

Basis for a Vector Space

Definition: If $S = \{v_1, v_2, \dots, v_n\}$ is a set of vectors in a finite-dimensional vector space V , then we say S is a **basis** for V if

(a) S spans V

& (b) S is linearly independent

Example standard unit vectors e_1, e_2, \dots, e_n for \mathbb{R}^n
 \Rightarrow standard basis for \mathbb{R}^n

Example: $S = \{1, x, x^2, \dots, x^n\}$ is a basis for P_n

\Rightarrow standard basis for P_n

Example: $M_1 = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, M_2 = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, M_3 = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, M_4 = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$

form a basis for M_{22}

Theorem [uniqueness of basis representation]

If $S = \{v_1, v_2, \dots, v_n\}$ is a basis for a vector space V , then every $v \in V$ can be expressed as $v = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$ in "exactly one way".

(Proof): S is a basis for $V \Rightarrow v$ can be expressed as a linear combination of vectors in S

Suppose that

$$v = c_1 v_1 + c_2 v_2 + \dots + c_n v_n \quad \text{--- (1)}$$

$$\text{and } v = d_1 v_1 + d_2 v_2 + \dots + d_n v_n \quad \text{--- (2)}$$

$$\text{--- (1) --- (2)} \Rightarrow (c_1 - d_1) v_1 + (c_2 - d_2) v_2 + \dots + (c_n - d_n) v_n = 0$$

$\because S$ is a basis $\Rightarrow S$ is a linearly independent set

$$\therefore c_1 - d_1 = 0, c_2 - d_2 = 0, \dots, c_n - d_n = 0$$

That is, $c_1 = d_1, c_2 = d_2, \dots, c_n = d_n$, So the representation is unique

Since the basis representation is unique :

Definition Let $S = \{v_1, v_2, \dots, v_n\}$ be an ordered basis for

a vector space V , and $v = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$

Then the scalars c_1, c_2, \dots, c_n are the **coordinates of v relative to the basis S**

The coordinate "vector" of v relative to S :

$$[v]_S = (c_1, c_2, \dots, c_n)$$

$$\text{or } [v]_S = \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix}$$

Example: $P(x) = c_0 + c_1 x + c_2 x^2 + \dots + c_n x^n$

standard basis for $P_n : \{1, x, x^2, \dots, x^n\} \subset S$

$$\text{then } [P]_S = (c_0, c_1, \dots, c_n)$$

Example: $B = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ the standard basis for $M_{2,2}$:

$$S = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$$

$$\text{then } [B]_S = (a, b, c, d)$$

Example: Suppose we have $\{v_1 = (1, 2, 1),$

$$v_2 = (2, 9, 0),$$

$$v_3 = (3, 3, 4)\} = S$$

S is a basis for \mathbb{R}^3 . Consider $[v]_S = (-1, 3, 2)$

$$\text{Then } v = (-1) \cdot (1, 2, 1) + 3 \cdot (2, 9, 0) + 2 \cdot (3, 3, 4)$$

$$= (11, 31, 7)$$

If we have $v = (5, -1, 9)$

we can assume that $v = c_1 v_1 + c_2 v_2 + c_3 v_3$

$$\Rightarrow (5, -1, 9) = c_1 (1, 2, 1) + c_2 (2, 9, 0) + c_3 (3, 3, 4)$$

Solving the system $\Rightarrow c_1 = 1, c_2 = -1, c_3 = 2$

Then we know that $[v]_S = (1, -1, 2)$

Dimension and Change of Basis

* 4.6.2 Theorem Let V be a finite-dimensional vector space

and let $\{v_1, v_2, \dots, v_n\}$ be any basis for V

(a) If a set in V has more than n vectors,
then it is linearly dependent

(b) If a set in V has fewer than n vectors,
then it does NOT span V

* 4.6.3 Theorem All bases for a finite-dimensional vector space have the same number of vectors.

What's the "dimension" of a vector space?

