

Lecture 08

Chapter 3. Vector Spaces

3.4. Basis and Dimension

3.5. Changing of Basis

3.4 Basis and Dimension

The elements of a **minimal spanning set** form the basic building blocks for the whole vector space.

Basis of Vector Space

Definition 1. The vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ form a **basis** [基, 基底] for a vector space V **if and only if**

- (i) $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ are linearly independent;
- (ii) $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ span V .

Example 1. In \mathbf{R}^3 , the vectors $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$ form a basis of the vector space \mathbf{R}^3 , which are called the **standard basis** [标准基] for \mathbf{R}^3 .

However, there are many bases for \mathbf{R}^3 . For example,

$$\left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix} \right\} \text{ and } \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \right\} \text{ are both bases for } \mathbf{R}^3.$$

Example 2. In $\mathbf{R}^{2 \times 2}$, consider the set $\{E_{11}, E_{12}, E_{21}, E_{22}\}$, where

$$E_{11} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, E_{12} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, E_{21} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, E_{22} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}.$$

(i) If
$$c_1 E_{11} + c_2 E_{12} + c_3 E_{21} + c_4 E_{22} = O,$$

then
$$\begin{pmatrix} c_1 & c_2 \\ c_3 & c_4 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix},$$

so $c_1 = c_2 = c_3 = c_4 = 0$. $E_{11}, E_{12}, E_{21}, E_{22}$ are linearly independent.

(ii) If A is in $\mathbf{R}^{2 \times 2}$, then

$$A = a_{11}E_{11} + a_{12}E_{12} + a_{21}E_{21} + a_{22}E_{22}.$$

Thus, $E_{11}, E_{12}, E_{21}, E_{22}$ span $\mathbf{R}^{2 \times 2}$ and hence form a basis for $\mathbf{R}^{2 \times 2}$.

$\{E_{11}, E_{12}, E_{21}, E_{22}\}$: **Standard basis** for $\mathbf{R}^{2 \times 2}$

Example 3. In the vector space P_3 , the vector set $\{1, x, x^2\}$ is a basis, since all vectors in P_3 can be represented as a linear combination of them.

This basis is called the **standard basis** for P_3 .

Theorem 1. If $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is a **spanning set** for a vector space V , then any collection of m vectors in V , where $m > n$, is linearly dependent.

Proof. Let $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$ be m vectors in V where $m > n$.

- Since $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ span V , we have

$$\mathbf{u}_i = a_{i1}\mathbf{v}_1 + a_{i2}\mathbf{v}_2 + \dots + a_{in}\mathbf{v}_n \quad \text{for } i = 1, 2, \dots, m.$$

- A linear combination $c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_m\mathbf{u}_m$ can be written in the form

$$c_1 \sum_{j=1}^n a_{1j}\mathbf{v}_j + c_2 \sum_{j=1}^n a_{2j}\mathbf{v}_j + \dots + c_m \sum_{j=1}^n a_{mj}\mathbf{v}_j.$$

Rearranging the terms, we have

Theorem 1. If $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is a **spanning set** for a vector space V , then any collection of m vectors in V , where $m > n$, is linearly dependent.

Proof. (continue)

$$c_1 \mathbf{u}_1 + c_2 \mathbf{u}_2 + \dots + c_m \mathbf{u}_m = \sum_{i=1}^m \left[c_i \left(\sum_{j=1}^n a_{ij} \mathbf{v}_j \right) \right] = \sum_{j=1}^n \left(\sum_{i=1}^m a_{ij} c_i \right) \mathbf{v}_j.$$

Consider the system of equations

$$\sum_{i=1}^m a_{ij} c_i = 0, \quad j = 1, 2, \dots, n,$$

where $c_i, i = 1, 2, \dots, m$ are unknowns. This is a homogeneous system with more unknowns (m) than equations (n), then the system must have a nontrivial solution $(\hat{c}_1, \hat{c}_2, \dots, \hat{c}_m)$. Then

$$\hat{c}_1 \mathbf{u}_1 + \hat{c}_2 \mathbf{u}_2 + \dots + \hat{c}_m \mathbf{u}_m = \sum_{j=1}^n 0 \mathbf{v}_j = \mathbf{0}.$$

Corollary 1. If $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ and $\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$ are both **bases** for a vector space V , then $n = m$.

