

SC223 - Linear Algebra

Aditya Tatu

Lecture 22



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Summary: Lecture 21

- Let $U \subset V$. U is said to **span** V if $\text{span}(U) = V$
- $W = \{v_1, \dots, v_n\} \subset V$ is a set of linear independent vectors, if

$$a_1 v_1 + \dots + a_n v_n = \theta \Rightarrow a_i = 0, i = 1, \dots, n$$

- **Proposition 11:** For any FDVS, the number of vectors in a linearly independent set of vectors cannot be more than the number of vectors in a spanning set.

Basis of a Vector space

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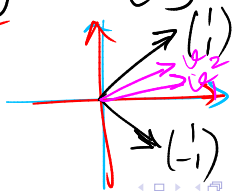
● Examples:

$$1. V = \mathbb{R}^2, \beta_1 = \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right\}. \beta_2 = \left\{ \begin{pmatrix} 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 2 \end{pmatrix} \right\}$$

$$\beta_3 = \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}.$$

$$\begin{bmatrix} x \\ y \end{bmatrix} = k_1 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} + k_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} \Rightarrow \begin{matrix} k_2 = y \\ k_1 = x - y \end{matrix}$$

$$\beta_4 = \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right\}.$$

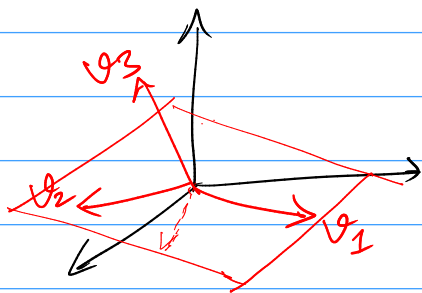


② $V = \mathbb{R}^3$

$$\beta_1 = \left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right\} \quad \beta_2 = \left\{ \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} \right\}$$

$$\begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} k_1 \\ k_2 \\ k_3 \end{bmatrix} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

\mathbb{R}^3



④ $V = \mathcal{P}(\mathbb{R})$

$$\beta_1 = \{1, x, x^2, x^3, \dots\}$$

$$\beta_2 = \{1, x+x^2, x^2, x^3, \dots\}$$

$$1 \cdot (x+x^2) - 1 \cdot x^2 = x$$

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⑤ $V = \mathbb{R}^{n \times n}$

$$U = \left\{ \begin{bmatrix} h_0 & h_1 & \dots & h_{n-1} \\ h_{n-1} & h_0 & h_1 & \dots & h_{n-2} \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ h_1 & h_2 & \dots & h_{n-1} & h_0 \end{bmatrix} \mid \forall h_0, h_1, \dots, h_{n-1} \in \mathbb{R} \right\}$$

$$I_n = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix}$$

$$D = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & 0 & \dots & 0 & 0 \end{bmatrix}$$

$$D^2 = \begin{bmatrix} 0 & 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 1 & 0 & \dots & 0 & 0 \end{bmatrix}$$

Any $H \in U$, $H = h_0 I + h_1 D + h_2 D^2 + \dots + h_{n-1} D^{n-1}$

$H \in \mathbb{R}^{n \times n}$, $x \in \mathbb{R}^n$

$$(Hx)(k) = ((h_0 I + h_1 D + \dots + h_{n-1} D^{n-1})x)(k)$$

$$= (h_0 x + h_1 D x + \dots + h_{n-1} D^{n-1} x)(k)$$

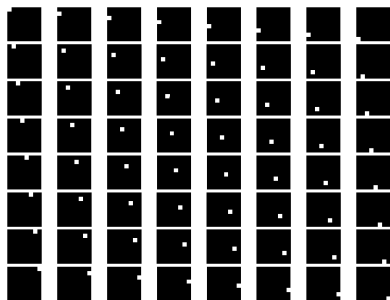
$$(D x)(k) = x[(k+1) \bmod N].$$

$$= h_0 x(k) + h_1 x[(k+1) \bmod N] + \dots + h_{n-1} x[(k+n-1) \bmod N]$$

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● Examples: $V = \mathbb{R}^{8 \times 8}$



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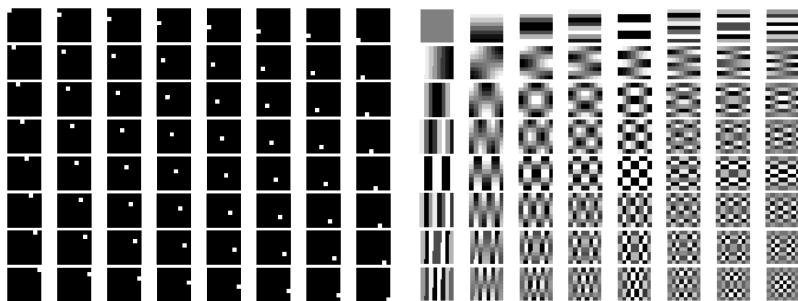


Figure: (left) Standard basis, (right) 2D-DCT basis for $\mathbb{R}^{8 \times 8}$

- For speech/audio/1D signals:

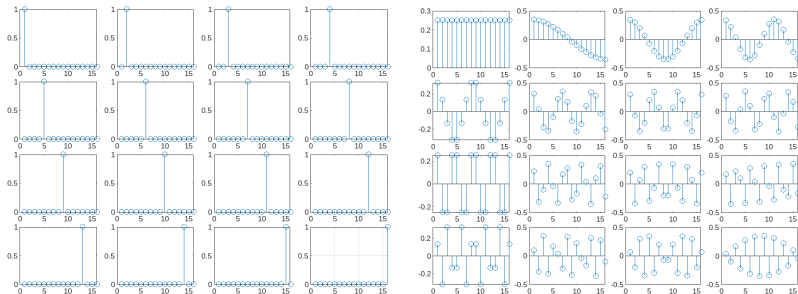


Figure: (left) Standard Basis, (right) DCT basis for \mathbb{R}^{16}

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$$A: \mathbb{R}^n \longrightarrow \mathbb{R}^m$$

