

# Linear Algebra 2

## Lecture #24

### Bilinear and quadratic forms

Irena Penev

May 15, 2024

- This lecture has four parts:

- This lecture has four parts:
  - ① A formula for products of the form  $\mathbf{x}^T \mathbf{A} \mathbf{y}$

- This lecture has four parts:
  - ① A formula for products of the form  $\mathbf{x}^T \mathbf{A} \mathbf{y}$
  - ② Bilinear forms

- This lecture has four parts:
  - ① A formula for products of the form  $\mathbf{x}^T \mathbf{A} \mathbf{y}$
  - ② Bilinear forms
  - ③ Quadratic forms

- This lecture has four parts:
  - ① A formula for products of the form  $\mathbf{x}^T \mathbf{A} \mathbf{y}$
  - ② Bilinear forms
  - ③ Quadratic forms
  - ④ Quadratic forms on  $\mathbb{R}^n$

- 1 A formula for products of the form  $\mathbf{x}^T \mathbf{A} \mathbf{y}$

- ① A formula for products of the form  $\mathbf{x}^T \mathbf{A} \mathbf{y}$

### Proposition 9.1.1

Let  $\mathbb{F}$  be a field, let  $\mathcal{E}_n = \{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  be the standard basis of  $\mathbb{F}^n$ , and let  $A = [a_{i,j}]_{n \times n}$  be a matrix in  $\mathbb{F}^{n \times n}$ . Then both the following hold:

- Ⓐ for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  and  $\mathbf{y} = [y_1 \ \dots \ y_n]^T$  in  $\mathbb{F}^n$ , we have that

$$\mathbf{x}^T \mathbf{A} \mathbf{y} = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i y_j;$$

- Ⓑ for all indices  $i, j \in \{1, \dots, n\}$ , we have that  $\mathbf{e}_i^T \mathbf{A} \mathbf{e}_j = a_{i,j}$ .

*Proof.*



- ① A formula for products of the form  $\mathbf{x}^T \mathbf{A} \mathbf{y}$

### Proposition 9.1.1

Let  $\mathbb{F}$  be a field, let  $\mathcal{E}_n = \{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  be the standard basis of  $\mathbb{F}^n$ , and let  $A = [a_{i,j}]_{n \times n}$  be a matrix in  $\mathbb{F}^{n \times n}$ . Then both the following hold:

- Ⓐ for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  and  $\mathbf{y} = [y_1 \ \dots \ y_n]^T$  in  $\mathbb{F}^n$ , we have that

$$\mathbf{x}^T \mathbf{A} \mathbf{y} = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i y_j;$$

- Ⓑ for all indices  $i, j \in \{1, \dots, n\}$ , we have that  $\mathbf{e}_i^T \mathbf{A} \mathbf{e}_j = a_{i,j}$ .

*Proof.* Obviously, (a) implies (b). So, let us prove (a).

*Proof (continued).*

*Proof (continued).* For any vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  and  $\mathbf{y} = [y_1 \ \dots \ y_n]^T$  in  $\mathbb{F}^n$ , we have the following:

$$\begin{aligned}\mathbf{x}^T \mathbf{A} \mathbf{y} &= [x_1 \ x_2 \ \dots \ x_n] \begin{bmatrix} a_{1,1} & a_{1,2} & \dots & a_{1,n} \\ a_{2,1} & a_{2,2} & \dots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n,1} & a_{n,2} & \dots & a_{n,n} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \\ &= [x_1 \ x_2 \ \dots \ x_n] \begin{bmatrix} \sum_{j=1}^n a_{1,j} y_j \\ \sum_{j=1}^n a_{2,j} y_j \\ \vdots \\ \sum_{j=1}^n a_{n,j} y_j \end{bmatrix} \\ &= \sum_{i=1}^n x_i \left( \sum_{j=1}^n a_{i,j} y_j \right) = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i y_j.\end{aligned}$$

This proves (a).  $\square$

### Proposition 9.1.1

Let  $\mathbb{F}$  be a field, let  $\mathcal{E}_n = \{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  be the standard basis of  $\mathbb{F}^n$ , and let  $A = [a_{i,j}]_{n \times n}$  be a matrix in  $\mathbb{F}^{n \times n}$ . Then both the following hold:

- Ⓐ for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  and  $\mathbf{y} = [y_1 \ \dots \ y_n]^T$  in  $\mathbb{F}^n$ , we have that

$$\mathbf{x}^T \mathbf{A} \mathbf{y} = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i y_j;$$

- Ⓑ for all indices  $i, j \in \{1, \dots, n\}$ , we have that  $\mathbf{e}_i^T \mathbf{A} \mathbf{e}_j = a_{i,j}$ .

## ② Bilinear forms

## 2 Bilinear forms

### Definition

A *bilinear form* on a vector space  $V$  over a field  $\mathbb{F}$  is a function  $f : V \times V \rightarrow \mathbb{F}$  that satisfies the following four axioms:

- b.1.  $\forall \mathbf{x}_1, \mathbf{x}_2, \mathbf{y} \in V: f(\mathbf{x}_1 + \mathbf{x}_2, \mathbf{y}) = f(\mathbf{x}_1, \mathbf{y}) + f(\mathbf{x}_2, \mathbf{y});$
- b.2.  $\forall \mathbf{x}, \mathbf{y} \in V$  and  $\alpha \in \mathbb{F}: f(\alpha \mathbf{x}, \mathbf{y}) = \alpha f(\mathbf{x}, \mathbf{y});$
- b.3.  $\forall \mathbf{x}, \mathbf{y}_1, \mathbf{y}_2 \in V: f(\mathbf{x}, \mathbf{y}_1 + \mathbf{y}_2) = f(\mathbf{x}, \mathbf{y}_1) + f(\mathbf{x}, \mathbf{y}_2);$
- b.4.  $\forall \mathbf{x}, \mathbf{y} \in V, \alpha \in \mathbb{F}: f(\mathbf{x}, \alpha \mathbf{y}) = \alpha f(\mathbf{x}, \mathbf{y}).$

The bilinear form  $f$  is said to be *symmetric* if it further satisfies the property that  $f(\mathbf{x}, \mathbf{y}) = f(\mathbf{y}, \mathbf{x})$  for all  $\mathbf{x}, \mathbf{y} \in V$ .

- Reminder:

## Definition

A *scalar product* (also called *inner product*) in a **real** vector space  $V$  is a function  $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$  that satisfies the following four axioms:

r.1.  $\forall \mathbf{x} \in V: \langle \mathbf{x}, \mathbf{x} \rangle \geq 0$ , and equality holds iff  $\mathbf{x} = \mathbf{0}$ ;

r.2.  $\forall \mathbf{x}, \mathbf{y}, \mathbf{z} \in V: \langle \mathbf{x} + \mathbf{y}, \mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{z} \rangle + \langle \mathbf{y}, \mathbf{z} \rangle$ ;

r.3.  $\forall \mathbf{x}, \mathbf{y} \in V, \alpha \in \mathbb{R}: \langle \alpha \mathbf{x}, \mathbf{y} \rangle = \alpha \langle \mathbf{x}, \mathbf{y} \rangle$ ;

r.4.  $\forall \mathbf{x}, \mathbf{y} \in V: \langle \mathbf{x}, \mathbf{y} \rangle = \langle \mathbf{y}, \mathbf{x} \rangle$ .

r.2'.  $\forall \mathbf{x}, \mathbf{y}, \mathbf{z} \in V, \langle \mathbf{x}, \mathbf{y} + \mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{z} \rangle$ ;

r.3'.  $\forall \mathbf{x}, \mathbf{y} \in V$  and  $\alpha \in \mathbb{R}, \langle \mathbf{x}, \alpha \mathbf{y} \rangle = \alpha \langle \mathbf{x}, \mathbf{y} \rangle$ .

- Reminder:

### Definition

A *scalar product* (also called *inner product*) in a **real** vector space  $V$  is a function  $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{R}$  that satisfies the following four axioms:

r.1.  $\forall \mathbf{x} \in V: \langle \mathbf{x}, \mathbf{x} \rangle \geq 0$ , and equality holds iff  $\mathbf{x} = \mathbf{0}$ ;

r.2.  $\forall \mathbf{x}, \mathbf{y}, \mathbf{z} \in V: \langle \mathbf{x} + \mathbf{y}, \mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{z} \rangle + \langle \mathbf{y}, \mathbf{z} \rangle$ ;

r.3.  $\forall \mathbf{x}, \mathbf{y} \in V, \alpha \in \mathbb{R}: \langle \alpha \mathbf{x}, \mathbf{y} \rangle = \alpha \langle \mathbf{x}, \mathbf{y} \rangle$ ;

r.4.  $\forall \mathbf{x}, \mathbf{y} \in V: \langle \mathbf{x}, \mathbf{y} \rangle = \langle \mathbf{y}, \mathbf{x} \rangle$ .

r.2'.  $\forall \mathbf{x}, \mathbf{y}, \mathbf{z} \in V, \langle \mathbf{x}, \mathbf{y} + \mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{z} \rangle$ ;

r.3'.  $\forall \mathbf{x}, \mathbf{y} \in V$  and  $\alpha \in \mathbb{R}, \langle \mathbf{x}, \alpha \mathbf{y} \rangle = \alpha \langle \mathbf{x}, \mathbf{y} \rangle$ .

- **Remark:** every scalar product  $\langle \cdot, \cdot \rangle$  in a **real** vector space  $V$  is a symmetric bilinear form.
  - Indeed, r.2, r.3, r.2', and r.3' are precisely the axioms b.1, b.2, b.3, and b.4, respectively.
  - Moreover, by r.4, scalar products in real vector spaces are symmetric.



- Reminder:

### Definition

A *scalar product* (also called *inner product*) in a **complex** vector space  $V$  is a function  $\langle \cdot, \cdot \rangle : V \times V \rightarrow \mathbb{C}$  that satisfies the following four axioms:

c.1.  $\forall \mathbf{x} \in V$ :  $\langle \mathbf{x}, \mathbf{x} \rangle$  is a real number,  $\langle \mathbf{x}, \mathbf{x} \rangle \geq 0$ , and equality holds iff  $\mathbf{x} = \mathbf{0}$ ;

c.2.  $\forall \mathbf{x}, \mathbf{y}, \mathbf{z} \in V$ :  $\langle \mathbf{x} + \mathbf{y}, \mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{z} \rangle + \langle \mathbf{y}, \mathbf{z} \rangle$ ;

c.3.  $\forall \mathbf{x}, \mathbf{y} \in V, \alpha \in \mathbb{C}$ :  $\langle \alpha \mathbf{x}, \mathbf{y} \rangle = \alpha \langle \mathbf{x}, \mathbf{y} \rangle$ ;

c.4.  $\forall \mathbf{x}, \mathbf{y} \in V$ :  $\langle \mathbf{x}, \mathbf{y} \rangle = \overline{\langle \mathbf{y}, \mathbf{x} \rangle}$ .

c.2'.  $\forall \mathbf{x}, \mathbf{y}, \mathbf{z} \in V$ :  $\langle \mathbf{x}, \mathbf{y} + \mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{y} \rangle + \langle \mathbf{x}, \mathbf{z} \rangle$ ;

c.3'.  $\forall \mathbf{x}, \mathbf{y} \in V, \alpha \in \mathbb{C}$ :  $\langle \mathbf{x}, \alpha \mathbf{y} \rangle = \bar{\alpha} \langle \mathbf{x}, \mathbf{y} \rangle$ .

- **Remark:** scalar products in non-trivial **complex** vector spaces are not bilinear forms, since c.1 and c.3' together contradict axiom b.4 (next slide).

c.1.  $\forall \mathbf{x} \in V$ :  $\langle \mathbf{x}, \mathbf{x} \rangle$  is a real number,  $\langle \mathbf{x}, \mathbf{x} \rangle \geq 0$ , and equality holds iff  $\mathbf{x} = \mathbf{0}$ ;

c.3'.  $\forall \mathbf{x}, \mathbf{y} \in V, \alpha \in \mathbb{C}$ :  $\langle \mathbf{x}, \alpha \mathbf{y} \rangle = \bar{\alpha} \langle \mathbf{x}, \mathbf{y} \rangle$ .

- Indeed, if  $\langle \cdot, \cdot \rangle$  is a scalar product in a non-trivial complex vector space  $V$ , then for any  $\mathbf{x} \in V \setminus \{\mathbf{0}\}$ , c.1 guarantees that  $\langle \mathbf{x}, \mathbf{x} \rangle \neq 0$ ,

c.1.  $\forall \mathbf{x} \in V$ :  $\langle \mathbf{x}, \mathbf{x} \rangle$  is a real number,  $\langle \mathbf{x}, \mathbf{x} \rangle \geq 0$ , and equality holds iff  $\mathbf{x} = \mathbf{0}$ ;

c.3'.  $\forall \mathbf{x}, \mathbf{y} \in V, \alpha \in \mathbb{C}$ :  $\langle \mathbf{x}, \alpha \mathbf{y} \rangle = \bar{\alpha} \langle \mathbf{x}, \mathbf{y} \rangle$ .

- Indeed, if  $\langle \cdot, \cdot \rangle$  is a scalar product in a non-trivial complex vector space  $V$ , then for any  $\mathbf{x} \in V \setminus \{\mathbf{0}\}$ , c.1 guarantees that  $\langle \mathbf{x}, \mathbf{x} \rangle \neq 0$ , and so

$$\langle \mathbf{x}, i\mathbf{x} \rangle \stackrel{\text{c.3}'}{=} \bar{i} \langle \mathbf{x}, \mathbf{x} \rangle = -i \langle \mathbf{x}, \mathbf{x} \rangle \neq i \langle \mathbf{x}, \mathbf{x} \rangle,$$

and we see that b.4 does not hold.

### Proposition 9.2.1

Let  $V$  be a vector space over a field  $\mathbb{F}$ , and let  $f$  be a bilinear form on  $V$ . Then all the following hold:

- a)  $\forall \mathbf{x} \in V: f(\mathbf{x}, \mathbf{0}) = 0;$
- b)  $\forall \mathbf{y} \in V: f(\mathbf{0}, \mathbf{y}) = 0;$
- c)  $f(\mathbf{0}, \mathbf{0}) = 0.$

*Proof.*

### Proposition 9.2.1

Let  $V$  be a vector space over a field  $\mathbb{F}$ , and let  $f$  be a bilinear form on  $V$ . Then all the following hold:

- Ⓐ  $\forall \mathbf{x} \in V: f(\mathbf{x}, \mathbf{0}) = 0;$
- Ⓑ  $\forall \mathbf{y} \in V: f(\mathbf{0}, \mathbf{y}) = 0;$
- Ⓒ  $f(\mathbf{0}, \mathbf{0}) = 0.$

*Proof.* For (a), we fix a vector  $\mathbf{x} \in V$ , and we compute:

$$f(\mathbf{x}, \mathbf{0}) = f(\mathbf{x}, \mathbf{0} + \mathbf{0}) \stackrel{\text{b.3}}{=} f(\mathbf{x}, \mathbf{0}) + f(\mathbf{x}, \mathbf{0}).$$

By subtracting  $f(\mathbf{x}, \mathbf{0})$  from both sides, we obtain  $\mathbf{0} = f(\mathbf{x}, \mathbf{0})$ .