Proof. By Theorem 1,

- since $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ span V and $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$ are linearly independent, we have $m \leq n$;
- by the same reason, $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$ span V , and $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ are linearly independent, so $n \leq m$.

Therefore, $n = m$.

Dimension of Vector Space

Definition 2. Let V be a vector space. If V has a basis consisting of n vectors, we say that V has **dimension [维数] n** , denoted by $\dim V = n$.

- The subspace $\{\mathbf{0}\}$ of V is said to have dimension 0;
- V is said to be **finite-dimensional [有限维]** if there is a finite set of vectors that spans V ; otherwise, we say that V is **infinite-dimensional [无限维]**.

Example.

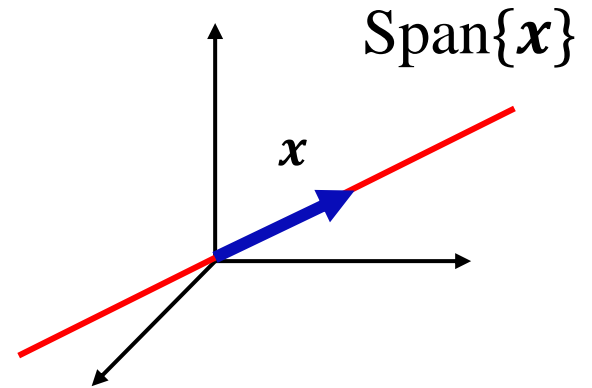
- \mathbf{R}^3 : standard basis $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$, $\dim \mathbf{R}^3 = 3$.
- $\mathbf{R}^{2 \times 2}$: standard basis $\{E_{11}, E_{12}, E_{21}, E_{22}\}$, $\dim \mathbf{R}^{2 \times 2} = 4$.
- P_n : standard basis $\{1, x, \dots, x^{n-1}\}$, $\dim P_n = n$.

Example 4.

- Let $\mathbf{x} = (x_1, x_2, x_3)^T$ be a **nonzero** vector in \mathbf{R}^3 , then

$$\text{Span}\{\mathbf{x}\} = \{\alpha\mathbf{x} | \alpha \in \mathbf{R}\}$$

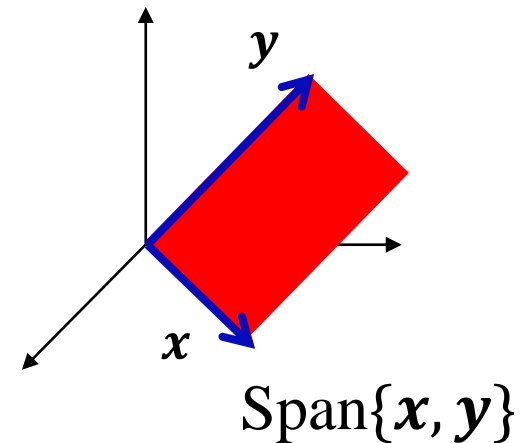
is a one-dimensional vector space.



- If \mathbf{x} and \mathbf{y} are two **linearly independent** vectors in \mathbf{R}^3 , then

$$\text{Span}\{\mathbf{x}, \mathbf{y}\} = \{\alpha\mathbf{x} + \beta\mathbf{y} | \alpha, \beta \in \mathbf{R}\}$$

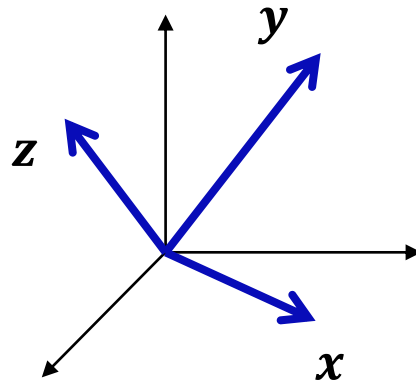
is a two-dimensional vector space.



- If \mathbf{x} , \mathbf{y} and \mathbf{z} are **linearly independent**, then

$$\text{Span}\{\mathbf{x}, \mathbf{y}, \mathbf{z}\} = \{\alpha\mathbf{x} + \beta\mathbf{y} + \gamma\mathbf{z} \mid \alpha, \beta, \gamma \in \mathbf{R}\}$$

is just \mathbf{R}^3 .



$$\text{Span}\{\mathbf{x}, \mathbf{y}, \mathbf{z}\} = \mathbf{R}^3$$

Example 5. The vector space of all polynomials P is infinite-dimensional.