Definition: The dimension of a vector space V

$\Rightarrow \dim(V)$: the number of vectors
in a basis for V

Note: $\dim(\emptyset) := 0$

\hookrightarrow is defined

Example:

$$\dim(\mathbb{R}^n) = n$$

$$\dim(P_n) = n+1$$

$$\dim(M_{mn}) = mn$$

Example: If $S = \{v_1, v_2, \dots, v_r\}$, then every vector in $\text{span}(S)$ is expressible as linear combination of the vectors in S

If v_1, v_2, \dots, v_r are linearly independent

$\Rightarrow \{v_1, v_2, \dots, v_r\}$ is a basis for $\text{span}(S)$

$\Rightarrow \dim(\text{span}(\{v_1, v_2, \dots, v_r\})) = r$

Example (dimension of a Solution Space)

Find a basis for and the dimension of the solution space of

$$x_1 + 3x_2 - 2x_3 + 2x_5 = 0$$

$$2x_1 + 6x_2 - 5x_3 - 2x_4 + 4x_5 - 3x_6 = 0$$

$$5x_3 + 10x_4 + 15x_6 = 0$$

$$2x_1 + 6x_2 + 8x_4 + 4x_5 + 18x_6 = 0$$

(sol):

Solve the system above, we have:

$$x_1 = -3r - 4s - 2t$$

$$x_2 = r$$

$$x_3 = -2s$$

$$x_4 = s$$

$$x_5 = t$$

$$x_6 = 0$$

$$\begin{aligned} &= (-3r - 4s - 2t, r, -2s, s, t, 0) \\ &= r(-3, 1, 0, 0, 0, 0) + s(-4, 0, -2, 1, 0, 0) \end{aligned}$$

$$+ t(-2, 0, 0, 1, 0)$$

$$(-3, 1, 0, 0, 0, 0), (-4, 0, -2, 1, 0, 0), (-2, 0, 0, 1, 0, 0)$$

spans the solution space

↳ Check that whether they are linearly independent

$$\Rightarrow \text{dimension} = 3$$

Change of Basis / (Recall)

Let V be a vector space of $\dim(V) = 2$

Let $\beta_1 = \{u_1, u_2\}$, $\beta_2 = \{w_1, w_2\}$ be two bases of V

Suppose that

$$[u_1]_{\beta_2} = \begin{bmatrix} a \\ b \end{bmatrix}, \quad [u_2]_{\beta_2} = \begin{bmatrix} c \\ d \end{bmatrix}$$

That is,

$$u_1 = aw_1 + bw_2$$

$$u_2 = cw_1 + dw_2$$

Now, let $v \in V$ be any vector, suppose that

$$[v]_{\beta_1} = \begin{bmatrix} k_1 \\ k_2 \end{bmatrix}, \text{ that is, } v = k_1 u_1 + k_2 u_2$$

$$\begin{aligned} v &= k_1(aw_1 + bw_2) + k_2(cw_1 + dw_2) \\ &= (k_1a + k_2c)w_1 + (k_1b + k_2d)w_2 \end{aligned}$$

$$\begin{aligned} \text{Thus, } [v]_{\beta_2} &= \begin{bmatrix} k_1a + k_2c \\ k_1b + k_2d \end{bmatrix} = \begin{bmatrix} a & c \\ b & d \end{bmatrix} \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} \\ &= \underbrace{\begin{bmatrix} a & c \\ b & d \end{bmatrix}}_{[I]_{\beta_2}} [v]_{\beta_1} \end{aligned}$$

Therefore,

$$[v]_{\beta_2} = [I]_{\beta_1}^{\beta_2} [v]_{\beta_1}$$

$$[v]_{\beta_1} = [I]_{\beta_2}^{\beta_1} [v]_{\beta_2}$$

Note: $[I]_{\beta_1}^{\beta_2} [I]_{\beta_2}^{\beta_1} = I \rightsquigarrow$ Later it will be used for an efficient algorithm.