This proves (a).

### Proposition 9.2.1

Let  $V$  be a vector space over a field  $\mathbb{F}$ , and let  $f$  be a bilinear form on  $V$ . Then all the following hold:

- Ⓐ  $\forall \mathbf{x} \in V: f(\mathbf{x}, \mathbf{0}) = 0;$
- Ⓑ  $\forall \mathbf{y} \in V: f(\mathbf{0}, \mathbf{y}) = 0;$
- Ⓒ  $f(\mathbf{0}, \mathbf{0}) = 0.$

*Proof.* For (a), we fix a vector  $\mathbf{x} \in V$ , and we compute:

$$f(\mathbf{x}, \mathbf{0}) = f(\mathbf{x}, \mathbf{0} + \mathbf{0}) \stackrel{\text{b.3}}{=} f(\mathbf{x}, \mathbf{0}) + f(\mathbf{x}, \mathbf{0}).$$

By subtracting  $f(\mathbf{x}, \mathbf{0})$  from both sides, we obtain  $\mathbf{0} = f(\mathbf{x}, \mathbf{0})$ .

This proves (a).

The proof of (b) is similar.

### Proposition 9.2.1

Let  $V$  be a vector space over a field  $\mathbb{F}$ , and let  $f$  be a bilinear form on  $V$ . Then all the following hold:

- (a)  $\forall \mathbf{x} \in V: f(\mathbf{x}, \mathbf{0}) = 0$ ;
- (b)  $\forall \mathbf{y} \in V: f(\mathbf{0}, \mathbf{y}) = 0$ ;
- (c)  $f(\mathbf{0}, \mathbf{0}) = 0$ .

*Proof.* For (a), we fix a vector  $\mathbf{x} \in V$ , and we compute:

$$f(\mathbf{x}, \mathbf{0}) = f(\mathbf{x}, \mathbf{0} + \mathbf{0}) \stackrel{\text{b.3}}{=} f(\mathbf{x}, \mathbf{0}) + f(\mathbf{x}, \mathbf{0}).$$

By subtracting  $f(\mathbf{x}, \mathbf{0})$  from both sides, we obtain  $\mathbf{0} = f(\mathbf{x}, \mathbf{0})$ .

This proves (a).

The proof of (b) is similar. Finally, (c) is a special case of (a) for  $\mathbf{x} = \mathbf{0}$ .  $\square$

- Reminder:

### Theorem 4.5.1

Let  $U$  and  $V$  be non-trivial, finite-dimensional vector spaces over a field  $\mathbb{F}$ . Let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_m\}$  be a basis of  $U$ , let  $\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}$  be a basis of  $V$ , and let  $f : U \rightarrow V$  be a linear function. Then exists a unique matrix in  $\mathbb{F}^{n \times m}$ , denoted by  ${}_c [ f ]_B$  and called the *matrix of  $f$  with respect to  $\mathcal{B}$  and  $\mathcal{C}$* , s.t. for all  $\mathbf{u} \in U$ , we have that

$${}_c [ f ]_B [ \mathbf{u} ]_B = [ f(\mathbf{u}) ]_C.$$

Moreover, the matrix  ${}_c [ f ]_B$  is given by

$${}_c [ f ]_B = [ [ f(\mathbf{b}_1) ]_C \quad \dots \quad [ f(\mathbf{b}_m) ]_C ].$$



- Reminder:

### Theorem 4.5.1

Let  $U$  and  $V$  be non-trivial, finite-dimensional vector spaces over a field  $\mathbb{F}$ . Let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_m\}$  be a basis of  $U$ , let  $\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}$  be a basis of  $V$ , and let  $f : U \rightarrow V$  be a linear function. Then exists a unique matrix in  $\mathbb{F}^{n \times m}$ , denoted by  ${}_c[f]_{\mathcal{B}}$  and called the *matrix of  $f$  with respect to  $\mathcal{B}$  and  $\mathcal{C}$* , s.t. for all  $\mathbf{u} \in U$ , we have that

$${}_c[f]_{\mathcal{B}} [\mathbf{u}]_{\mathcal{B}} = [f(\mathbf{u})]_{\mathcal{C}}.$$

Moreover, the matrix  ${}_c[f]_{\mathcal{B}}$  is given by

$${}_c[f]_{\mathcal{B}} = \left[ [f(\mathbf{b}_1)]_{\mathcal{C}} \quad \dots \quad [f(\mathbf{b}_m)]_{\mathcal{C}} \right].$$

- For bilinear forms, we have the following (next slide).

## Theorem 9.2.2

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  be a basis of  $V$ .

- (a) For every matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , the function  $f : V \times V \rightarrow \mathbb{F}$  given by

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V$$

is a bilinear form on  $V$ , and moreover, all the following hold:

- (a.1)  $f(\mathbf{b}_i, \mathbf{b}_j) = a_{i,j}$  for all  $i, j \in \{1, \dots, n\}$ ,  
(a.2)  $f\left(\sum_{i=1}^n c_i \mathbf{b}_i, \sum_{j=1}^n d_j \mathbf{b}_j\right) = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} c_i d_j$  for all  $c_1, \dots, c_n, d_1, \dots, d_n \in \mathbb{F}$ ,  
(a.3)  $f$  is symmetric if and only if  $A$  is symmetric.

- (b) For every bilinear form  $f$  on  $V$ , there exists a unique matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , called the *matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$* , that satisfies the property that

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

Moreover, the entries of the matrix  $A$  are given by

$$a_{i,j} = f(\mathbf{b}_i, \mathbf{b}_j) \text{ for all indices } i, j \in \{1, \dots, n\}.$$

*Proof.* (a) Fix a matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , and define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

*Proof.* (a) Fix a matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , and define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

Let us first check that  $f$  is bilinear.

*Proof.* (a) Fix a matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , and define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

Let us first check that  $f$  is bilinear. We must check that  $f$  satisfies axioms b.1-b.4.

*Proof.* (a) Fix a matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , and define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

Let us first check that  $f$  is bilinear. We must check that  $f$  satisfies axioms b.1-b.4. For b.1, we observe that for all vectors  $\mathbf{x}_1, \mathbf{x}_2, \mathbf{y} \in V$ , we have the following:

$$\begin{aligned} f(\mathbf{x}_1 + \mathbf{x}_2, \mathbf{y}) &= [\mathbf{x}_1 + \mathbf{x}_2]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \\ &\stackrel{(*)}{=} \left( [\mathbf{x}_1]_{\mathcal{B}} + [\mathbf{x}_2]_{\mathcal{B}} \right)^T A [\mathbf{y}]_{\mathcal{B}} \\ &= \left( [\mathbf{x}_1]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \right) + \left( [\mathbf{x}_2]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \right) \\ &= f(\mathbf{x}_1, \mathbf{y}) + f(\mathbf{x}_2, \mathbf{y}), \end{aligned}$$

where (\*) follows from the linearity of  $[\cdot]_{\mathcal{B}}$ . Thus,  $f$  satisfies b.1, and similarly, it satisfies b.3.

*Proof (continued).* For b.2, we observe that for all vectors  $\mathbf{x}, \mathbf{y} \in V$  and scalars  $\alpha \in V$ , we have the following:

$$\begin{aligned} f(\alpha\mathbf{x}, \mathbf{y}) &= [\alpha\mathbf{x}]_B^T A [\mathbf{y}]_B \\ &\stackrel{(*)}{=} (\alpha [\mathbf{x}]_B)^T A [\mathbf{y}]_B \\ &= \alpha \left( [\mathbf{x}]_B^T A [\mathbf{y}]_B \right) \\ &= \alpha f(\mathbf{x}, \mathbf{y}), \end{aligned}$$

where (\*) follows from the linearity of  $[\cdot]_B$ . Thus,  $f$  satisfies b.2, and similarly, it satisfies b.4. This proves that  $f$  is indeed bilinear.

*Proof (continued).* Next, to prove (a.1), we fix indices  $i, j \in \{1, \dots, n\}$ , and we compute:

$$f(\mathbf{b}_i, \mathbf{b}_j) = [\mathbf{b}_i]_{\mathcal{B}}^T A [\mathbf{b}_j]_{\mathcal{B}} = \mathbf{e}_i^T A \mathbf{e}_j \stackrel{(*)}{=} a_{i,j},$$

where (\*) follows from Proposition 9.1.1(b).



*Proof (continued).* Next, to prove (a.1), we fix indices  $i, j \in \{1, \dots, n\}$ , and we compute:

$$f(\mathbf{b}_i, \mathbf{b}_j) = [\mathbf{b}_i]_{\mathcal{B}}^T A [\mathbf{b}_j]_{\mathcal{B}} = \mathbf{e}_i^T A \mathbf{e}_j \stackrel{(*)}{=} a_{i,j},$$

where (\*) follows from Proposition 9.1.1(b).

For (a.2), we fix scalars  $c_1, \dots, c_n, d_1, \dots, d_n \in \mathbb{F}$ , and we compute:

$$f\left(\sum_{i=1}^n c_i \mathbf{b}_i, \sum_{j=1}^n d_j \mathbf{b}_j\right) \stackrel{(*)}{=} \sum_{i=1}^n \sum_{j=1}^n c_i d_j f(\mathbf{b}_i, \mathbf{b}_j) \stackrel{(a.1)}{=} \sum_{i=1}^n \sum_{j=1}^n a_{i,j} c_i d_j,$$

where (\*) follows from the fact that  $f$  is bilinear.

*Proof (continued).* It remains to prove (a.3).

*Proof (continued).* It remains to prove (a.3). Suppose first that  $A$  is symmetric. Then for all  $\mathbf{x}, \mathbf{y} \in V$ , we have that

$$\begin{aligned} f(\mathbf{x}, \mathbf{y}) &= [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \stackrel{(*)}{=} \left( [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \right)^T \\ &= [\mathbf{y}]_{\mathcal{B}}^T A^T [\mathbf{x}]_{\mathcal{B}} \stackrel{(**)}{=} [\mathbf{y}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}} = f(\mathbf{y}, \mathbf{x}), \end{aligned}$$

where (\*) follows from the fact that  $[\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}}$  is a  $1 \times 1$  matrix (and is therefore symmetric), and (\*\*) follows from the fact that  $A$  is symmetric. So,  $f$  is symmetric.

*Proof (continued).* It remains to prove (a.3). Suppose first that  $A$  is symmetric. Then for all  $\mathbf{x}, \mathbf{y} \in V$ , we have that

$$\begin{aligned} f(\mathbf{x}, \mathbf{y}) &= [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \stackrel{(*)}{=} \left( [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \right)^T \\ &= [\mathbf{y}]_{\mathcal{B}}^T A^T [\mathbf{x}]_{\mathcal{B}} \stackrel{(**)}{=} [\mathbf{y}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}} = f(\mathbf{y}, \mathbf{x}), \end{aligned}$$

where  $(*)$  follows from the fact that  $[\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}}$  is a  $1 \times 1$  matrix (and is therefore symmetric), and  $(**)$  follows from the fact that  $A$  is symmetric. So,  $f$  is symmetric.

Suppose, conversely, that  $f$  is symmetric. Then for all indices  $i, j \in \{1, \dots, n\}$ , we have the following:

$$a_{i,j} \stackrel{(a.1)}{=} f(\mathbf{b}_i, \mathbf{b}_j) \stackrel{(*)}{=} f(\mathbf{b}_j, \mathbf{b}_i) \stackrel{(a.1)}{=} a_{j,i},$$

where  $(*)$  follows from the fact that  $f$  is symmetric. So,  $A$  is symmetric.

*Proof (continued).* (b) Fix a bilinear form  $f$  on  $V$ .

*Proof (continued).* (b) Fix a bilinear form  $f$  on  $V$ .

First of all, if  $A = [ a_{i,j} ]_{n \times n}$  is any matrix in  $\mathbb{F}^{n \times n}$  that satisfies the property that  $f(\mathbf{x}, \mathbf{y}) = [ \mathbf{x} ]_{\mathcal{B}}^T A [ \mathbf{y} ]_{\mathcal{B}}$  for all  $\mathbf{x}, \mathbf{y} \in V$ , then (a) guarantees that  $a_{i,j} = f(\mathbf{b}_i, \mathbf{b}_j)$  for all indices  $i, j \in \{1, \dots, n\}$ . This, in particular, proves the uniqueness part of (b).

*Proof (continued).* For existence, we must show that the matrix  $A = [ a_{i,j} ]_{n \times n}$  given by the formula  $a_{i,j} = f(\mathbf{b}_i, \mathbf{b}_j)$  for all indices  $i, j \in \{1, \dots, n\}$ , does indeed satisfy the property that  $f(\mathbf{x}, \mathbf{y}) = [ \mathbf{x} ]_{\mathcal{B}}^T A [ \mathbf{y} ]_{\mathcal{B}}$  for all  $\mathbf{x}, \mathbf{y} \in V$ .

*Proof (continued).* For existence, we must show that the matrix  $A = [a_{i,j}]_{n \times n}$  given by the formula  $a_{i,j} = f(\mathbf{b}_i, \mathbf{b}_j)$  for all indices  $i, j \in \{1, \dots, n\}$ , does indeed satisfy the property that  $f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}}$  for all  $\mathbf{x}, \mathbf{y} \in V$ .

So, fix vectors  $\mathbf{x}, \mathbf{y} \in V$ . Since  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  is a basis of  $V$ , we know that there exist scalars  $c_1, \dots, c_n, d_1, \dots, d_n \in \mathbb{F}$  s.t.  $\mathbf{x} = \sum_{i=1}^n c_i \mathbf{b}_i$  and  $\mathbf{y} = \sum_{j=1}^n d_j \mathbf{b}_j$ , so that  $[\mathbf{x}]_{\mathcal{B}} = [c_1 \ \dots \ c_n]^T$  and  $[\mathbf{y}]_{\mathcal{B}} = [d_1 \ \dots \ d_n]^T$ . We then compute:

$$\begin{aligned} f(\mathbf{x}, \mathbf{y}) &= f\left(\sum_{i=1}^n c_i \mathbf{b}_i, \sum_{j=1}^n d_j \mathbf{b}_j\right) \stackrel{(*)}{=} \sum_{i=1}^n \sum_{j=1}^n c_i d_j \underbrace{f(\mathbf{b}_i, \mathbf{b}_j)}_{=a_{i,j}} \\ &= \sum_{i=1}^n \sum_{j=1}^n a_{i,j} c_i d_j \stackrel{(**)}{=} [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}}, \end{aligned}$$

where (\*) follows from the fact that  $f$  is bilinear, and (\*\*) follows from Proposition 9.1.1(a).  $\square$



## Theorem 9.2.2

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , and let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  be a basis of  $V$ .

- (a) For every matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , the function  $f : V \times V \rightarrow \mathbb{F}$  given by

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V$$

is a bilinear form on  $V$ , and moreover, all the following hold:

- (a.1)  $f(\mathbf{b}_i, \mathbf{b}_j) = a_{i,j}$  for all  $i, j \in \{1, \dots, n\}$ ,  
(a.2)  $f\left(\sum_{i=1}^n c_i \mathbf{b}_i, \sum_{j=1}^n d_j \mathbf{b}_j\right) = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} c_i d_j$  for all  $c_1, \dots, c_n, d_1, \dots, d_n \in \mathbb{F}$ ,  
(a.3)  $f$  is symmetric if and only if  $A$  is symmetric.