Proof. In fact, if P is finite-dimensional, say of dimension n , any set of $n + 1$ vectors would be linearly dependent. However, we can prove that $1, x, x^2, \dots, x^n$ are linearly independent. This means that P cannot be of dimension n . Since n is arbitrary, P must be infinite-dimensional.

Note: The same argument shows that the vector space $C[a, b]$ is infinite-dimensional.

Theorem. If V is a vector space of dimension $n > 0$:

- (i) Any set of n linearly independent vectors span V ;
- (ii) Any n vectors that span V are linearly independent.

Proof. (i) Suppose that $\mathbf{v}_1, \dots, \mathbf{v}_n$ are linearly independent and \mathbf{v} is any other vector in V . Since V has dimension n , it has a basis consisting of n vectors and these vectors span V . Then $\mathbf{v}_1, \dots, \mathbf{v}_n, \mathbf{v}$ must be linearly dependent. Therefore, any $\mathbf{v} \in V$ can be written in form of linear combination of $\mathbf{v}_1, \dots, \mathbf{v}_n$. This means that $\mathbf{v}_1, \dots, \mathbf{v}_n$ span V .

Theorem. If V is a vector space of dimension $n > 0$:

- (i) Any set of n linearly independent vectors span V ;
- (ii) Any n vectors that span V are linearly independent.

Proof. (ii) Suppose that $\mathbf{v}_1, \dots, \mathbf{v}_n$ span V . If $\mathbf{v}_1, \dots, \mathbf{v}_n$ are linearly dependent, then one of the \mathbf{v}_i 's, say \mathbf{v}_n , can be written as a linear combination of the others. It follows that $\mathbf{v}_1, \dots, \mathbf{v}_{n-1}$ will still span V and this means that the dimension of V must be smaller than n . This contradicts with $\dim V = n$.

Example 6. Show that $\left\{ \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}, \begin{pmatrix} -2 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \right\}$ is a basis for \mathbf{R}^3 .

Proof. Since $\dim \mathbf{R}^3 = 3$, it is enough to show that the three vectors are linearly independent. This follows since

$$\begin{vmatrix} 1 & -2 & 1 \\ 2 & 1 & 0 \\ 3 & 0 & 1 \end{vmatrix} = 2 \neq 0.$$

Theorem. If V is a vector space of dimension $n \geq 0$, then

- (i) No set of less than n vectors can span V .
- (ii) Any subset of less than n linearly independent vectors can be extended to form a basis for V .
- (iii) Any spanning set containing more than n vectors can be pared down to form a basis for V .

Note: This theorem can be easily proved and this is left to the reader.

Some Vector Spaces and their Standard Basis

- (1) \mathbf{R}^n : the set $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$.
- (2) $\mathbf{R}^{m \times n}$: the set $\{E_{ij} | i = 1, 2, \dots, m; j = 1, 2, \dots, n\}$, where E_{ij} is the $m \times n$ matrix with all zero entries except the (i, j) th entry.
- (3) P_n : the set $\{1, x, x^2, \dots, x^{n-1}\}$.

3.5 Changing of basis

Coordinate of Vector

If $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ form a **basis** of a finite-dimensional vector space V , then they are linearly independent and they span the whole space V .

Any vector \mathbf{x} can be written **uniquely** as a linear combination of $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$

$$\mathbf{x} = \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n.$$

By using this basis, we find a simple way to represent all vectors in the space V . (**coordinate system** [坐标系])

Definition 1. Suppose $E = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is an **ordered basis** of vector space V . Then any vector \mathbf{x} in V can be **uniquely** represented as

$$\mathbf{x} = \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n$$

where $\alpha_i, i = 1, 2, \dots, n$ are scalars and are called **coordinates [坐标]** of the vector \mathbf{x} in V with respect to the ordered basis E .

The vector $(\alpha_1, \alpha_2, \dots, \alpha_n)^T_E$ is called **coordinate vector [坐标向量]** of vector \mathbf{x} in V with respect to basis E , denoted by $[\mathbf{x}]_E$.

$$\begin{aligned} \mathbf{x} = \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n &= (\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n) \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix} \\ &= (\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n) [\mathbf{x}]_E. \end{aligned}$$

Example 1. Suppose $E = \{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$, $F = \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$

are two vector sets of \mathbf{R}^3 . Show that E and F are both bases of \mathbf{R}^3 and find the coordinate vectors of $\mathbf{x} = (1, 2, 3)^T$ w.r.t. the ordered bases E and F .

