

The preceding bilinear form is a special case of the next example.

Example 2 Let  $V = F^n$ , where the vectors are considered as column vectors. For any  $A \in M_{n \times n}(F)$ , define  $H: V \times V \rightarrow F$

by  $H(x, y) = x^t A y$ , for  $x, y \in V$ . Notice that  $H(x, y)$  is a  $1 \times 1$  matrix. We identify this matrix with its single entry.

The bilinearity of  $H$  follows as in example 1. For example, for  $a \in F$ , and  $x_1, x_2, y \in V$ , we have

$$\begin{aligned} H(ax_1 + x_2, y) &= (ax_1 + x_2)^t A y = (ax_1^t + x_2^t) A y \\ &= (ax_1^t) A y + x_2^t A y = a(x_1^t A y) + x_2^t A y = aH(x_1, y) + H(x_2, y) \end{aligned}$$

Similarly, we can show  $H(x, ay_1 + y_2) = aH(x, y_1) + H(x, y_2)$

Now we list several properties possessed by all bilinear forms.

For any bilinear form  $H$  on a vector space  $V$  over a field  $F$ , the following properties hold:

1. If, for any  $x \in V$ , the functions  $L_x, R_x: V \rightarrow F$  are defined by  $L_x(y) = H(x, y)$  and  $R_x(y) = H(y, x)$  for all  $y \in V$ , then  $L_x$  and  $R_x$  are linear.

Proof: Let  $a \in F$  and  $y_1, y_2 \in V$ . Then,

$$\begin{aligned} L_x(ay_1 + y_2) &= H(x, ay_1 + y_2) = aH(x, y_1) + \underbrace{H(x, y_2)}_{H(x, y_2)} \quad (\text{As } H \text{ is bilinear}) \\ &= aL_x(y_1) + L_x(y_2) \end{aligned}$$

So,  $L_x$  is linear. Similarly,  $R_x$  is linear.

2.  $H(\theta, x) = H(x, \theta) = 0$  for all  $x \in V$

Proof: Exercise.

3. For all  $x, y, z, w \in V$ ,

$$H(x+y, z+w) = H(x, z) + H(x, w) + H(y, z) + H(y, w)$$

Proof:  $H(x+y, z+w) = H(x, z+w) + H(y, z+w)$   
 $= H(x, z) + H(x, w) + H(y, z) + H(y, w)$  (As  $H$  is bilinear)

4. If  $J: V \times V \rightarrow F$  be defined by  $J(x, y) = H(y, x)$ , then  $J$  is a bilinear form.

Proof: Exercise.

Definition 4.4.2. Let  $V$  be a vector space, let  $H_1$  and  $H_2$  be two bilinear forms on  $V$ , and let  $a$  be a scalar. We define the sum  $H_1 + H_2$  and the scalar product  $aH_1$  by the equations  $(H_1 + H_2)(x, y) = H_1(x, y) + H_2(x, y)$  and

$$(aH_1)(x, y) = a(H_1(x, y)) \quad \text{for all } x, y \in V$$

The following theorem is an immediate consequence of the definition:

Theorem 4.4.3 For any vector space  $V$ , the sum of two bilinear forms and the product of a scalar and a bilinear form on  $V$  are again bilinear forms on  $V$ . Furthermore the set  $\mathcal{B}(V) = \{H: H \text{ is a bilinear form on } V\}$  is a vector space with respect to these operations.

Proof: Exercise.

Let  $\beta = \{v_1, v_2, \dots, v_n\}$  be an order basis for an  $n$ -dimensional vector space  $V$ , let  $H \in \mathcal{B}(V)$ . We can associate with  $H$  an  $n \times n$  matrix  $A = (a_{ij})_{n \times n}$  where  $a_{ij} = H(v_i, v_j)$ ,  $i, j = 1, 2, \dots, n$

**Definition 4.4.4** The matrix  $A$  defined earlier is called the matrix representation of  $H$  with respect to the ordered basis  $\beta$  and is denoted by  $\Psi_\beta(H)$ .