- (b) For every bilinear form  $f$  on  $V$ , there exists a unique matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , called the *matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$* , that satisfies the property that

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

Moreover, the entries of the matrix  $A$  are given by

$$a_{i,j} = f(\mathbf{b}_i, \mathbf{b}_j) \quad \text{for all indices } i, j \in \{1, \dots, n\}.$$

- As a special case of Theorem 9.2.2 for the special case of  $V = \mathbb{F}^n$  (where  $\mathbb{F}$  is a field), and  $\mathcal{B} = \mathcal{E}_n$  (the standard basis of  $\mathbb{F}^n$ ), we get the following corollary (next slide).

### Corollary 9.2.3

Let  $\mathbb{F}$  be a field, and let  $\mathcal{E}_n = \{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  be the standard basis of  $\mathbb{F}^n$ .

- (a) For every matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$ , the function  $f : \mathbb{F}^n \times \mathbb{F}^n \rightarrow \mathbb{F}$  given by

$$f(\mathbf{x}, \mathbf{y}) = \mathbf{x}^T \mathbf{A} \mathbf{y} \quad \text{for all } \mathbf{x}, \mathbf{y} \in \mathbb{F}^n$$

is a bilinear form on  $\mathbb{F}^n$ , and moreover, all the following hold:

(a.1)  $f(\mathbf{e}_i, \mathbf{e}_j) = a_{i,j}$  for all  $i, j \in \{1, \dots, n\}$ ,

(a.2)  $f(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i y_j$  for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$

and  $\mathbf{y} = [y_1 \ \dots \ y_n]^T$  in  $\mathbb{F}^n$ ,

(a.3)  $f$  is symmetric iff  $A$  is symmetric.

- (b) For every bilinear form  $f$  on  $\mathbb{F}^n$ , there exists a unique matrix  $A = [a_{i,j}]_{n \times n}$  in  $\mathbb{F}^{n \times n}$  that satisfies the property that

$$f(\mathbf{x}, \mathbf{y}) = \mathbf{x}^T \mathbf{A} \mathbf{y} \quad \text{for all } \mathbf{x}, \mathbf{y} \in \mathbb{F}^n.$$

Moreover, the entries of the matrix  $A$  are given by

$$a_{i,j} = f(\mathbf{e}_i, \mathbf{e}_j) \text{ for all indices } i, j \in \{1, \dots, n\}.$$

- **Remark:** Corollary 9.2.3 implies that, for a field  $\mathbb{F}$ , the bilinear forms on  $\mathbb{F}^n$  are precisely the functions  $f : \mathbb{F}^n \times \mathbb{F}^n \rightarrow \mathbb{F}$  given by

$$f(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i y_j \quad \text{for all } \mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ and } \mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \text{ in } \mathbb{F}^n,$$

where the  $a_{i,j}$ 's are some scalars in  $\mathbb{F}$ .

- **Remark:** Corollary 9.2.3 implies that, for a field  $\mathbb{F}$ , the bilinear forms on  $\mathbb{F}^n$  are precisely the functions  $f : \mathbb{F}^n \times \mathbb{F}^n \rightarrow \mathbb{F}$  given by

$$f(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i y_j \quad \text{for all } \mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ and } \mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \text{ in } \mathbb{F}^n,$$

where the  $a_{i,j}$ 's are some scalars in  $\mathbb{F}$ .

- Moreover, such a bilinear form is symmetric if and only if  $a_{i,j} = a_{j,i}$  for all indices  $i, j \in \{1, \dots, n\}$ .

- **Remark:** Corollary 9.2.3 implies that, for a field  $\mathbb{F}$ , the bilinear forms on  $\mathbb{F}^n$  are precisely the functions  $f : \mathbb{F}^n \times \mathbb{F}^n \rightarrow \mathbb{F}$  given by

$$f(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i y_j \quad \text{for all } \mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ and } \mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \text{ in } \mathbb{F}^n,$$

where the  $a_{i,j}$ 's are some scalars in  $\mathbb{F}$ .

- Moreover, such a bilinear form is symmetric if and only if  $a_{i,j} = a_{j,i}$  for all indices  $i, j \in \{1, \dots, n\}$ .
- The matrix of this bilinear form with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{R}^n$  is  $\begin{bmatrix} a_{i,j} \end{bmatrix}_{n \times n}$  (so, the  $i, j$ -th entry of the matrix is the coefficient in front of  $x_i y_j$  from the formula for  $f$  above).

- For example, functions  $f_1, f_2 : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$  given by the formulas

- $f_1(\mathbf{x}, \mathbf{y}) = x_1y_1 - 3x_1y_2 - 3x_2y_1 + 7x_2y_2,$

- $f_2(\mathbf{x}, \mathbf{y}) = x_1y_1 - 2x_1y_2 + 3x_2y_1 - 3x_2y_2,$

for all  $\mathbf{x} = [x_1 \ x_2]^T$  and  $\mathbf{y} = [y_1 \ y_2]^T$  in  $\mathbb{R}^2$ , are bilinear forms on  $\mathbb{R}^2$ .

- For example, functions  $f_1, f_2 : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$  given by the formulas

- $f_1(\mathbf{x}, \mathbf{y}) = x_1y_1 - 3x_1y_2 - 3x_2y_1 + 7x_2y_2,$

- $f_2(\mathbf{x}, \mathbf{y}) = x_1y_1 - 2x_1y_2 + 3x_2y_1 - 3x_2y_2,$

for all  $\mathbf{x} = [x_1 \ x_2]^T$  and  $\mathbf{y} = [y_1 \ y_2]^T$  in  $\mathbb{R}^2$ , are bilinear forms on  $\mathbb{R}^2$ .

- The bilinear form  $f_1$  is symmetric, whereas the bilinear form  $f_2$  is not.



- For example, functions  $f_1, f_2 : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$  given by the formulas

- $f_1(\mathbf{x}, \mathbf{y}) = x_1y_1 - 3x_1y_2 - 3x_2y_1 + 7x_2y_2,$

- $f_2(\mathbf{x}, \mathbf{y}) = x_1y_1 - 2x_1y_2 + 3x_2y_1 - 3x_2y_2,$

for all  $\mathbf{x} = \begin{bmatrix} x_1 & x_2 \end{bmatrix}^T$  and  $\mathbf{y} = \begin{bmatrix} y_1 & y_2 \end{bmatrix}^T$  in  $\mathbb{R}^2$ , are bilinear forms on  $\mathbb{R}^2$ .

- The bilinear form  $f_1$  is symmetric, whereas the bilinear form  $f_2$  is not.
- The matrices of the bilinear forms  $f_1$  and  $f_2$  with respect to the standard basis  $\mathcal{E}_2$  of  $\mathbb{R}^2$  are

$$A_1 = \begin{bmatrix} 1 & -3 \\ -3 & 7 \end{bmatrix} \quad \text{and} \quad A_2 = \begin{bmatrix} 1 & -2 \\ 3 & -3 \end{bmatrix},$$

respectively.

- For example, functions  $f_1, f_2 : \mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R}$  given by the formulas

- $f_1(\mathbf{x}, \mathbf{y}) = x_1y_1 - 3x_1y_2 - 3x_2y_1 + 7x_2y_2,$

- $f_2(\mathbf{x}, \mathbf{y}) = x_1y_1 - 2x_1y_2 + 3x_2y_1 - 3x_2y_2,$

for all  $\mathbf{x} = \begin{bmatrix} x_1 & x_2 \end{bmatrix}^T$  and  $\mathbf{y} = \begin{bmatrix} y_1 & y_2 \end{bmatrix}^T$  in  $\mathbb{R}^2$ , are bilinear forms on  $\mathbb{R}^2$ .

- The bilinear form  $f_1$  is symmetric, whereas the bilinear form  $f_2$  is not.
- The matrices of the bilinear forms  $f_1$  and  $f_2$  with respect to the standard basis  $\mathcal{E}_2$  of  $\mathbb{R}^2$  are

$$A_1 = \begin{bmatrix} 1 & -3 \\ -3 & 7 \end{bmatrix} \quad \text{and} \quad A_2 = \begin{bmatrix} 1 & -2 \\ 3 & -3 \end{bmatrix},$$

respectively.

- Note that  $A_1$  is symmetric, whereas  $A_2$  is not; this is consistent with the fact that  $f_1$  is symmetric, whereas  $f_2$  is not.

- Reminder:

### Theorem 4.3.2

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and assume that  $U$  is finite-dimensional. Let  $\mathcal{B} = \{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis of  $U$ , and let  $\mathbf{v}_1, \dots, \mathbf{v}_n \in V$ . Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, if the vector space  $U$  is non-trivial (i.e.  $n \neq 0$ ), then this unique linear function  $f : U \rightarrow V$  satisfies the following: for all  $\mathbf{u} \in U$ , we have that

$$f(\mathbf{u}) = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n,$$

where  $[\mathbf{u}]_{\mathcal{B}} = [\alpha_1 \ \dots \ \alpha_n]^T$ . On the other hand, if  $U$  is trivial (i.e.  $U = \{\mathbf{0}\}$ ), then  $f : U \rightarrow V$  is given by  $f(\mathbf{0}) = \mathbf{0}$ .

- Reminder:

### Theorem 4.3.2

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and assume that  $U$  is finite-dimensional. Let  $\mathcal{B} = \{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis of  $U$ , and let  $\mathbf{v}_1, \dots, \mathbf{v}_n \in V$ . Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, if the vector space  $U$  is non-trivial (i.e.  $n \neq 0$ ), then this unique linear function  $f : U \rightarrow V$  satisfies the following: for all  $\mathbf{u} \in U$ , we have that

$$f(\mathbf{u}) = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n,$$

where  $[\mathbf{u}]_{\mathcal{B}} = [\alpha_1 \ \dots \ \alpha_n]^T$ . On the other hand, if  $U$  is trivial (i.e.  $U = \{\mathbf{0}\}$ ), then  $f : U \rightarrow V$  is given by  $f(\mathbf{0}) = \mathbf{0}$ .

- Theorem 4.3.2 essentially states that a linear function can be fully determined by specifying what the vectors of some basis of the domain get mapped to.

- Reminder:

### Theorem 4.3.2

Let  $U$  and  $V$  be vector spaces over a field  $\mathbb{F}$ , and assume that  $U$  is finite-dimensional. Let  $\mathcal{B} = \{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  be a basis of  $U$ , and let  $\mathbf{v}_1, \dots, \mathbf{v}_n \in V$ . Then there exists a unique linear function  $f : U \rightarrow V$  s.t.  $f(\mathbf{u}_1) = \mathbf{v}_1, \dots, f(\mathbf{u}_n) = \mathbf{v}_n$ . Moreover, if the vector space  $U$  is non-trivial (i.e.  $n \neq 0$ ), then this unique linear function  $f : U \rightarrow V$  satisfies the following: for all  $\mathbf{u} \in U$ , we have that

$$f(\mathbf{u}) = \alpha_1 \mathbf{v}_1 + \dots + \alpha_n \mathbf{v}_n,$$

where  $[\mathbf{u}]_{\mathcal{B}} = [\alpha_1 \ \dots \ \alpha_n]^T$ . On the other hand, if  $U$  is trivial (i.e.  $U = \{\mathbf{0}\}$ ), then  $f : U \rightarrow V$  is given by  $f(\mathbf{0}) = \mathbf{0}$ .

- Theorem 4.3.2 essentially states that a linear function can be fully determined by specifying what the vectors of some basis of the domain get mapped to.
- For bilinear forms, Theorem 9.2.2 yields the following analogous result.

### Corollary 9.2.4

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  be a basis of  $V$ , and let  $A = [a_{i,j}]_{n \times n}$  be a matrix in  $\mathbb{F}^{n \times n}$ . Then there exists a unique bilinear form  $f$  on  $V$  that satisfies the property that  $f(\mathbf{b}_i, \mathbf{b}_j) = a_{i,j}$  for all indices  $i, j \in \{1, \dots, n\}$ . Moreover, the matrix of this bilinear form with respect to the basis  $\mathcal{B}$  is precisely the matrix  $A$ .

*Proof.*

### Corollary 9.2.4

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  be a basis of  $V$ , and let  $A = [a_{i,j}]_{n \times n}$  be a matrix in  $\mathbb{F}^{n \times n}$ . Then there exists a unique bilinear form  $f$  on  $V$  that satisfies the property that  $f(\mathbf{b}_i, \mathbf{b}_j) = a_{i,j}$  for all indices  $i, j \in \{1, \dots, n\}$ . Moreover, the matrix of this bilinear form with respect to the basis  $\mathcal{B}$  is precisely the matrix  $A$ .

*Proof. Existence.* By Theorem 9.2.2(a), the function  $f : V \times V \rightarrow \mathbb{F}$  given by

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V$$

is bilinear, and moreover, part (a.1) of Theorem 9.2.2(a) guarantees that  $f(\mathbf{b}_i, \mathbf{b}_j) = a_{i,j}$  for all indices  $i, j \in \{1, \dots, n\}$ . Clearly,  $A$  is the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ .

### Corollary 9.2.4

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , let  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  be a basis of  $V$ , and let  $A = [a_{i,j}]_{n \times n}$  be a matrix in  $\mathbb{F}^{n \times n}$ . Then there exists a unique bilinear form  $f$  on  $V$  that satisfies the property that  $f(\mathbf{b}_i, \mathbf{b}_j) = a_{i,j}$  for all indices  $i, j \in \{1, \dots, n\}$ . Moreover, the matrix of this bilinear form with respect to the basis  $\mathcal{B}$  is precisely the matrix  $A$ .

*Proof (continued).* **Uniqueness.** Suppose that  $f'$  is any bilinear form on  $V$  that satisfies  $f'(\mathbf{b}_i, \mathbf{b}_j) = a_{i,j}$  for all  $i, j \in \{1, \dots, n\}$ . Then Theorem 9.2.2(b) guarantees that the matrix of the bilinear form  $f'$  with respect to the basis  $\mathcal{B}$  is precisely the matrix  $A = [a_{i,j}]_{n \times n}$ , i.e.  $f'(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{y}]_{\mathcal{B}}$  for all  $\mathbf{x}, \mathbf{y} \in V$ .  $\square$



### Theorem 9.2.5 [Change of basis for bilinear forms]

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , let  $f$  be a bilinear form on  $V$ , and let  $\mathcal{B}$  and  $\mathcal{C}$  be bases of  $V$ . Further, let  $B$  be the matrix of  $f$  with respect to  $\mathcal{B}$ , and let  $C$  be the matrix of  $f$  with respect to  $\mathcal{C}$ . Then

$$C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C.$$

*Proof.*

### Theorem 9.2.5 [Change of basis for bilinear forms]

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , let  $f$  be a bilinear form on  $V$ , and let  $\mathcal{B}$  and  $\mathcal{C}$  be bases of  $V$ . Further, let  $B$  be the matrix of  $f$  with respect to  $\mathcal{B}$ , and let  $C$  be the matrix of  $f$  with respect to  $\mathcal{C}$ . Then

$$C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C.$$

*Proof.* For all  $\mathbf{x}, \mathbf{y} \in V$ , we have that

$$\begin{aligned} f(\mathbf{x}, \mathbf{y}) &\stackrel{(*)}{=} [\mathbf{x}]_B^T B [\mathbf{y}]_B \\ &= \left( {}_B[\text{Id}_V]_C [\mathbf{x}]_C \right)^T B \left( {}_B[\text{Id}_V]_C [\mathbf{y}]_C \right) \\ &= [\mathbf{x}]_C^T \left( {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C \right) [\mathbf{y}]_C, \end{aligned}$$

where (\*) follows from the fact that  $B$  is the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ .

### Theorem 9.2.5 [Change of basis for bilinear forms]

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , let  $f$  be a bilinear form on  $V$ , and let  $\mathcal{B}$  and  $\mathcal{C}$  be bases of  $V$ . Further, let  $B$  be the matrix of  $f$  with respect to  $\mathcal{B}$ , and let  $C$  be the matrix of  $f$  with respect to  $\mathcal{C}$ . Then

$$C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C.$$

*Proof (continued).* Reminder:

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_C^T ({}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C) [\mathbf{y}]_C.$$

### Theorem 9.2.5 [Change of basis for bilinear forms]

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$ , let  $f$  be a bilinear form on  $V$ , and let  $\mathcal{B}$  and  $\mathcal{C}$  be bases of  $V$ . Further, let  $B$  be the matrix of  $f$  with respect to  $\mathcal{B}$ , and let  $C$  be the matrix of  $f$  with respect to  $\mathcal{C}$ . Then

$$C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C.$$

*Proof (continued).* Reminder:

$$f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_C^T ({}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C) [\mathbf{y}]_C.$$

But now we have that

$${}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C$$

is the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{C}$ , that is,  $C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C$ .  $\square$

## Definition

Let  $\mathbb{F}$  be a field. A matrix  $A \in \mathbb{F}^{n \times n}$  is said to be *congruent* to a matrix  $B \in \mathbb{F}^{n \times n}$  if there exists an invertible matrix  $P \in \mathbb{F}^{n \times n}$  s.t.  $B = P^T A P$ .

## Definition

Let  $\mathbb{F}$  be a field. A matrix  $A \in \mathbb{F}^{n \times n}$  is said to be *congruent* to a matrix  $B \in \mathbb{F}^{n \times n}$  if there exists an invertible matrix  $P \in \mathbb{F}^{n \times n}$  s.t.  $B = P^T A P$ .

- Like matrix similarity (see Proposition 4.5.13), matrix congruence is an equivalence relation on  $\mathbb{F}^{n \times n}$ .

## Proposition 9.2.6

Let  $\mathbb{F}$  be a field. Then all the following hold:

- Ⓐ for all matrices  $A \in \mathbb{F}^{n \times n}$ ,  $A$  is congruent to  $A$ ;
- Ⓑ for all matrices  $A, B \in \mathbb{F}^{n \times n}$ , if  $A$  is congruent to  $B$ , then  $B$  is congruent to  $A$ ;
- Ⓒ for all matrices  $A, B, C \in \mathbb{F}^{n \times n}$ , if  $A$  is congruent to  $B$  and  $B$  is congruent to  $C$ , then  $A$  is congruent to  $C$ .

- Proof: Lecture Notes (easy).

## Definition

Let  $\mathbb{F}$  be a field. A matrix  $A \in \mathbb{F}^{n \times n}$  is said to be *congruent* to a matrix  $B \in \mathbb{F}^{n \times n}$  if there exists an invertible matrix  $P \in \mathbb{F}^{n \times n}$  s.t.  $B = P^T A P$ .

- Theorem 4.5.16 essentially states that two square matrices are similar iff they represent the same linear function, but possibly with respect to different bases.

## Definition

Let  $\mathbb{F}$  be a field. A matrix  $A \in \mathbb{F}^{n \times n}$  is said to be *congruent* to a matrix  $B \in \mathbb{F}^{n \times n}$  if there exists an invertible matrix  $P \in \mathbb{F}^{n \times n}$  s.t.  $B = P^T A P$ .

- Theorem 4.5.16 essentially states that two square matrices are similar iff they represent the same linear function, but possibly with respect to different bases.
- Theorem 9.2.7 (next slide) is an analog of Theorem 4.5.16 for congruent matrices: it states that two square matrices are congruent if and only if they represent the same bilinear form, but possibly with respect to different bases.



### Theorem 9.2.7

Let  $\mathbb{F}$  be a field, let  $B, C \in \mathbb{F}^{n \times n}$  be matrices, and let  $V$  be an  $n$ -dimensional vector space over the field  $\mathbb{F}$ . Then the following are equivalent:

- (a)  $B$  and  $C$  are congruent;
- (b) for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;
- (c) there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof.*

### Theorem 9.2.7

Let  $\mathbb{F}$  be a field, let  $B, C \in \mathbb{F}^{n \times n}$  be matrices, and let  $V$  be an  $n$ -dimensional vector space over the field  $\mathbb{F}$ . Then the following are equivalent:

- (a)  $B$  and  $C$  are congruent;
- (b) for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;
- (c) there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof.* We will prove the implications  
“(a)  $\implies$  (b)  $\implies$  (c)  $\implies$  (a).”

### Theorem 9.2.7

- Ⓐ  $B$  and  $C$  are congruent;
- Ⓑ for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;

*Proof (continued).* We first assume (a) and prove (b).

### Theorem 9.2.7

- Ⓐ  $B$  and  $C$  are congruent;
- Ⓑ for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;

*Proof (continued).* We first assume (a) and prove (b). By (a), there exists an invertible matrix  $P \in \mathbb{F}^{n \times n}$  s.t.  $C = P^T B P$ .

### Theorem 9.2.7

- (a)  $B$  and  $C$  are congruent;
- (b) for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;

*Proof (continued).* We first assume (a) and prove (b). By (a), there exists an invertible matrix  $P \in \mathbb{F}^{n \times n}$  s.t.  $C = P^T B P$ . Now, to prove (b), we fix a basis  $\mathcal{B}$  of  $V$  and a bilinear form  $f$  on  $V$  such that  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ .

### Theorem 9.2.7

- (a)  $B$  and  $C$  are congruent;
- (b) for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;

*Proof (continued).* We first assume (a) and prove (b). By (a), there exists an invertible matrix  $P \in \mathbb{F}^{n \times n}$  s.t.  $C = P^T B P$ . Now, to prove (b), we fix a basis  $\mathcal{B}$  of  $V$  and a bilinear form  $f$  on  $V$  such that  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ .

Since  $P$  is invertible, Proposition 4.5.12 guarantees that there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $P = {}_{\mathcal{B}} [\text{Id}_V]_{\mathcal{C}}$ .

### Theorem 9.2.7

- (a)  $B$  and  $C$  are congruent;
- (b) for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;

*Proof (continued).* We first assume (a) and prove (b). By (a), there exists an invertible matrix  $P \in \mathbb{F}^{n \times n}$  s.t.  $C = P^T B P$ . Now, to prove (b), we fix a basis  $\mathcal{B}$  of  $V$  and a bilinear form  $f$  on  $V$  such that  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ .

Since  $P$  is invertible, Proposition 4.5.12 guarantees that there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $P = {}_{\mathcal{B}}[\text{Id}_V]_{\mathcal{C}}$ . But then Theorem 9.2.5 guarantees that the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{C}$  is precisely the matrix

$${}_{\mathcal{B}}[\text{Id}_V]_{\mathcal{C}}^T B {}_{\mathcal{B}}[\text{Id}_V]_{\mathcal{C}} = P^T B P = C.$$

This proves (b).

### Theorem 9.2.7

- ⓑ for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;
- ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Next, we assume (b) and prove (c).



### Theorem 9.2.7

- ⓑ for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;
- ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Next, we assume (b) and prove (c). Fix any basis  $\mathcal{B}$  of  $V$ , and define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = \begin{bmatrix} \mathbf{x} \end{bmatrix}_{\mathcal{B}}^T B \begin{bmatrix} \mathbf{y} \end{bmatrix}_{\mathcal{B}} \text{ for all } \mathbf{x}, \mathbf{y} \in V.$$

### Theorem 9.2.7

- ⓑ for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;
- ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Next, we assume (b) and prove (c). Fix any basis  $\mathcal{B}$  of  $V$ , and define  $f : V \times V \rightarrow \mathbb{F}$  by setting  $f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T B [\mathbf{y}]_{\mathcal{B}}$  for all  $\mathbf{x}, \mathbf{y} \in V$ . By Theorem 9.2.2,  $f$  is a bilinear form on  $V$ , and obviously,  $B$  is the matrix of  $f$  with respect to the basis  $\mathcal{B}$ .

### Theorem 9.2.7

- ⓑ for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;
- ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Next, we assume (b) and prove (c). Fix any basis  $\mathcal{B}$  of  $V$ , and define  $f : V \times V \rightarrow \mathbb{F}$  by setting  $f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T B [\mathbf{y}]_{\mathcal{B}}$  for all  $\mathbf{x}, \mathbf{y} \in V$ . By Theorem 9.2.2,  $f$  is a bilinear form on  $V$ , and obviously,  $B$  is the matrix of  $f$  with respect to the basis  $\mathcal{B}$ .

Using (b), we now fix a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of the bilinear form  $f$  with respect to  $\mathcal{C}$ .

### Theorem 9.2.7

- ⓑ for all bases  $\mathcal{B}$  of  $V$  and bilinear forms  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ ;
- ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Next, we assume (b) and prove (c). Fix any basis  $\mathcal{B}$  of  $V$ , and define  $f : V \times V \rightarrow \mathbb{F}$  by setting  $f(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T B [\mathbf{y}]_{\mathcal{B}}$  for all  $\mathbf{x}, \mathbf{y} \in V$ . By Theorem 9.2.2,  $f$  is a bilinear form on  $V$ , and obviously,  $B$  is the matrix of  $f$  with respect to the basis  $\mathcal{B}$ .

Using (b), we now fix a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of the bilinear form  $f$  with respect to  $\mathcal{C}$ . We have now constructed bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$ , and a bilinear form  $f$  on  $V$ , s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ . This proves (c).

### Theorem 9.2.7

- Ⓐ  $B$  and  $C$  are congruent;
- Ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Finally, we assume (c) and prove (a).

### Theorem 9.2.7

- Ⓐ  $B$  and  $C$  are congruent;
- Ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Finally, we assume (c) and prove (a).

Using (c), we fix bases  $\mathcal{B}$  and  $\mathcal{C}$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

### Theorem 9.2.7

- Ⓐ  $B$  and  $C$  are congruent;
- Ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Finally, we assume (c) and prove (a).

Using (c), we fix bases  $\mathcal{B}$  and  $\mathcal{C}$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

Set  $P := {}_{\mathcal{B}}[\text{Id}_V]_{\mathcal{C}}$ .

### Theorem 9.2.7

- Ⓐ  $B$  and  $C$  are congruent;
- Ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

*Proof (continued).* Finally, we assume (c) and prove (a).

Using (c), we fix bases  $\mathcal{B}$  and  $\mathcal{C}$  and a bilinear form  $f$  on  $V$  s.t.  $B$  is the matrix of  $f$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $f$  with respect to  $\mathcal{C}$ .

Set  $P := {}_{\mathcal{B}}[\text{Id}_V]_{\mathcal{C}}$ . By Proposition 4.5.12,  $P$  is invertible, and by Theorem 9.2.5, we have that  $C = P^T B P$ . This proves (a).  $\square$



## Definition

The *characteristic* of a field  $\mathbb{F}$  is the smallest positive integer  $n$  (if it exists) s.t. in the field  $\mathbb{F}$ , we have that

$$\underbrace{1 + \cdots + 1}_n = 0,$$

where the 1's and the 0 are understood to be in the field  $\mathbb{F}$ . If no such  $n$  exists, then  $\text{char}(\mathbb{F}) := 0$ .

- Fields  $\mathbb{Q}$ ,  $\mathbb{R}$ , and  $\mathbb{C}$  all have characteristic 0.
- On the other hand, for all prime numbers  $p$ , we have that  $\text{char}(\mathbb{Z}_p) = p$ .
- By Theorem 2.4.5, the characteristic of any field is either 0 or a prime number.

### Proposition 9.2.8

Let  $f$  and  $g$  be **symmetric** bilinear forms on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2, and assume that for all  $\mathbf{x} \in V$ , we have that  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ . Then  $f = g$ .

- Proof: next slide.
- **Remark:** Proposition 9.2.8 applies to bilinear forms over vector spaces of characteristic other than 2.
  - In such fields, we can divide by  $2 := 1 + 1$ , since  $2 = 1 + 1 \neq 0$ .
  - The only field of characteristic 2 that we have seen is  $\mathbb{Z}_2$ , but other fields of characteristic 2 do exist.

### Proposition 9.2.8

Let  $f$  and  $g$  be **symmetric** bilinear forms on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2, and assume that for all  $\mathbf{x} \in V$ , we have that  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ . Then  $f = g$ .

*Proof.*

### Proposition 9.2.8

Let  $f$  and  $g$  be **symmetric** bilinear forms on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2, and assume that for all  $\mathbf{x} \in V$ , we have that  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ . Then  $f = g$ .

*Proof.* Fix  $\mathbf{x}, \mathbf{y} \in V$ . We must show that  $f(\mathbf{x}, \mathbf{y}) = g(\mathbf{x}, \mathbf{y})$ .

### Proposition 9.2.8

Let  $f$  and  $g$  be **symmetric** bilinear forms on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2, and assume that for all  $\mathbf{x} \in V$ , we have that  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ . Then  $f = g$ .

*Proof.* Fix  $\mathbf{x}, \mathbf{y} \in V$ . We must show that  $f(\mathbf{x}, \mathbf{y}) = g(\mathbf{x}, \mathbf{y})$ . By hypothesis, all the following hold:

- ①  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ ;
- ②  $f(\mathbf{y}, \mathbf{y}) = g(\mathbf{y}, \mathbf{y})$ ;
- ③  $f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y})$ .

### Proposition 9.2.8

Let  $f$  and  $g$  be **symmetric** bilinear forms on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2, and assume that for all  $\mathbf{x} \in V$ , we have that  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ . Then  $f = g$ .

*Proof.* Fix  $\mathbf{x}, \mathbf{y} \in V$ . We must show that  $f(\mathbf{x}, \mathbf{y}) = g(\mathbf{x}, \mathbf{y})$ . By hypothesis, all the following hold:

- ①  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ ;
- ②  $f(\mathbf{y}, \mathbf{y}) = g(\mathbf{y}, \mathbf{y})$ ;
- ③  $f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y})$ .

On the other hand, since  $f$  and  $g$  are bilinear, we have that

- (4)  $f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = f(\mathbf{x}, \mathbf{x}) + f(\mathbf{x}, \mathbf{y}) + f(\mathbf{y}, \mathbf{x}) + f(\mathbf{y}, \mathbf{y})$ ;
- (5)  $g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x}, \mathbf{x}) + g(\mathbf{x}, \mathbf{y}) + g(\mathbf{y}, \mathbf{x}) + g(\mathbf{y}, \mathbf{y})$ .

### Proposition 9.2.8

Let  $f$  and  $g$  be **symmetric** bilinear forms on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2, and assume that for all  $\mathbf{x} \in V$ , we have that  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ . Then  $f = g$ .