Solution. E is the standard basis of \mathbf{R}^3 . To show that F forms a basis of \mathbf{R}^3 it is enough to show that the three vectors are linearly independent. This can be done by check

$$\begin{vmatrix} 1 & 1 & 0 \\ 1 & 1 & 1 \\ 1 & 0 & 1 \end{vmatrix} = 1 \neq 0.$$

Solution. (continue) $\mathbf{x} = (1,2,3)^T$

- Suppose that $[\mathbf{x}]_E = (x_1, x_2, x_3)^T$, then we have

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \mathbf{x} = x_1 \mathbf{e}_1 + x_2 \mathbf{e}_2 + x_3 \mathbf{e}_3 = x_1 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + x_3 \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

which implies that $[\mathbf{x}]_E = (1,2,3)^T$.

- Suppose that $[\mathbf{x}]_F = (y_1, y_2, y_3)^T$, then we have

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \mathbf{x} = y_1 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} + y_2 \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} + y_3 \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} y_1 + y_2 \\ y_1 + y_2 + y_3 \\ y_1 + y_3 \end{pmatrix}$$

This leads to the linear system

$$\begin{aligned} y_1 + y_2 &= 1 \\ y_1 + y_2 + y_3 &= 2 \\ y_1 + y_3 &= 3 \end{aligned}$$

By solving the system, we obtain $[\mathbf{x}]_F = (2, -1, 1)^T$.

Remark.

- The coordinate vectors of \mathbf{x} w.r.t. different bases will be generally different from each other, so when we refer to a coordinate vector of a vector, we must **make sure** which basis is used.
- When $V = \mathbf{R}^n$ and $E = \{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$ is the standard basis, for any vector $x \in \mathbf{R}^n$ we have $[\mathbf{x}]_E = \mathbf{x}$.

Example 2. Show that

$$E = \{1, x, x^2\} \text{ and } F = \{1 + x, x(1 + x), x^2\}$$

are both bases of P_3 . Find the coordinate vectors of polynomial $p(x) = (1 + x)^2$ w.r.t. the ordered bases E and F .

Solution. E is the standard basis of P_3 . To show F is a basis of P_3 , it is enough to show that the three polynomials in F are linearly independent. For this purpose, assume that

$$\alpha_1(1 + x) + \alpha_2x(1 + x) + \alpha_3x^2 = 0.$$

$$\alpha_1 = 0$$

$$\text{We then have } \alpha_1 + \alpha_2 = 0 \implies \alpha_1 = \alpha_2 = \alpha_3 = 0$$

$$\alpha_2 + \alpha_3 = 0$$

Thus the three vectors in the set F are linearly independent and F forms a basis of P_3 .

Solution. (continue) $p(x) = 1 + 2x + x^2$

- Suppose $[p(x)]_E = (c_1, c_2, c_3)^T$. Then

$$p(x) = c_1 \cdot 1 + c_2 \cdot x + c_3 \cdot x^2$$

By comparing the coefficients, we see that $[p(x)]_E = (1, 2, 1)^T$.