We can therefore regard  $\Psi_\beta$  as a mapping from  $\mathcal{B}(V)$  to  $M_{n \times n}(F)$ , where  $F$  is the field of scalars for  $V$ , that takes a bilinear form  $H$  into its matrix representation  $\Psi_\beta(H)$ .

We first consider an example and then show that  $\Psi_\beta$  is an isomorphism.

**Example 3:** Consider the bilinear form  $H$  of Example 1 (Page-52),

and let  $\beta = \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\}$  and  $B = \Psi_\beta(H) = [b_{ij}]_{2 \times 2}$

$$\text{Then } b_{11} = H \left( \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right) = 2 + 3 + 4 - 1 = 8$$

$$b_{12} = H \left( \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right) = 2 - 3 + 4 + 1 = 4$$

$$b_{21} = H \left( \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right) = 2 + 3 - 4 + 1 = 2$$

$$b_{22} = H \left( \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right) = 2 - 3 - 4 - 1 = -6$$

$$\text{So, } \Psi_\beta(H) = \begin{pmatrix} 8 & 4 \\ 2 & -6 \end{pmatrix}$$

If  $\gamma$  be the standard ordered basis for  $\mathbb{R}^2$ , ~~you can~~ the student can verify that  $\Psi_\gamma(H) = \begin{pmatrix} 2 & 3 \\ 4 & -1 \end{pmatrix}$

**Theorem 4.4.5** For any  $n$ -dimensional vector space  $V$  over  $F$  and any ordered basis  $\beta$  for  $V$ ,  $\Psi_\beta$  is an isomorphism.

**Proof:** Let  $\beta = \{v_1, v_2, \dots, v_n\}$ . Let  $H_1, H_2 \in \mathcal{B}(V)$  and  $a \in F$ . Then  $\Psi_\beta(aH_1 + H_2) = A$  where  $a_{ij} = (aH_1 + H_2)(v_i, v_j)$

$$\text{So, } a_{ij} = (aH_1 + H_2)(v_i, v_j) = a(H_1(v_i, v_j)) + H_2(v_i, v_j)$$

$$\text{So, } a_{ij} = a b_{ij} + c_{ij} \text{ where } \psi_\beta(H_1) = B = (b_{ij})_{n \times n} \text{ and } \psi_\beta(H_2) = C = (c_{ij})_{n \times n}$$

$$\text{So, } A = aB + C$$

$$\text{So, } \psi_\beta(aH_1 + H_2) = a\psi_\beta(H_1) + \psi_\beta(H_2)$$

So,  $\psi_\beta$  is linear.

To show  $\psi_\beta$  is one-to-one or injective, suppose that

$$\psi_\beta(H) = O_{n \times n} \text{ where } O_{n \times n} \text{ is the null matrix of order } n$$

for some  $H \in \mathcal{B}(V)$ . Fix  $v_i \in \beta$  and recall the mapping

$L_{v_i}: V \rightarrow F$  is linear by property 1 of page-53. By

hypothesis  $L_{v_i}(v_j) = H(v_i, v_j) = 0$  for all  $v_j \in \beta$ . Hence

$L_{v_i}$  is the zero linear transformation from  $V$  to  $F$ . So,

$$H(v_i, x) = L_{v_i}(x) = 0 \text{ for all } x \in V \text{ and } v_i \in \beta. \dots (1)$$

Next fix an arbitrary  $y \in V$ , and recall the linear mapping

$R_y: V \rightarrow F$  defined in the property 1 on page-53

By (1),  $R_y(v_i) = H(v_i, y) = 0$  for all  $v_i \in \beta$  and hence

$R_y$  is the zero linear transformation,

$$\text{So, } H(x, y) = R_y(x) = 0 \text{ for all } x, y \in V. \text{ So } H$$

is the zero bilinear form, and so  $\psi_\beta$  is one-to-one.

To show that  $\psi_\beta$  is onto, consider any  $A \in M_{n \times n}(F)$

Now, we recall that  $\phi_\beta$  the isomorphism  $\phi_\beta: V \rightarrow F^n$

defined by  $\phi_\beta(x) = [x]_\beta$  where  $[x]_\beta$  is the