*Proof.* Fix  $\mathbf{x}, \mathbf{y} \in V$ . We must show that  $f(\mathbf{x}, \mathbf{y}) = g(\mathbf{x}, \mathbf{y})$ . By hypothesis, all the following hold:

- ①  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ ;
- ②  $f(\mathbf{y}, \mathbf{y}) = g(\mathbf{y}, \mathbf{y})$ ;
- ③  $f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y})$ .

On the other hand, since  $f$  and  $g$  are bilinear, we have that

$$(4) \quad f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = f(\mathbf{x}, \mathbf{x}) + f(\mathbf{x}, \mathbf{y}) + f(\mathbf{y}, \mathbf{x}) + f(\mathbf{y}, \mathbf{y});$$

$$(5) \quad g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x}, \mathbf{x}) + g(\mathbf{x}, \mathbf{y}) + g(\mathbf{y}, \mathbf{x}) + g(\mathbf{y}, \mathbf{y}).$$

By (1)-(5), it follows that  $f(\mathbf{x}, \mathbf{y}) + f(\mathbf{y}, \mathbf{x}) = g(\mathbf{x}, \mathbf{y}) + g(\mathbf{y}, \mathbf{x})$ .

### Proposition 9.2.8

Let  $f$  and  $g$  be **symmetric** bilinear forms on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2, and assume that for all  $\mathbf{x} \in V$ , we have that  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ . Then  $f = g$ .

*Proof.* Fix  $\mathbf{x}, \mathbf{y} \in V$ . We must show that  $f(\mathbf{x}, \mathbf{y}) = g(\mathbf{x}, \mathbf{y})$ . By hypothesis, all the following hold:

- ①  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ ;
- ②  $f(\mathbf{y}, \mathbf{y}) = g(\mathbf{y}, \mathbf{y})$ ;
- ③  $f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y})$ .

On the other hand, since  $f$  and  $g$  are bilinear, we have that

$$(4) \quad f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = f(\mathbf{x}, \mathbf{x}) + f(\mathbf{x}, \mathbf{y}) + f(\mathbf{y}, \mathbf{x}) + f(\mathbf{y}, \mathbf{y});$$

$$(5) \quad g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x}, \mathbf{x}) + g(\mathbf{x}, \mathbf{y}) + g(\mathbf{y}, \mathbf{x}) + g(\mathbf{y}, \mathbf{y}).$$

By (1)-(5), it follows that  $f(\mathbf{x}, \mathbf{y}) + f(\mathbf{y}, \mathbf{x}) = g(\mathbf{x}, \mathbf{y}) + g(\mathbf{y}, \mathbf{x})$ .

But since  $f$  and  $g$  are symmetric, we further have that

$f(\mathbf{x}, \mathbf{y}) = f(\mathbf{y}, \mathbf{x})$  and  $g(\mathbf{x}, \mathbf{y}) = g(\mathbf{y}, \mathbf{x})$ , and it follows that

$$2f(\mathbf{x}, \mathbf{y}) = 2g(\mathbf{x}, \mathbf{y}).$$



### Proposition 9.2.8

Let  $f$  and  $g$  be **symmetric** bilinear forms on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2, and assume that for all  $\mathbf{x} \in V$ , we have that  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ . Then  $f = g$ .

*Proof.* Fix  $\mathbf{x}, \mathbf{y} \in V$ . We must show that  $f(\mathbf{x}, \mathbf{y}) = g(\mathbf{x}, \mathbf{y})$ . By hypothesis, all the following hold:

- ①  $f(\mathbf{x}, \mathbf{x}) = g(\mathbf{x}, \mathbf{x})$ ;
- ②  $f(\mathbf{y}, \mathbf{y}) = g(\mathbf{y}, \mathbf{y})$ ;
- ③  $f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y})$ .

On the other hand, since  $f$  and  $g$  are bilinear, we have that

$$(4) \quad f(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = f(\mathbf{x}, \mathbf{x}) + f(\mathbf{x}, \mathbf{y}) + f(\mathbf{y}, \mathbf{x}) + f(\mathbf{y}, \mathbf{y});$$

$$(5) \quad g(\mathbf{x} + \mathbf{y}, \mathbf{x} + \mathbf{y}) = g(\mathbf{x}, \mathbf{x}) + g(\mathbf{x}, \mathbf{y}) + g(\mathbf{y}, \mathbf{x}) + g(\mathbf{y}, \mathbf{y}).$$

By (1)-(5), it follows that  $f(\mathbf{x}, \mathbf{y}) + f(\mathbf{y}, \mathbf{x}) = g(\mathbf{x}, \mathbf{y}) + g(\mathbf{y}, \mathbf{x})$ .

But since  $f$  and  $g$  are symmetric, we further have that

$f(\mathbf{x}, \mathbf{y}) = f(\mathbf{y}, \mathbf{x})$  and  $g(\mathbf{x}, \mathbf{y}) = g(\mathbf{y}, \mathbf{x})$ , and it follows that

$2f(\mathbf{x}, \mathbf{y}) = 2g(\mathbf{x}, \mathbf{y})$ . Since  $\text{char}(\mathbb{F}) \neq 2$  (and consequently,

$2 = 1 + 1 \neq 0$  in our field  $\mathbb{F}$ ), we deduce that  $f(\mathbf{x}, \mathbf{y}) = g(\mathbf{x}, \mathbf{y})$ .  $\square$

### ③ Quadratic forms

### 3 Quadratic forms

#### Definition

A *quadratic form* on a vector space  $V$  over a field  $\mathbb{F}$  is any function  $q : V \rightarrow \mathbb{F}$  for which there exists a bilinear form  $f : V \times V \rightarrow \mathbb{F}$  s.t.  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$  for all  $\mathbf{x} \in V$ .

### 3 Quadratic forms

#### Definition

A *quadratic form* on a vector space  $V$  over a field  $\mathbb{F}$  is any function  $q : V \rightarrow \mathbb{F}$  for which there exists a bilinear form  $f : V \times V \rightarrow \mathbb{F}$  s.t.  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$  for all  $\mathbf{x} \in V$ .

- Quadratic forms are defined for vector spaces over fields of any characteristic.

### 3 Quadratic forms

#### Definition

A *quadratic form* on a vector space  $V$  over a field  $\mathbb{F}$  is any function  $q : V \rightarrow \mathbb{F}$  for which there exists a bilinear form  $f : V \times V \rightarrow \mathbb{F}$  s.t.  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$  for all  $\mathbf{x} \in V$ .

- Quadratic forms are defined for vector spaces over fields of any characteristic.
- However, in all our results that follow, we assume that the field in question is of characteristic other than 2, so that we can divide by 2.

### Theorem 9.3.1

Let  $q$  be a quadratic form on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2. Then there exists a unique **symmetric** bilinear form  $f$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ . Furthermore, if the vector space  $V$  is non-trivial and finite-dimensional, then for any basis  $\mathcal{B}$  of  $V$ , there exists a unique **symmetric** matrix  $A \in \mathbb{F}^{n \times n}$  s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V,$$

and moreover, this unique symmetric matrix  $A$  is precisely the matrix of the symmetric bilinear form  $f$  with respect to the basis  $\mathcal{B}$ .

### Theorem 9.3.1

Let  $q$  be a quadratic form on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2. Then there exists a unique **symmetric** bilinear form  $f$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ . Furthermore, if the vector space  $V$  is non-trivial and finite-dimensional, then for any basis  $\mathcal{B}$  of  $V$ , there exists a unique **symmetric** matrix  $A \in \mathbb{F}^{n \times n}$  s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V,$$

and moreover, this unique symmetric matrix  $A$  is precisely the matrix of the symmetric bilinear form  $f$  with respect to the basis  $\mathcal{B}$ .

- **Terminology:** The symmetric matrix  $A$  from the statement of Theorem 9.3.1 is called the *matrix of the quadratic form  $q$  with respect to the basis  $\mathcal{B}$* .
  - For emphasis, we may optionally refer to  $A$  as the **symmetric matrix of the quadratic form  $q$  with respect to the basis  $\mathcal{B}$** .

### Theorem 9.3.1

Let  $q$  be a quadratic form on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2. Then there exists a unique **symmetric** bilinear form  $f$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ . Furthermore, if the vector space  $V$  is non-trivial and finite-dimensional, then for any basis  $\mathcal{B}$  of  $V$ , there exists a unique **symmetric** matrix  $A \in \mathbb{F}^{n \times n}$  s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V,$$

and moreover, this unique symmetric matrix  $A$  is precisely the matrix of the symmetric bilinear form  $f$  with respect to the basis  $\mathcal{B}$ .

- **Warning:** There may possibly exist more than one matrix  $A \in \mathbb{F}^{n \times n}$  that satisfies the property that  $q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}}$  for all  $\mathbf{x} \in V$ .
  - However, only one such matrix is symmetric.
  - This (unique) symmetric matrix is the one that we refer to as the matrix of  $q$  with respect to  $\mathcal{B}$ .



### Theorem 9.3.1

Let  $q$  be a quadratic form on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2. Then there exists a unique **symmetric** bilinear form  $f$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ . Furthermore, if the vector space  $V$  is non-trivial and finite-dimensional, then for any basis  $\mathcal{B}$  of  $V$ , there exists a unique **symmetric** matrix  $A \in \mathbb{F}^{n \times n}$  s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V,$$

and moreover, this unique symmetric matrix  $A$  is precisely the matrix of the symmetric bilinear form  $f$  with respect to the basis  $\mathcal{B}$ .

- **Warning:** There may possibly exist more than one matrix  $A \in \mathbb{F}^{n \times n}$  that satisfies the property that  $q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}}$  for all  $\mathbf{x} \in V$ .
  - However, only one such matrix is symmetric.
  - This (unique) symmetric matrix is the one that we refer to as the matrix of  $q$  with respect to  $\mathcal{B}$ .
- Now let's prove the theorem!

*Proof.* We first prove the existence and uniqueness of the symmetric bilinear form  $f$ .

*Proof.* We first prove the existence and uniqueness of the symmetric bilinear form  $f$ . By the definition of a quadratic form, there exists some bilinear form  $h$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x})$ .

*Proof.* We first prove the existence and uniqueness of the symmetric bilinear form  $f$ . By the definition of a quadratic form, there exists some bilinear form  $h$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x})$ . Now, using the fact that  $\text{char}(\mathbb{F}) \neq 2$ , we define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = \frac{1}{2} \left( h(\mathbf{x}, \mathbf{y}) + h(\mathbf{y}, \mathbf{x}) \right) \quad \text{for all } \mathbf{x} \in V.$$

*Proof.* We first prove the existence and uniqueness of the symmetric bilinear form  $f$ . By the definition of a quadratic form, there exists some bilinear form  $h$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x})$ . Now, using the fact that  $\text{char}(\mathbb{F}) \neq 2$ , we define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = \frac{1}{2} \left( h(\mathbf{x}, \mathbf{y}) + h(\mathbf{y}, \mathbf{x}) \right) \quad \text{for all } \mathbf{x} \in V.$$

It is then straightforward to check that  $f$  is a symmetric bilinear form on  $V$  (details?).

*Proof.* We first prove the existence and uniqueness of the symmetric bilinear form  $f$ . By the definition of a quadratic form, there exists some bilinear form  $h$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x})$ . Now, using the fact that  $\text{char}(\mathbb{F}) \neq 2$ , we define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = \frac{1}{2} \left( h(\mathbf{x}, \mathbf{y}) + h(\mathbf{y}, \mathbf{x}) \right) \quad \text{for all } \mathbf{x} \in V.$$

It is then straightforward to check that  $f$  is a symmetric bilinear form on  $V$  (details?). Moreover, for all  $\mathbf{x} \in V$ , we have that

$$q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x}) = \frac{1}{2} \left( h(\mathbf{x}, \mathbf{x}) + h(\mathbf{x}, \mathbf{x}) \right) = f(\mathbf{x}, \mathbf{x}),$$

which is what we needed.

*Proof.* We first prove the existence and uniqueness of the symmetric bilinear form  $f$ . By the definition of a quadratic form, there exists some bilinear form  $h$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x})$ . Now, using the fact that  $\text{char}(\mathbb{F}) \neq 2$ , we define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = \frac{1}{2} \left( h(\mathbf{x}, \mathbf{y}) + h(\mathbf{y}, \mathbf{x}) \right) \quad \text{for all } \mathbf{x} \in V.$$

It is then straightforward to check that  $f$  is a symmetric bilinear form on  $V$  (details?). Moreover, for all  $\mathbf{x} \in V$ , we have that

$$q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x}) = \frac{1}{2} \left( h(\mathbf{x}, \mathbf{x}) + h(\mathbf{x}, \mathbf{x}) \right) = f(\mathbf{x}, \mathbf{x}),$$

which is what we needed. This completes the proof of existence.

*Proof.* We first prove the existence and uniqueness of the symmetric bilinear form  $f$ . By the definition of a quadratic form, there exists some bilinear form  $h$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x})$ . Now, using the fact that  $\text{char}(\mathbb{F}) \neq 2$ , we define  $f : V \times V \rightarrow \mathbb{F}$  by setting

$$f(\mathbf{x}, \mathbf{y}) = \frac{1}{2} \left( h(\mathbf{x}, \mathbf{y}) + h(\mathbf{y}, \mathbf{x}) \right) \quad \text{for all } \mathbf{x} \in V.$$

It is then straightforward to check that  $f$  is a symmetric bilinear form on  $V$  (details?). Moreover, for all  $\mathbf{x} \in V$ , we have that

$$q(\mathbf{x}) = h(\mathbf{x}, \mathbf{x}) = \frac{1}{2} \left( h(\mathbf{x}, \mathbf{x}) + h(\mathbf{x}, \mathbf{x}) \right) = f(\mathbf{x}, \mathbf{x}),$$

which is what we needed. This completes the proof of existence. Uniqueness follows immediately from Proposition 9.2.8.

- Indeed, suppose that  $f_1$  and  $f_2$  are symmetric bilinear forms on  $V$  s.t.  $q(\mathbf{x}) = f_1(\mathbf{x}, \mathbf{x})$  and  $q(\mathbf{x}) = f_2(\mathbf{x}, \mathbf{x})$  for all  $\mathbf{x} \in V$ . Then  $f_1(\mathbf{x}, \mathbf{x}) = f_2(\mathbf{x}, \mathbf{x})$  for all  $\mathbf{x} \in V$ . But then by Proposition 9.2.8, we have that  $f_1 = f_2$ .



*Proof (continued).* Let us now assume that the vector space  $V$  is non-trivial and finite-dimensional, and let  $\mathcal{B}$  be a basis of  $V$ .

*Proof (continued).* Let us now assume that the vector space  $V$  is non-trivial and finite-dimensional, and let  $\mathcal{B}$  be a basis of  $V$ . Let  $A \in \mathbb{F}^{n \times n}$  be the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ ; by Theorem 9.2.2, the matrix  $A$  is symmetric.

*Proof (continued).* Let us now assume that the vector space  $V$  is non-trivial and finite-dimensional, and let  $\mathcal{B}$  be a basis of  $V$ . Let  $A \in \mathbb{F}^{n \times n}$  be the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ ; by Theorem 9.2.2, the matrix  $A$  is symmetric. Obviously,  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}}$  for all  $\mathbf{x} \in V$ .