- Assume that $[p(x)]_F = (a_1, a_2, a_3)^T$ and then

$$p(x) = a_1(1 + x) + a_2x(1 + x) + a_3x^2$$

$$1 + 2x + x^2 = a_1 + (a_1 + a_2)x + (a_2 + a_3)x^2.$$

By comparing the coefficients, we see that

$$a_1 = 1$$

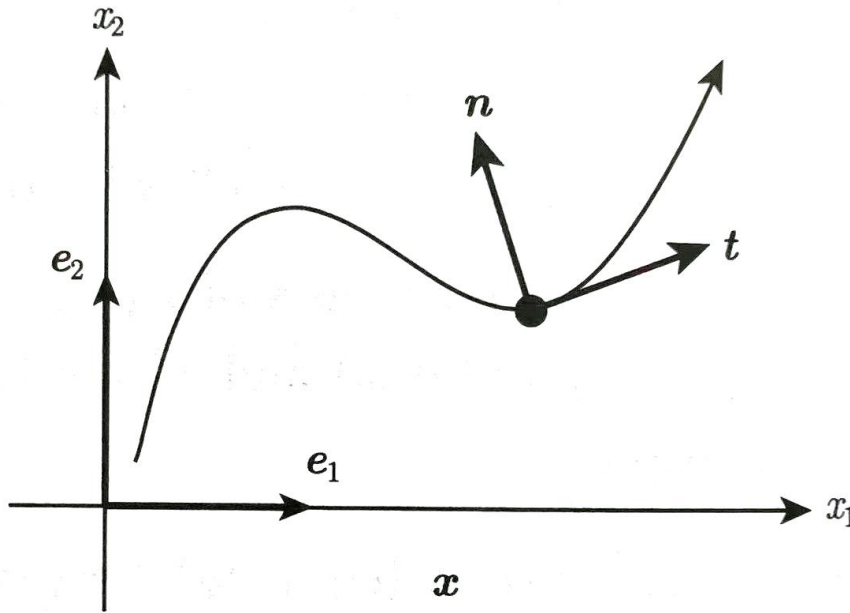
$$a_1 + a_2 = 2$$

$$a_2 + a_3 = 1$$

Therefore $[p(x)]_F = (1, 1, 0)^T$.

Changing of basis in \mathbb{R}^2

Method of **changing basis**
[基变换]



E : the original basis of V

F : the new basis of V

Question: How does the coordinate vector of $x \in V$ change with the changing of basis?

(1) Given $[x]_E$, find $[x]_F$;

(2) Given $[x]_F$, find $[x]_E$.

Let $E = \{\mathbf{e}_1, \mathbf{e}_2\}$, $F = \left\{ \mathbf{u}_1 = \begin{pmatrix} 0 \\ 2 \end{pmatrix}, \mathbf{u}_2 = \begin{pmatrix} 3 \\ 1 \end{pmatrix} \right\}$ be bases of \mathbf{R}^2 .

Notice that

$$\begin{aligned}\mathbf{u}_1 &= 0\mathbf{e}_1 + 2\mathbf{e}_2, \\ \mathbf{u}_2 &= 3\mathbf{e}_1 + 1\mathbf{e}_2.\end{aligned}$$

If $[\mathbf{x}]_F = (c_1, c_2)^T$, we have

$$\begin{aligned}\mathbf{x} &= c_1\mathbf{u}_1 + c_2\mathbf{u}_2 = (0c_1 + 3c_2)\mathbf{e}_1 + (2c_1 + 1c_2)\mathbf{e}_2 \\ &= (\mathbf{e}_1, \mathbf{e}_2) \begin{pmatrix} 0c_1 + 3c_2 \\ 2c_1 + 1c_2 \end{pmatrix} = (\mathbf{e}_1, \mathbf{e}_2) \begin{pmatrix} 0 & 3 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \end{pmatrix}.\end{aligned}$$

Let

$$U = \begin{pmatrix} 0 & 3 \\ 2 & 1 \end{pmatrix} = (\mathbf{u}_1, \mathbf{u}_2).$$

Then

$$[\mathbf{x}]_E = U[\mathbf{x}]_F.$$

transition matrix from F to E

Let $E = \{\mathbf{e}_1, \mathbf{e}_2\}$, $F = \left\{ \mathbf{u}_1 = \begin{pmatrix} 0 \\ 2 \end{pmatrix}, \mathbf{u}_2 = \begin{pmatrix} 3 \\ 1 \end{pmatrix} \right\}$ be bases of \mathbf{R}^2 .

$$[\mathbf{x}]_E = U[\mathbf{x}]_F$$

- The matrix U is called the **transition matrix** [过渡矩阵] from the ordered basis F to the standard basis E .
- U is invertible, since each column of U is a vector of the basis F .
- If we know $[\mathbf{x}]_E$ and the transition matrix from basis F to basis E is U , then
$$[\mathbf{x}]_F = U^{-1}[\mathbf{x}]_E.$$
 U^{-1} is the transition matrix from standard basis E to basis F .

Exercise. Let $\mathbf{x} = (1,1)^T$. Find the coordinates of \mathbf{x} w.r.t. the basis F .

Let $\{\mathbf{u}_1, \mathbf{u}_2\}, \{\mathbf{v}_1, \mathbf{v}_2\}$ be two bases of \mathbf{R}^2 .