Example Consider two bases $B = \{u_1, u_2\}$, $B' = \{u'_1, u'_2\}$ for \mathbb{R}^2

where $u_1 = (1, 0)$, $u_2 = (0, 1)$, $u'_1 = (1, 1)$, $u'_2 = (2, 1)$

Find $[I]_{B'}^{B'}$ and $[I]_B^{B'}$

(sol): $u_1 = -u'_1 + u'_2 \Rightarrow [I]_B^{B'} = \begin{bmatrix} -1 & 2 \\ 1 & -1 \end{bmatrix}$

$$\begin{aligned} u'_1 &= u_1 + u_2 \\ u'_2 &= 2u_1 + u_2 \end{aligned} \Rightarrow [I]_{B'}^B = \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix}$$

Example Given $[v]_B = \begin{bmatrix} -3 \\ 5 \end{bmatrix}$, B and B' are two bases as above

$$\begin{aligned} \text{Then, } [v]_{B'} &= [I]_{B'}^{B'} [v]_B = \begin{bmatrix} -1 & 2 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} -3 \\ 5 \end{bmatrix} \\ &= \begin{bmatrix} 13 \\ -8 \end{bmatrix} \end{aligned}$$

< An Efficient Method of Computing Transition Matrices >

$$[\text{new basis} \mid \text{old basis}] \xrightarrow{\substack{\text{row} \\ \text{operations}}} [I \mid [I]_{\text{old}}^{\text{new}}]$$

Example: Let $B = \{u_1, u_2\}$, $B' = \{u'_1, u'_2\}$ as above

$$[\text{new basis} \mid \text{old basis}] = \left[\begin{array}{cc|cc} 1 & 2 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ \hline u'_1 & u'_2 & u_1 & u_2 \end{array} \right]$$

$$\xrightarrow{\text{row operations}} [I \mid [I]_{\text{old}}^{\text{new}}] = \left[\begin{array}{cc|cc} 1 & 0 & -1 & 2 \\ 0 & 1 & 1 & -1 \end{array} \right]$$

$$\Rightarrow [I]_B^{B'} = \begin{bmatrix} -1 & 2 \\ 1 & -1 \end{bmatrix}$$

4.6.3 (PLUS/MINUS)

Theorem Let $S \neq \emptyset$ be a set of vectors in a vector space V

(a) If S is linearly independent, and $v \in V$ is outside $\text{span}(S)$, then $S \cup \{v\}$ is still linearly independent.

(b) If $v \in S$ that is a linear combination of other vectors in S , then $S - \{v\}$ spans the same space as S . That is, $\text{span}(S) = \text{span}(S - \{v\})$.

Example:

$P_1 = 1 - x^2$, $P_2 = 2 - x^2$, $P_3 = x^3$ are linearly independent?

(proof): Show that $S = \{P_1, P_2\}$ is linearly independent

② P_3 cannot be expressed as a linear combination of S

$\therefore S \cup \{P_3\} = \{P_1, P_2, P_3\}$ is a linearly independent set.

Recall

$w \in V$, $w = k_1 v_1 + k_2 v_2 + \dots + k_r v_r$

where k_1, k_2, \dots, k_r are scalars,

w is a "linear combination" of v_1, v_2, \dots, v_r

Theorem If $S = \{w_1, w_2, \dots, w_r\}$ is a nonempty set of vectors in a vector space V , then

(a) $W = \text{all possible linear combinations of the vectors in } S$
 W is a subspace of V

(b) The set W in (a) is the "smallest" subspace of V that contains all of the vectors in S

(sketch of the proof of Theorem):

$$(a) u = c_1 w_1 + c_2 w_2 + \dots + c_r w_r, v = k_1 w_1 + k_2 w_2 + \dots + k_r w_r$$

$$\textcircled{1} u+v = (c_1+k_1)w_1 + (c_2+k_2)w_2 + \dots + (c_r+k_r)w_r \rightsquigarrow \in W$$

$$\textcircled{2} \text{ Let } t \in \mathbb{R},$$

$$tu = (tc_1)w_1 + (tc_2)w_2 + \dots + (tc_r)w_r \rightsquigarrow \in W$$

(b) Let W' be any subspace of V that contains all of the vectors in S

$\because W'$ is a subspace of V

$\therefore W'$ is closed under addition and scalar multiplication

$\Rightarrow W'$ contains all linear combinations of the vectors in S

$$\Rightarrow W' \supseteq W$$

(proof of Theorem):

(a) Let $S' = \{w_1, w_2, \dots, w_m\}$, $m > n$, be any set of m vectors in V

$\because S = \{v_1, v_2, \dots, v_n\}$ is a basis of V

\therefore Let each $w_i = a_{1i}v_1 + a_{2i}v_2 + \dots + a_{ni}v_n$, for $1 \leq i \leq m$