*Proof (continued).* Let us now assume that the vector space  $V$  is non-trivial and finite-dimensional, and let  $\mathcal{B}$  be a basis of  $V$ . Let  $A \in \mathbb{F}^{n \times n}$  be the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ ; by Theorem 9.2.2, the matrix  $A$  is symmetric. Obviously,  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}}$  for all  $\mathbf{x} \in V$ . For uniqueness, suppose that  $A' \in \mathbb{F}^{n \times n}$  is a symmetric matrix s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V.$$

WTS  $A' = A$ .

*Proof (continued).* Let us now assume that the vector space  $V$  is non-trivial and finite-dimensional, and let  $\mathcal{B}$  be a basis of  $V$ . Let  $A \in \mathbb{F}^{n \times n}$  be the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ ; by Theorem 9.2.2, the matrix  $A$  is symmetric. Obviously,  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}}$  for all  $\mathbf{x} \in V$ . For uniqueness, suppose that  $A' \in \mathbb{F}^{n \times n}$  is a symmetric matrix s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V.$$

WTS  $A' = A$ . Define  $f' : V \times V \rightarrow \mathbb{F}$  by setting

$$f'(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

*Proof (continued).* Let us now assume that the vector space  $V$  is non-trivial and finite-dimensional, and let  $\mathcal{B}$  be a basis of  $V$ . Let  $A \in \mathbb{F}^{n \times n}$  be the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ ; by Theorem 9.2.2, the matrix  $A$  is symmetric. Obviously,  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}}$  for all  $\mathbf{x} \in V$ . For uniqueness, suppose that  $A' \in \mathbb{F}^{n \times n}$  is a symmetric matrix s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V.$$

WTS  $A' = A$ . Define  $f' : V \times V \rightarrow \mathbb{F}$  by setting

$$f'(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

By Theorem 9.2.2(a),  $f'$  is a symmetric bilinear form.

*Proof (continued).* Let us now assume that the vector space  $V$  is non-trivial and finite-dimensional, and let  $\mathcal{B}$  be a basis of  $V$ . Let  $A \in \mathbb{F}^{n \times n}$  be the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ ; by Theorem 9.2.2, the matrix  $A$  is symmetric. Obviously,  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}}$  for all  $\mathbf{x} \in V$ . For uniqueness, suppose that  $A' \in \mathbb{F}^{n \times n}$  is a symmetric matrix s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V.$$

WTS  $A' = A$ . Define  $f' : V \times V \rightarrow \mathbb{F}$  by setting

$$f'(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

By Theorem 9.2.2(a),  $f'$  is a symmetric bilinear form. But then for all  $\mathbf{x} \in V$ , we have that  $f'(\mathbf{x}, \mathbf{x}) = q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ , and so by Proposition 9.2.8,  $f' = f$ .

*Proof (continued).* Let us now assume that the vector space  $V$  is non-trivial and finite-dimensional, and let  $\mathcal{B}$  be a basis of  $V$ . Let  $A \in \mathbb{F}^{n \times n}$  be the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$ ; by Theorem 9.2.2, the matrix  $A$  is symmetric. Obviously,  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}}$  for all  $\mathbf{x} \in V$ . For uniqueness, suppose that  $A' \in \mathbb{F}^{n \times n}$  is a symmetric matrix s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V.$$

WTS  $A' = A$ . Define  $f' : V \times V \rightarrow \mathbb{F}$  by setting

$$f'(\mathbf{x}, \mathbf{y}) = [\mathbf{x}]_{\mathcal{B}}^T A' [\mathbf{y}]_{\mathcal{B}} \quad \text{for all } \mathbf{x}, \mathbf{y} \in V.$$

By Theorem 9.2.2(a),  $f'$  is a symmetric bilinear form. But then for all  $\mathbf{x} \in V$ , we have that  $f'(\mathbf{x}, \mathbf{x}) = q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ , and so by Proposition 9.2.8,  $f' = f$ . The uniqueness part of Theorem 9.2.2(b) now guarantees that  $A' = A$ , and we are done.  $\square$



### Theorem 9.3.1

Let  $q$  be a quadratic form on a vector space  $V$  over a field  $\mathbb{F}$  of characteristic other than 2. Then there exists a unique **symmetric** bilinear form  $f$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ . Furthermore, if the vector space  $V$  is non-trivial and finite-dimensional, then for any basis  $\mathcal{B}$  of  $V$ , there exists a unique **symmetric** matrix  $A \in \mathbb{F}^{n \times n}$  s.t.

$$q(\mathbf{x}) = [\mathbf{x}]_{\mathcal{B}}^T A [\mathbf{x}]_{\mathcal{B}} \quad \text{for all } \mathbf{x} \in V,$$

and moreover, this unique symmetric matrix  $A$  is precisely the matrix of the symmetric bilinear form  $f$  with respect to the basis  $\mathcal{B}$ .

- **Remark:** Let  $\mathbb{F}$  be a field. Then quadratic forms  $q$  on  $\mathbb{F}^n$  are all of the form

$$q(\mathbf{x}) = \sum_{i=1}^n \sum_{j=1}^n b_{i,j} x_i x_j \quad \text{for all } \mathbf{x} = [x_1 \ \dots \ x_n]^T \text{ in } \mathbb{F}^n,$$

where the  $b_{i,j}$ 's are some elements of  $\mathbb{F}$ .

- **Remark:** Let  $\mathbb{F}$  be a field. Then quadratic forms  $q$  on  $\mathbb{F}^n$  are all of the form

$$q(\mathbf{x}) = \sum_{i=1}^n \sum_{j=1}^n b_{i,j} x_i x_j \quad \text{for all } \mathbf{x} = [x_1 \ \dots \ x_n]^T \text{ in } \mathbb{F}^n,$$

where the  $b_{i,j}$ 's are some elements of  $\mathbb{F}$ .

- If  $\text{char}(\mathbb{F}) \neq 2$ , then the matrix of such a quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{F}^n$  is the matrix  $A = [a_{i,j}]_{n \times n}$  whose entries are given by  $a_{i,j} = \frac{1}{2}(b_{i,j} + b_{j,i})$  for all  $i, j \in \{1, \dots, n\}$ .

- **Remark:** Let  $\mathbb{F}$  be a field. Then quadratic forms  $q$  on  $\mathbb{F}^n$  are all of the form

$$q(\mathbf{x}) = \sum_{i=1}^n \sum_{j=1}^n b_{i,j} x_i x_j \quad \text{for all } \mathbf{x} = [x_1 \ \dots \ x_n]^T \text{ in } \mathbb{F}^n,$$

where the  $b_{i,j}$ 's are some elements of  $\mathbb{F}$ .

- If  $\text{char}(\mathbb{F}) \neq 2$ , then the matrix of such a quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{F}^n$  is the matrix  $A = [a_{i,j}]_{n \times n}$  whose entries are given by  $a_{i,j} = \frac{1}{2}(b_{i,j} + b_{j,i})$  for all  $i, j \in \{1, \dots, n\}$ .
- Indeed, by construction,  $A$  is symmetric, and we see that for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  in  $\mathbb{F}^n$ , we have the following:

$$\begin{aligned} \mathbf{x}^T A \mathbf{x} &\stackrel{(*)}{=} \sum_{i=1}^n \sum_{j=1}^n a_{i,j} x_i x_j = \sum_{i=1}^n \sum_{j=1}^n \frac{1}{2}(b_{i,j} + b_{j,i}) x_i x_j \\ &= \frac{1}{2} \left( \left( \sum_{i=1}^n \sum_{j=1}^n b_{i,j} x_i x_j \right) + \left( \sum_{i=1}^n \sum_{j=1}^n b_{j,i} x_i x_j \right) \right) \\ &= \sum_{i=1}^n \sum_{j=1}^n b_{i,j} x_i x_j = q(\mathbf{x}), \end{aligned}$$

where (\*) follows from Proposition 9.1.1(a).

### Example 9.3.2

Consider the quadratic form  $q$  on  $\mathbb{R}^3$  given by

$$q(\mathbf{x}) = 3x_1^2 + 2x_1x_2 - 4x_1x_3 + 5x_2^2 - 6x_2x_3 + 2x_3^2$$

for all vectors  $\mathbf{x} = [x_1 \ x_2 \ x_3]^T$  in  $\mathbb{R}^3$ . Then the matrix of  $q$  with respect to the standard basis  $\mathcal{E}_3$  of  $\mathbb{R}^3$  is the matrix

$$A := \begin{bmatrix} 3 & 1 & -2 \\ 1 & 5 & -3 \\ -2 & -3 & 2 \end{bmatrix}.$$

### Corollary 9.3.3 [Change of basis for quadratic forms]

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$  of characteristic other than 2, let  $q$  be a quadratic form on  $V$ , and let  $\mathcal{B}$  and  $\mathcal{C}$  be bases of  $V$ . Further, let  $B$  be the (symmetric) matrix of  $q$  with respect to  $\mathcal{B}$ , and let  $C$  be the (symmetric) matrix of  $q$  with respect to  $\mathcal{C}$ . Then

$$C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C.$$

*Proof.*

### Corollary 9.3.3 [Change of basis for quadratic forms]

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$  of characteristic other than 2, let  $q$  be a quadratic form on  $V$ , and let  $\mathcal{B}$  and  $\mathcal{C}$  be bases of  $V$ . Further, let  $B$  be the (symmetric) matrix of  $q$  with respect to  $\mathcal{B}$ , and let  $C$  be the (symmetric) matrix of  $q$  with respect to  $\mathcal{C}$ . Then

$$C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C.$$

*Proof.* By Theorem 9.3.1, there exists a unique symmetric bilinear form  $f$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ .

### Corollary 9.3.3 [Change of basis for quadratic forms]

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$  of characteristic other than 2, let  $q$  be a quadratic form on  $V$ , and let  $\mathcal{B}$  and  $\mathcal{C}$  be bases of  $V$ . Further, let  $B$  be the (symmetric) matrix of  $q$  with respect to  $\mathcal{B}$ , and let  $C$  be the (symmetric) matrix of  $q$  with respect to  $\mathcal{C}$ . Then

$$C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C.$$

*Proof.* By Theorem 9.3.1, there exists a unique symmetric bilinear form  $f$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ . Theorem 9.3.1 further guarantees that  $B$  (resp.  $C$ ) is the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$  (resp.  $\mathcal{C}$ ) of  $\mathbb{F}^n$ .



### Corollary 9.3.3 [Change of basis for quadratic forms]

Let  $V$  be a non-trivial, finite-dimensional vector space over a field  $\mathbb{F}$  of characteristic other than 2, let  $q$  be a quadratic form on  $V$ , and let  $\mathcal{B}$  and  $\mathcal{C}$  be bases of  $V$ . Further, let  $B$  be the (symmetric) matrix of  $q$  with respect to  $\mathcal{B}$ , and let  $C$  be the (symmetric) matrix of  $q$  with respect to  $\mathcal{C}$ . Then

$$C = {}_B[\text{Id}_V]_C^T B {}_B[\text{Id}_V]_C.$$

*Proof.* By Theorem 9.3.1, there exists a unique symmetric bilinear form  $f$  on  $V$  s.t. for all  $\mathbf{x} \in V$ , we have that  $q(\mathbf{x}) = f(\mathbf{x}, \mathbf{x})$ .

Theorem 9.3.1 further guarantees that  $B$  (resp.  $C$ ) is the matrix of the bilinear form  $f$  with respect to the basis  $\mathcal{B}$  (resp.  $\mathcal{C}$ ) of  $\mathbb{F}^n$ . The result now follows immediately from Theorem 9.5.2.  $\square$

### Theorem 9.3.4

Let  $\mathbb{F}$  be a field of characteristic other than 2, let  $B, C \in \mathbb{F}^{n \times n}$  be **symmetric** matrices, and let  $V$  be an  $n$ -dimensional vector space over the field  $\mathbb{F}$ . Then the following are equivalent:

- Ⓐ  $B$  and  $C$  are congruent;
- Ⓑ for all bases  $\mathcal{B}$  of  $V$  and quadratic forms  $q$  on  $V$  s.t.  $B$  is the matrix of  $q$  with respect to  $\mathcal{B}$ , there exists a basis  $\mathcal{C}$  of  $V$  s.t.  $C$  is the matrix of  $q$  with respect to  $\mathcal{C}$ ;
- Ⓒ there exist bases  $\mathcal{B}$  and  $\mathcal{C}$  of  $V$  and a quadratic form  $q$  on  $V$  s.t.  $B$  is the matrix of  $q$  with respect to  $\mathcal{B}$ , and  $C$  is the matrix of  $q$  with respect to  $\mathcal{C}$ .

- Proof: Lecture Notes

④ Quadratic forms on  $\mathbb{R}^n$

④ Quadratic forms on  $\mathbb{R}^n$

- In what follows, orthogonality and orthonormality in  $\mathbb{R}^n$  are assumed to be with respect to the standard scalar product  $\cdot$  and the induced norm  $\|\cdot\|$ .

#### ④ Quadratic forms on $\mathbb{R}^n$

- In what follows, orthogonality and orthonormality in  $\mathbb{R}^n$  are assumed to be with respect to the standard scalar product  $\cdot$  and the induced norm  $\|\cdot\|$ .
- By Corollary 8.7.4, any symmetric matrix in  $\mathbb{R}^{n \times n}$  has  $n$  real eigenvalues (when algebraic multiplicities are taken into account).

#### ④ Quadratic forms on $\mathbb{R}^n$

- In what follows, orthogonality and orthonormality in  $\mathbb{R}^n$  are assumed to be with respect to the standard scalar product  $\cdot$  and the induced norm  $\|\cdot\|$ .
- By Corollary 8.7.4, any symmetric matrix in  $\mathbb{R}^{n \times n}$  has  $n$  real eigenvalues (when algebraic multiplicities are taken into account).
  - With this in mind, we define the following (next slide).

## Definition

The *signature* of a symmetric matrix  $A \in \mathbb{R}^{n \times n}$  to be the ordered triple  $(n_+, n_-, n_0)$ , where

- $n_+$  is the number of positive eigenvalues of  $A$  (counting algebraic multiplicities),
- $n_-$  is the number of negative eigenvalues of  $A$  (counting algebraic multiplicities),
- $n_0 := n - n_+ - n_-$ .

## Definition

The *signature* of a symmetric matrix  $A \in \mathbb{R}^{n \times n}$  to be the ordered triple  $(n_+, n_-, n_0)$ , where

- $n_+$  is the number of positive eigenvalues of  $A$  (counting algebraic multiplicities),
  - $n_-$  is the number of negative eigenvalues of  $A$  (counting algebraic multiplicities),
  - $n_0 := n - n_+ - n_-$ .
- Note that 0 is an eigenvalue of  $A$  iff  $n_0 > 0$ , and in this case, the algebraic multiplicity of the eigenvalue 0 is precisely  $n_0$ .