Compute the **transition matrix** from basis $\{\mathbf{u}_1, \mathbf{u}_2\}$ to $\{\mathbf{v}_1, \mathbf{v}_2\}$:

Step 1. Find the transition matrix U from basis $\{\mathbf{u}_1, \mathbf{u}_2\}$ to the standard basis $\{\mathbf{e}_1, \mathbf{e}_2\}$,

$$U = (\mathbf{u}_1, \mathbf{u}_2).$$

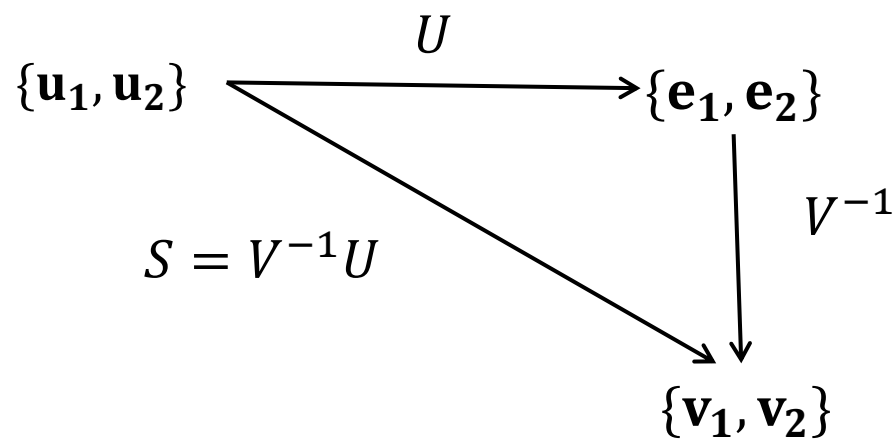
Step 2. Find the transition matrix V from basis $\{\mathbf{v}_1, \mathbf{v}_2\}$ to the standard basis $\{\mathbf{e}_1, \mathbf{e}_2\}$,

$$V = (\mathbf{v}_1, \mathbf{v}_2).$$

Step 3. The transition matrix S from basis $\{\mathbf{u}_1, \mathbf{u}_2\}$ to $\{\mathbf{v}_1, \mathbf{v}_2\}$ can be calculated by

$$S = V^{-1}U.$$

Let $\{\mathbf{u}_1, \mathbf{u}_2\}, \{\mathbf{v}_1, \mathbf{v}_2\}$ be two bases of \mathbf{R}^2 .



Example 3. Suppose

$$\mathbf{u}_1 = \begin{pmatrix} 5 \\ 2 \end{pmatrix}, \mathbf{u}_2 = \begin{pmatrix} 7 \\ 3 \end{pmatrix}, \quad \mathbf{v}_1 = \begin{pmatrix} 3 \\ 2 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}.$$

Find the transition matrix corresponding to the change of basis from $\{\mathbf{u}_1, \mathbf{u}_2\}$ to $\{\mathbf{v}_1, \mathbf{v}_2\}$.

Solution. • The transition matrix from basis $\{\mathbf{u}_1, \mathbf{u}_2\}$ to the standard

basis is $U = (\mathbf{u}_1, \mathbf{u}_2) = \begin{pmatrix} 5 & 7 \\ 2 & 3 \end{pmatrix}$.

• The transition matrix from basis $\{\mathbf{v}_1, \mathbf{v}_2\}$ to the standard basis is

$$V = (\mathbf{v}_1, \mathbf{v}_2) = \begin{pmatrix} 3 & 1 \\ 2 & 1 \end{pmatrix}, \quad V^{-1} = \begin{pmatrix} 1 & -1 \\ -2 & 3 \end{pmatrix}.$$

• The transition matrix S from $\{\mathbf{u}_1, \mathbf{u}_2\}$ to $\{\mathbf{v}_1, \mathbf{v}_2\}$ is

$$S = V^{-1}U = \begin{pmatrix} 3 & 4 \\ -4 & -5 \end{pmatrix}.$$

Changing of basis in an n -dim. vector space

Case $V = \mathbf{R}^n$.