Let $k_1 w_1 + k_2 w_2 + \dots + k_m w_m = 0 \rightarrow$ Want to show that not all k_i 's are zeros

$$[w_1 | w_2 | \dots | w_m] = [v_1 | v_2 | \dots | v_n] \begin{bmatrix} a_{11} & a_{21} & \dots & a_{n1} \\ a_{12} & a_{22} & \dots & a_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{nn} \end{bmatrix}$$

$$\therefore [w_1 | w_2 | \dots | w_m] \begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_m \end{bmatrix} = [v_1 | v_2 | \dots | v_n] \begin{bmatrix} a_{11} & a_{21} & \dots & a_{n1} \\ a_{12} & a_{22} & \dots & a_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_m \end{bmatrix} = 0$$

where $\begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_m \end{bmatrix}$ is a nontrivial solution of

$$\begin{bmatrix} a_{11} & a_{21} & \dots & a_{n1} \\ a_{12} & a_{22} & \dots & a_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1n} & a_{2n} & \dots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_m \end{bmatrix} = 0$$

(b) Exercise for students!

~~4.6.5~~ 4.6.5

Theorem Let S be a finite set of vectors in a finite-dimensional vector space V .

(a) If S spans V but is not a basis for V , then

S can be reduced to a basis for V by removing appropriate vectors from S

$\rightarrow S$ is linear dependent \rightarrow some $v \in S$ is a linear combination of the others \rightarrow remove v from S
 $\rightarrow \dots \dots$

(b) If S is a linearly independent set that is not a basis for V , then S can be enlarged to a basis for V by inserting appropriate vectors into S .

$\rightarrow S$ fails to span $V \rightarrow$ some $v \in V$ is NOT in $\text{span}(S)$

\rightarrow insert v into $S \rightarrow \dots \dots$

~~4.6.6~~ 4.6.6

Theorem If W is a subspace of a finite-dimensional vector space V , then:

(a) W is finite-dimensional.

(b) $\dim(W) \leq \dim(V)$

(c) $W = V$ if and only if $\dim(W) = \dim(V)$

(sketch of the proof of Theorem 4.6.2(b)) "m < n"

$$w_1 = a_{11}v_1 + a_{12}v_2 + \dots + a_{1n}v_n$$

$$w_2 = a_{12}v_1 + a_{22}v_2 + \dots + a_{2n}v_n$$

⋮

$$w_m = a_{1m}v_1 + a_{2m}v_2 + \dots + a_{nm}v_n$$

where $\{v_1, v_2, \dots, v_n\}$ is a basis of V

$$\Rightarrow [w_1 | w_2 | \dots | w_m] = [v_1 | v_2 | \dots | v_n] \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{bmatrix}$$

$$\text{Let } \bar{v} = b_1v_1 + b_2v_2 + \dots + b_nv_n$$

be any vector in V , $b_1, b_2, \dots, b_n \in \mathbb{R}$

We want to show that $\exists k_1, k_2, \dots, k_m \in \mathbb{R}$ such that

$$[w_1 | w_2 | \dots | w_m] \begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_m \end{bmatrix} = [v_1 | v_2 | \dots | v_n] \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$\Rightarrow [v_1 | v_2 | \dots | v_n] \begin{bmatrix} A \\ \left[\begin{array}{c|ccccc} k_1 & k_2 & \dots & k_m \\ \hline b_1 & b_2 & \dots & b_n \end{array} \right] \end{bmatrix} = [v_1 | v_2 | \dots | v_n] \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

Solve $\begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_m \end{bmatrix}$ by the augmented matrix:

$$\left[\begin{array}{c|ccccc} A & b_1 \\ \hline b_1 & b_2 \\ \vdots & \vdots \\ b_n & \end{array} \right] \rightsquigarrow \left[\begin{array}{cccc|cc} 1 & 0 & \dots & 0 & b'_1 \\ 0 & 1 & \dots & 0 & b'_2 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & b'_m \end{array} \right] \Rightarrow \text{no solution for } k_1, k_2, \dots, k_m$$

! unique representation
of basis for any vector!

let it $b_n \neq 0$