## Definition

The *signature* of a symmetric matrix  $A \in \mathbb{R}^{n \times n}$  to be the ordered triple  $(n_+, n_-, n_0)$ , where

- $n_+$  is the number of positive eigenvalues of  $A$  (counting algebraic multiplicities),
  - $n_-$  is the number of negative eigenvalues of  $A$  (counting algebraic multiplicities),
  - $n_0 := n - n_+ - n_-$ .
- 
- Note that 0 is an eigenvalue of  $A$  iff  $n_0 > 0$ , and in this case, the algebraic multiplicity of the eigenvalue 0 is precisely  $n_0$ .
  - For example, if the spectrum of a symmetric matrix in  $\mathbb{R}^{9 \times 9}$  is  $\{0, 0, 1, 1, -2, -2, 5, 6, -7\}$ , then the signature of that matrix is  $(4, 3, 2)$ .

## Definition

The *signature* of a symmetric matrix  $A \in \mathbb{R}^{n \times n}$  to be the ordered triple  $(n_+, n_-, n_0)$ , where

- $n_+$  is the number of positive eigenvalues of  $A$  (counting algebraic multiplicities),
- $n_-$  is the number of negative eigenvalues of  $A$  (counting algebraic multiplicities),
- $n_0 := n - n_+ - n_-$ .

## Definition

The *signature* of a symmetric matrix  $A \in \mathbb{R}^{n \times n}$  to be the ordered triple  $(n_+, n_-, n_0)$ , where

- $n_+$  is the number of positive eigenvalues of  $A$  (counting algebraic multiplicities),
- $n_-$  is the number of negative eigenvalues of  $A$  (counting algebraic multiplicities),
- $n_0 := n - n_+ - n_-$ .

- Our goal is to prove the following theorem.

## Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

## Definition

The *signature* of a symmetric matrix  $A \in \mathbb{R}^{n \times n}$  to be the ordered triple  $(n_+, n_-, n_0)$ , where

- $n_+$  is the number of positive eigenvalues of  $A$  (counting algebraic multiplicities),
- $n_-$  is the number of negative eigenvalues of  $A$  (counting algebraic multiplicities),
- $n_0 := n - n_+ - n_-$ .

- Our goal is to prove the following theorem.

## Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

- We begin with a proposition, which we will use to prove Theorem 9.4.3

### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

*Proof.*

### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

*Proof.* By the spectral theorem for symmetric matrices, we know that  $A$  is orthogonally diagonalizable.

### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

*Proof.* By the spectral theorem for symmetric matrices, we know that  $A$  is orthogonally diagonalizable. So, let  $D = D(\lambda_1, \dots, \lambda_n)$  be a diagonal and  $Q$  an orthogonal matrix, both in  $\mathbb{R}^{n \times n}$ , s.t.  $D = Q^T A Q$ .

### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

*Proof.* By the spectral theorem for symmetric matrices, we know that  $A$  is orthogonally diagonalizable. So, let  $D = D(\lambda_1, \dots, \lambda_n)$  be a diagonal and  $Q$  an orthogonal matrix, both in  $\mathbb{R}^{n \times n}$ , s.t.  $D = Q^T A Q$ . By Proposition 8.5.12,  $\{\lambda_1, \dots, \lambda_n\}$  is the spectrum of  $A$ .



### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

*Proof.* By the spectral theorem for symmetric matrices, we know that  $A$  is orthogonally diagonalizable. So, let  $D = D(\lambda_1, \dots, \lambda_n)$  be a diagonal and  $Q$  an orthogonal matrix, both in  $\mathbb{R}^{n \times n}$ , s.t.  $D = Q^T A Q$ . By Proposition 8.5.12,  $\{\lambda_1, \dots, \lambda_n\}$  is the spectrum of  $A$ .

After possibly permuting the  $\lambda_i$ 's and the corresponding columns of the orthogonal matrix  $Q$ , we may assume that the first  $n_+$  many  $\lambda_i$ 's are positive, the subsequent  $n_-$  many  $\lambda_i$ 's are negative, and the final  $n_0$  many  $\lambda_i$ 's are 0 (justification: Lecture Notes).

*Proof (continued).* Reminder:  $D = Q^T A Q$ ,  $D = D(\lambda_1, \dots, \lambda_n)$ ,  $Q$  is orthogonal; the first  $n_+$  many  $\lambda_i$ 's are positive, the subsequent  $n_-$  many  $\lambda_i$ 's are negative, and the final  $n_0$  many  $\lambda_i$ 's are 0.

*Proof (continued).* Reminder:  $D = Q^T A Q$ ,  $D = D(\lambda_1, \dots, \lambda_n)$ ,  $Q$  is orthogonal; the first  $n_+$  many  $\lambda_i$ 's are positive, the subsequent  $n_-$  many  $\lambda_i$ 's are negative, and the final  $n_0$  many  $\lambda_i$ 's are 0.

Now, set

$$\ell_i := \begin{cases} \frac{1}{\sqrt{|\lambda_i|}} & \text{if } \lambda_i \neq 0 \\ 1 & \text{if } \lambda_i = 0 \end{cases}$$

for all indices  $i \in \{1, \dots, n\}$ , and set  $L := D(\ell_1, \dots, \ell_n)$  and  $R := QL$ .

*Proof (continued).* Reminder:  $D = Q^T A Q$ ,  $D = D(\lambda_1, \dots, \lambda_n)$ ,  $Q$  is orthogonal; the first  $n_+$  many  $\lambda_i$ 's are positive, the subsequent  $n_-$  many  $\lambda_i$ 's are negative, and the final  $n_0$  many  $\lambda_i$ 's are 0.

Now, set

$$\ell_i := \begin{cases} \frac{1}{\sqrt{|\lambda_i|}} & \text{if } \lambda_i \neq 0 \\ 1 & \text{if } \lambda_i = 0 \end{cases}$$

for all indices  $i \in \{1, \dots, n\}$ , and set  $L := D(\ell_1, \dots, \ell_n)$  and  $R := QL$ . Since both  $Q$  and  $L$  are invertible, so is  $R$ .

- Since  $Q$  is orthogonal, Theorem 6.8.1 guarantees that it is invertible. On the other hand,  $L$  is a diagonal matrix, and all its entries on the main diagonal are non-zero; so, by Proposition 8.5.3(b),  $L$  is invertible.

*Proof (continued).* Reminder:  $D = Q^T A Q$ ,  $D = D(\lambda_1, \dots, \lambda_n)$ ,  $Q$  is orthogonal; the first  $n_+$  many  $\lambda_i$ 's are positive, the subsequent  $n_-$  many  $\lambda_i$ 's are negative, and the final  $n_0$  many  $\lambda_i$ 's are 0.

Now, set

$$\ell_i := \begin{cases} \frac{1}{\sqrt{|\lambda_i|}} & \text{if } \lambda_i \neq 0 \\ 1 & \text{if } \lambda_i = 0 \end{cases}$$

for all indices  $i \in \{1, \dots, n\}$ , and set  $L := D(\ell_1, \dots, \ell_n)$  and  $R := QL$ . Since both  $Q$  and  $L$  are invertible, so is  $R$ .

- Since  $Q$  is orthogonal, Theorem 6.8.1 guarantees that it is invertible. On the other hand,  $L$  is a diagonal matrix, and all its entries on the main diagonal are non-zero; so, by Proposition 8.5.3(b),  $L$  is invertible.

Moreover, since  $L$  is diagonal, Proposition 8.5.1(b) guarantees that the columns of  $R = QL$  are scalar multiples of the columns of  $Q$ ;

*Proof (continued).* Reminder:  $D = Q^T A Q$ ,  $D = D(\lambda_1, \dots, \lambda_n)$ ,  $Q$  is orthogonal; the first  $n_+$  many  $\lambda_i$ 's are positive, the subsequent  $n_-$  many  $\lambda_i$ 's are negative, and the final  $n_0$  many  $\lambda_i$ 's are 0.

Now, set

$$\ell_i := \begin{cases} \frac{1}{\sqrt{|\lambda_i|}} & \text{if } \lambda_i \neq 0 \\ 1 & \text{if } \lambda_i = 0 \end{cases}$$

for all indices  $i \in \{1, \dots, n\}$ , and set  $L := D(\ell_1, \dots, \ell_n)$  and  $R := QL$ . Since both  $Q$  and  $L$  are invertible, so is  $R$ .

- Since  $Q$  is orthogonal, Theorem 6.8.1 guarantees that it is invertible. On the other hand,  $L$  is a diagonal matrix, and all its entries on the main diagonal are non-zero; so, by Proposition 8.5.3(b),  $L$  is invertible.

Moreover, since  $L$  is diagonal, Proposition 8.5.1(b) guarantees that the columns of  $R = QL$  are scalar multiples of the columns of  $Q$ ; since the columns of  $Q$  are pairwise orthogonal (by Theorem 6.8.1), Proposition 6.1.4(b) guarantees that the columns of  $R$  are pairwise orthogonal.

*Proof (continued).* Reminder:  $D = Q^T A Q$ ,  $D = D(\lambda_1, \dots, \lambda_n)$ ,  $Q$  is orthogonal; the first  $n_+$  many  $\lambda_i$ 's are positive, the subsequent  $n_-$  many  $\lambda_i$ 's are negative, and the final  $n_0$  many  $\lambda_i$ 's are 0.

Now, set

$$\ell_i := \begin{cases} \frac{1}{\sqrt{|\lambda_i|}} & \text{if } \lambda_i \neq 0 \\ 1 & \text{if } \lambda_i = 0 \end{cases}$$

for all indices  $i \in \{1, \dots, n\}$ , and set  $L := D(\ell_1, \dots, \ell_n)$  and  $R := QL$ . Since both  $Q$  and  $L$  are invertible, so is  $R$ .

- Since  $Q$  is orthogonal, Theorem 6.8.1 guarantees that it is invertible. On the other hand,  $L$  is a diagonal matrix, and all its entries on the main diagonal are non-zero; so, by Proposition 8.5.3(b),  $L$  is invertible.

Moreover, since  $L$  is diagonal, Proposition 8.5.1(b) guarantees that the columns of  $R = QL$  are scalar multiples of the columns of  $Q$ ; since the columns of  $Q$  are pairwise orthogonal (by Theorem 6.8.1), Proposition 6.1.4(b) guarantees that the columns of  $R$  are pairwise orthogonal. Finally, we compute (next slide):

*Proof (continued).*

$$\begin{aligned} R^T A R &= (QL)^T A (QL) = L^T \underbrace{Q^T A Q}_{=D} L \stackrel{(*)}{=} LDL \\ &= D(\ell_1, \dots, \ell_n) D(\lambda_1, \dots, \lambda_n) D(\ell_1, \dots, \ell_n) \\ &\stackrel{(**)}{=} D(\lambda_1 \ell_1^2, \dots, \lambda_n \ell_n^2), \\ &\stackrel{(***)}{=} D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}), \end{aligned}$$

where (\*) follows from the fact that  $L$  is diagonal and therefore symmetric, (\*\*) follows from Proposition 8.5.2, and (\*\*\*) follows from the fact that, by construction,

$$\lambda_i \ell_i^2 = \begin{cases} 1 & \text{if } \lambda_i > 0 \\ -1 & \text{if } \lambda_i < 0 \\ 0 & \text{if } \lambda_i = 0 \end{cases}$$

for all indices  $i \in \{1, \dots, n\}$ , plus the fact that the first  $n_+$  many  $\lambda_i$ 's are positive, the subsequent  $n_-$  many  $\lambda_i$ 's are negative, and the final  $n_0$  many  $\lambda_i$ 's are zero.  $\square$



### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

- The proof of Proposition 9.4.1 is fully constructive (i.e. it allows us to construct a suitable matrix  $R$ , as long as we are able to factor the characteristic polynomial of  $A$ ).

### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

- The proof of Proposition 9.4.1 is fully constructive (i.e. it allows us to construct a suitable matrix  $R$ , as long as we are able to factor the characteristic polynomial of  $A$ ).
- For a numerical example, see the Lecture Notes.

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof.* Fix symmetric matrices  $B, C \in \mathbb{R}^{n \times n}$ , and suppose first that  $B$  and  $C$  both have the same signature, say  $(n_+, n_-, n_0)$ .

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof.* Fix symmetric matrices  $B, C \in \mathbb{R}^{n \times n}$ , and suppose first that  $B$  and  $C$  both have the same signature, say  $(n_+, n_-, n_0)$ .

Proposition 9.4.1 then guarantees  $B$  and  $C$  are both congruent to the diagonal matrix

$$D := D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof.* Fix symmetric matrices  $B, C \in \mathbb{R}^{n \times n}$ , and suppose first that  $B$  and  $C$  both have the same signature, say  $(n_+, n_-, n_0)$ .

Proposition 9.4.1 then guarantees  $B$  and  $C$  are both congruent to the diagonal matrix

$$D := D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

By Proposition 9.2.6, matrix congruence is an equivalence relation on  $\mathbb{R}^{n \times n}$ ; so, since  $B$  and  $C$  are congruent to the same matrix  $D$ , they are also congruent to each other.

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Suppose, conversely, that  $B$  and  $C$  are congruent.

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Suppose, conversely, that  $B$  and  $C$  are congruent. Let  $(p, q, n - p - q)$  be the signature of  $B$ , and let  $(s, t, n - s - t)$  be the signature of  $C$ ; WTS  $(p, q, n - p - q) = (s, t, n - s - t)$ .



### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Suppose, conversely, that  $B$  and  $C$  are congruent. Let  $(p, q, n - p - q)$  be the signature of  $B$ , and let  $(s, t, n - s - t)$  be the signature of  $C$ ; WTS  $(p, q, n - p - q) = (s, t, n - s - t)$ . Clearly, it suffices to show that  $p = s$  and  $p + q = s + t$ .

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Suppose, conversely, that  $B$  and  $C$  are congruent. Let  $(p, q, n - p - q)$  be the signature of  $B$ , and let  $(s, t, n - s - t)$  be the signature of  $C$ ; WTS  $(p, q, n - p - q) = (s, t, n - s - t)$ . Clearly, it suffices to show that  $p = s$  and  $p + q = s + t$ .

First, by Proposition 9.4.1,  $B$  is congruent to the matrix

$$D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q}),$$

and  $C$  is congruent to the matrix

$$D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t}).$$

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Suppose, conversely, that  $B$  and  $C$  are congruent. Let  $(p, q, n - p - q)$  be the signature of  $B$ , and let  $(s, t, n - s - t)$  be the signature of  $C$ ; WTS  $(p, q, n - p - q) = (s, t, n - s - t)$ . Clearly, it suffices to show that  $p = s$  and  $p + q = s + t$ .

First, by Proposition 9.4.1,  $B$  is congruent to the matrix

$$D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q}),$$

and  $C$  is congruent to the matrix

$$D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t}).$$

Proposition 9.2.6 then guarantees that  $D_B$  and  $D_C$  are congruent to each other.

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Reminder: Matrices

$$\bullet D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$$

$$\bullet D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$$

are congruent to each other; WTS  $p = s$  and  $p + q = s + t$ .

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Reminder: Matrices

$$\bullet D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$$

$$\bullet D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$$

are congruent to each other; WTS  $p = s$  and  $p + q = s + t$ .

By definition, this means that there exists an invertible matrix  $P \in \mathbb{R}^{n \times n}$  s.t.  $D_C = P^T D_B P$ ; we will use this to prove that  $p + q = r + s$ .

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Reminder: Matrices

$$\bullet D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$$

$$\bullet D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$$

are congruent to each other; WTS  $p = s$  and  $p + q = s + t$ .