- $E = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$, $F = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$: two bases of \mathbf{R}^n .
- $\mathbf{x} \in \mathbf{R}^n$

$$\begin{aligned}\mathbf{x} &= \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \dots + \alpha_n \mathbf{v}_n = (\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n) \begin{pmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{pmatrix} \\ &= (\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n) [\mathbf{x}]_E\end{aligned}$$

$$\begin{aligned}\mathbf{x} &= \beta_1 \mathbf{u}_1 + \beta_2 \mathbf{u}_2 + \dots + \beta_n \mathbf{u}_n = (\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n) \begin{pmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_n \end{pmatrix} \\ &= (\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n) [\mathbf{x}]_F.\end{aligned}$$

Case $V = \mathbb{R}^n$.

- We have $(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n)[\mathbf{x}]_E = (\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n)[\mathbf{x}]_F$.
- Let $V = (\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n)$ and $U = (\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n)$.

Then U and V are both invertible matrices.

- If we denote $S = V^{-1}U$,

then the relations between two coordinate vectors are

$$[\mathbf{x}]_E = S[\mathbf{x}]_F$$

and

$$[\mathbf{x}]_F = S^{-1}[\mathbf{x}]_E.$$

general n -dimensional vector space V

- $E = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$, $F = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$: two ordered bases of V .

$$\mathbf{u}_j = \sum_{i=1}^n a_{ij} \mathbf{v}_i, \quad j = 1, 2, \dots, n.$$

- $\mathbf{x} \in V$. $[\mathbf{x}]_E = (\alpha_1, \alpha_2, \dots, \alpha_n)^T$, $[\mathbf{x}]_F = (\beta_1, \beta_2, \dots, \beta_n)^T$

$$\mathbf{x} = \beta_1 \mathbf{u}_1 + \beta_2 \mathbf{u}_2 + \dots + \beta_n \mathbf{u}_n$$

$$= \beta_1 \sum_{i=1}^n a_{i1} \mathbf{v}_i + \beta_2 \sum_{i=1}^n a_{i2} \mathbf{v}_i + \dots + \beta_n \sum_{i=1}^n a_{in} \mathbf{v}_i$$

$$= \sum_{j=1}^n \beta_j \sum_{i=1}^n a_{ij} \mathbf{v}_i = \sum_{i=1}^n \left(\sum_{j=1}^n a_{ij} \beta_j \right) \mathbf{v}_i.$$

α_i

general n -dimensional vector space V

Compared with $\mathbf{x} = \alpha_1 \mathbf{v}_1 + \alpha_2 \mathbf{v}_2 + \cdots + \alpha_n \mathbf{v}_n$, we obtain

$$\alpha_i = \sum_{j=1}^n a_{ij} \beta_j .$$

This gives

$$[\mathbf{x}]_E = U[\mathbf{x}]_F$$

where $U = (a_{ij})$ is the transition matrix.

The matrix U is invertible.

Example. Suppose that in P_3 , we want to change from the ordered basis $\{1, x, x^2\}$ to the ordered basis $\{1, 2x, 4x^2 - 2\}$.

Solution. It is easier to find the transition matrix from $\{1, 2x, 4x^2 - 2\}$ to $\{1, x, x^2\}$, since

$$1 = 1 \cdot 1 + 0x + 0x^2$$

$$2x = 0 \cdot 1 + 2x + 0x^2$$

$$4x^2 - 2 = -2 \cdot 1 + 0x + 4x^2.$$

The transition matrix is $S = \begin{pmatrix} 1 & 0 & -2 \\ 0 & 2 & 0 \\ 0 & 0 & 4 \end{pmatrix}.$

Solution. (continue) The inverse of S will be the transition matrix from $\{1, x, x^2\}$ to $\{1, 2x, 4x^2 - 2\}$

$$S^{-1} = \begin{pmatrix} 1 & 0 & 1/2 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1/4 \end{pmatrix}.$$

Given any $p(x) = a + bx + cx^2$ in P_3 , to find the coordinates of $p(x)$ with respect to $\{1, 2x, 4x^2 - 2\}$, we simply multiply

$$\begin{pmatrix} 1 & 0 & 1/2 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1/4 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} a + \frac{1}{2}c \\ \frac{1}{2}b \\ \frac{1}{4}c \end{pmatrix}.$$

$$\text{Thus, } p(x) = \left(a + \frac{1}{2}c\right) \cdot 1 + \left(\frac{1}{2}b\right) \cdot 2x + \left(\frac{1}{4}c\right) \cdot (4x^2 - 2).$$

Review

- Basis and dimension
- Changing of basis

Preview

- ❖ Row space and column space