb{R}^n$
- 4.  $\det(A) \neq 0$