By definition, this means that there exists an invertible matrix  $P \in \mathbb{R}^{n \times n}$  s.t.  $D_C = P^T D_B P$ ; we will use this to prove that  $p + q = r + s$ .

On the other hand, by Theorem 9.4.1, there exist bases  $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$  and  $\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}$  of  $\mathbb{R}^n$ , as well as a quadratic form  $q$  on  $\mathbb{R}^n$ , s.t.  $D_B$  is the matrix of  $q$  w.r.t.  $\mathcal{B}$ , and  $D_C$  is the matrix of  $q$  w.r.t.  $\mathcal{C}$ ; we will use this to prove that  $p = s$ .

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Reminder:  $D_C = P^T D_B P$ , where

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$ ,
- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$ ,
- $P$  is invertible.

We first show that  $p + q = s + t$ .

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Reminder:  $D_C = P^T D_B P$ , where

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$ ,
- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$ ,
- $P$  is invertible.

We first show that  $p + q = s + t$ . Clearly,  $\text{rank}(D_B) = p + q$  and  $\text{rank}(D_C) = s + t$ , and so it is enough to show that  $\text{rank}(D_B) = \text{rank}(D_C)$ .



### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Reminder:  $D_C = P^T D_B P$ , where

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$ ,
- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$ ,
- $P$  is invertible.

We first show that  $p + q = s + t$ . Clearly,  $\text{rank}(D_B) = p + q$  and  $\text{rank}(D_C) = s + t$ , and so it is enough to show that  $\text{rank}(D_B) = \text{rank}(D_C)$ . Since the matrix  $P$  is invertible, the Invertible Matrix Theorem guarantees that  $P^T$  is also invertible.

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

*Proof (continued).* Reminder:  $D_C = P^T D_B P$ , where

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$ ,
- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$ ,
- $P$  is invertible.

We first show that  $p + q = s + t$ . Clearly,  $\text{rank}(D_B) = p + q$  and  $\text{rank}(D_C) = s + t$ , and so it is enough to show that  $\text{rank}(D_B) = \text{rank}(D_C)$ . Since the matrix  $P$  is invertible, the Invertible Matrix Theorem guarantees that  $P^T$  is also invertible. But then

$$\text{rank}(D_C) = \text{rank}(P^T D_B P) \stackrel{(*)}{=} \text{rank}(D_B),$$

where (\*) follows from Proposition 3.3.14 (since  $P^T$  and  $P$  are both invertible).

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}.$$

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}.$$

It remains to show that  $p = s$ .

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}.$$

It remains to show that  $p = s$ . Suppose otherwise. By symmetry, we may assume that  $p > s$ .

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}.$$

It remains to show that  $p = s$ . Suppose otherwise. By symmetry, we may assume that  $p > s$ . Now consider the subspaces  $U_B := \text{Span}(\mathbf{b}_1, \dots, \mathbf{b}_p)$  and  $U_C := \text{Span}(\mathbf{c}_{s+1}, \dots, \mathbf{c}_n)$  of  $\mathbb{R}^n$ .

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}.$$

It remains to show that  $p = s$ . Suppose otherwise. By symmetry, we may assume that  $p > s$ . Now consider the subspaces

$U_B := \text{Span}(\mathbf{b}_1, \dots, \mathbf{b}_p)$  and  $U_C := \text{Span}(\mathbf{c}_{s+1}, \dots, \mathbf{c}_n)$  of  $\mathbb{R}^n$ .

Then by Theorem 3.2.23, we have that

$$\dim(U_B) + \dim(U_C) = \dim(U_B + U_C) + \dim(U_B \cap U_C).$$

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}.$$

It remains to show that  $p = s$ . Suppose otherwise. By symmetry, we may assume that  $p > s$ . Now consider the subspaces

$U_B := \text{Span}(\mathbf{b}_1, \dots, \mathbf{b}_p)$  and  $U_C := \text{Span}(\mathbf{c}_{s+1}, \dots, \mathbf{c}_n)$  of  $\mathbb{R}^n$ .

Then by Theorem 3.2.23, we have that

$$\dim(U_B) + \dim(U_C) = \dim(U_B + U_C) + \dim(U_B \cap U_C).$$

But note that

- $\dim(U_B) + \dim(U_C) = p + (n - s) = n + (p - s) > n$ ,
- $\dim(U_B + U_C) \leq \dim(\mathbb{R}^n) = n$ .



*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\}.$$

It remains to show that  $p = s$ . Suppose otherwise. By symmetry, we may assume that  $p > s$ . Now consider the subspaces

$U_B := \text{Span}(\mathbf{b}_1, \dots, \mathbf{b}_p)$  and  $U_C := \text{Span}(\mathbf{c}_{s+1}, \dots, \mathbf{c}_n)$  of  $\mathbb{R}^n$ .

Then by Theorem 3.2.23, we have that

$$\dim(U_B) + \dim(U_C) = \dim(U_B + U_C) + \dim(U_B \cap U_C).$$

But note that

- $\dim(U_B) + \dim(U_C) = p + (n - s) = n + (p - s) > n$ ,
- $\dim(U_B + U_C) \leq \dim(\mathbb{R}^n) = n$ .

So,  $\dim(U_B \cap U_C) > 0$ , and it follows that  $U_B \cap U_C$  contains some non-zero vector  $\mathbf{u}$ .

*Proof (continued).*

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\},$$

- $U_B := \text{Span}(\mathbf{b}_1, \dots, \mathbf{b}_p)$ ,  $U_C := \text{Span}(\mathbf{c}_{s+1}, \dots, \mathbf{c}_n)$ ,
- $\mathbf{u} \in U_B \cap U_C$ ,  $\mathbf{u} \neq \mathbf{0}$ .

Set  $[\mathbf{u}]_{\mathcal{B}} = [x_1 \ \dots \ x_n]^T$  and  $[\mathbf{u}]_{\mathcal{C}} = [y_1 \ \dots \ y_n]^T$ .

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\},$$

- $U_B := \text{Span}(\mathbf{b}_1, \dots, \mathbf{b}_p)$ ,  $U_C := \text{Span}(\mathbf{c}_{s+1}, \dots, \mathbf{c}_n)$ ,
- $\mathbf{u} \in U_B \cap U_C$ ,  $\mathbf{u} \neq \mathbf{0}$ .

Set  $[\mathbf{u}]_{\mathcal{B}} = [x_1 \ \dots \ x_n]^T$  and  $[\mathbf{u}]_{\mathcal{C}} = [y_1 \ \dots \ y_n]^T$ .

Then at least one of  $x_1, \dots, x_p$  is non-zero,  $x_{p+1} = \dots = x_n = 0$ ,  
and  $y_1 = \dots = y_s = 0$ .

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\},$$

- $U_B := \text{Span}(\mathbf{b}_1, \dots, \mathbf{b}_p)$ ,  $U_C := \text{Span}(\mathbf{c}_{s+1}, \dots, \mathbf{c}_n)$ ,
- $\mathbf{u} \in U_B \cap U_C$ ,  $\mathbf{u} \neq \mathbf{0}$ .

Set  $[\mathbf{u}]_{\mathcal{B}} = [x_1 \ \dots \ x_n]^T$  and  $[\mathbf{u}]_{\mathcal{C}} = [y_1 \ \dots \ y_n]^T$ .

Then at least one of  $x_1, \dots, x_p$  is non-zero,  $x_{p+1} = \dots = x_n = 0$ , and  $y_1 = \dots = y_s = 0$ . We now have that

- $q(\mathbf{u}) = [\mathbf{u}]_{\mathcal{B}}^T D_B [\mathbf{u}]_{\mathcal{B}} \stackrel{(*)}{=} x_1^2 + \dots + x_p^2 > 0$ ,

- $q(\mathbf{u}) = [\mathbf{u}]_{\mathcal{C}}^T D_C [\mathbf{u}]_{\mathcal{C}} \stackrel{(*)}{=} -y_{s+1}^2 - \dots - y_{s+t}^2 \leq 0$ ,

where for both instances of (\*), we used the formula from Proposition 9.1.1(a).

*Proof (continued).* Reminder:

- $D_B := D(\underbrace{1, \dots, 1}_p, \underbrace{-1, \dots, -1}_q, \underbrace{0, \dots, 0}_{n-p-q})$  is the matrix of  $q$  w.r.t.

$$\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\},$$

- $D_C := D(\underbrace{1, \dots, 1}_s, \underbrace{-1, \dots, -1}_t, \underbrace{0, \dots, 0}_{n-s-t})$  is the matrix of  $q$  w.r.t.

$$\mathcal{C} = \{\mathbf{c}_1, \dots, \mathbf{c}_n\},$$

- $U_B := \text{Span}(\mathbf{b}_1, \dots, \mathbf{b}_p)$ ,  $U_C := \text{Span}(\mathbf{c}_{s+1}, \dots, \mathbf{c}_n)$ ,
- $\mathbf{u} \in U_B \cap U_C$ ,  $\mathbf{u} \neq \mathbf{0}$ .

Set  $[\mathbf{u}]_{\mathcal{B}} = [x_1 \ \dots \ x_n]^T$  and  $[\mathbf{u}]_{\mathcal{C}} = [y_1 \ \dots \ y_n]^T$ .

Then at least one of  $x_1, \dots, x_p$  is non-zero,  $x_{p+1} = \dots = x_n = 0$ , and  $y_1 = \dots = y_s = 0$ . We now have that

- $q(\mathbf{u}) = [\mathbf{u}]_{\mathcal{B}}^T D_B [\mathbf{u}]_{\mathcal{B}} \stackrel{(*)}{=} x_1^2 + \dots + x_p^2 > 0,$

- $q(\mathbf{u}) = [\mathbf{u}]_{\mathcal{C}}^T D_C [\mathbf{u}]_{\mathcal{C}} \stackrel{(*)}{=} -y_{s+1}^2 - \dots - y_{s+t}^2 \leq 0,$

where for both instances of  $(*)$ , we used the formula from Proposition 9.1.1(a). We have now derived a contradiction, and it follows that  $p = s$ . This completes the argument.  $\square$

### Proposition 9.4.1

Let  $A$  be a symmetric matrix in  $\mathbb{R}^{n \times n}$  with signature  $(n_+, n_-, n_0)$ . Then there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.

$$R^T A R = D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

### Theorem 9.4.3

Two symmetric matrices in  $\mathbb{R}^{n \times n}$  are congruent iff they have the same signature.

- Suppose that  $\mathbb{F}$  is a field and that  $D = D(a_1, \dots, a_n)$  is a diagonal matrix in  $\mathbb{F}^{n \times n}$ .



- Suppose that  $\mathbb{F}$  is a field and that  $D = D(a_1, \dots, a_n)$  is a diagonal matrix in  $\mathbb{F}^{n \times n}$ .

- Then for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  in  $\mathbb{F}^n$ , we have that

$$\mathbf{x}^T D \mathbf{x} = a_1 x_1^2 + \dots + a_n x_n^2,$$

as can be seen via routine computation, or by applying Proposition 9.1.1(a).

- Suppose that  $\mathbb{F}$  is a field and that  $D = D(a_1, \dots, a_n)$  is a diagonal matrix in  $\mathbb{F}^{n \times n}$ .

- Then for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  in  $\mathbb{F}^n$ , we have that

$$\mathbf{x}^T D \mathbf{x} = a_1 x_1^2 + \dots + a_n x_n^2,$$

as can be seen via routine computation, or by applying Proposition 9.1.1(a).

- This is a particularly nice formula, and for this reason, if  $q$  is a quadratic form over a field  $\mathbb{F}$ , it is helpful to have a basis  $\mathcal{B}$  with respect to which the matrix of  $q$  is diagonal.

- Suppose that  $\mathbb{F}$  is a field and that  $D = D(a_1, \dots, a_n)$  is a diagonal matrix in  $\mathbb{F}^{n \times n}$ .

- Then for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  in  $\mathbb{F}^n$ , we have that

$$\mathbf{x}^T D \mathbf{x} = a_1 x_1^2 + \dots + a_n x_n^2,$$

as can be seen via routine computation, or by applying Proposition 9.1.1(a).

- This is a particularly nice formula, and for this reason, if  $q$  is a quadratic form over a field  $\mathbb{F}$ , it is helpful to have a basis  $\mathcal{B}$  with respect to which the matrix of  $q$  is diagonal.
- Sylvester's law of inertia (in a couple of slides) states that when  $V = \mathbb{R}^n$ , such a basis  $\mathcal{B}$  always exists.

- Suppose that  $\mathbb{F}$  is a field and that  $D = D(a_1, \dots, a_n)$  is a diagonal matrix in  $\mathbb{F}^{n \times n}$ .

- Then for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  in  $\mathbb{F}^n$ , we have that

$$\mathbf{x}^T D \mathbf{x} = a_1 x_1^2 + \dots + a_n x_n^2,$$

as can be seen via routine computation, or by applying Proposition 9.1.1(a).

- This is a particularly nice formula, and for this reason, if  $q$  is a quadratic form over a field  $\mathbb{F}$ , it is helpful to have a basis  $\mathcal{B}$  with respect to which the matrix of  $q$  is diagonal.
- Sylvester's law of inertia (in a couple of slides) states that when  $V = \mathbb{R}^n$ , such a basis  $\mathcal{B}$  always exists.
- As we shall see, Sylvester's law of inertia is essentially a "translation" of Proposition 9.4.1 and Theorem 9.4.3 into the language of quadratic forms.

- Suppose that  $\mathbb{F}$  is a field and that  $D = D(a_1, \dots, a_n)$  is a diagonal matrix in  $\mathbb{F}^{n \times n}$ .

- Then for all vectors  $\mathbf{x} = [x_1 \ \dots \ x_n]^T$  in  $\mathbb{F}^n$ , we have that

$$\mathbf{x}^T D \mathbf{x} = a_1 x_1^2 + \dots + a_n x_n^2,$$

as can be seen via routine computation, or by applying Proposition 9.1.1(a).

- This is a particularly nice formula, and for this reason, if  $q$  is a quadratic form over a field  $\mathbb{F}$ , it is helpful to have a basis  $\mathcal{B}$  with respect to which the matrix of  $q$  is diagonal.
- Sylvester's law of inertia (in a couple of slides) states that when  $V = \mathbb{R}^n$ , such a basis  $\mathcal{B}$  always exists.
- As we shall see, Sylvester's law of inertia is essentially a "translation" of Proposition 9.4.1 and Theorem 9.4.3 into the language of quadratic forms.
- Before formally stating and proving the law, we need a definition.

## Definition

The *signature* of a quadratic form  $q$  on  $\mathbb{R}^n$  is defined to be the signature of the matrix of  $q$  with respect to **any** basis  $\mathcal{B}$  of  $\mathbb{R}^n$ . A *polar basis* of  $\mathbb{R}^n$  associated with the quadratic form  $q$  is any **orthogonal** basis  $\mathcal{B}$  of  $\mathbb{R}^n$  s.t. the matrix of  $q$  w.r.t.  $\mathcal{B}$  is a diagonal matrix with only 1's,  $-1$ 's, and 0's on the main diagonal.

- By Theorems 9.3.4 and 9.4.3, the signature of  $q$  is well defined.
  - Indeed, by Theorem 9.3.4, matrices of  $q$  with respect to all possible bases of  $\mathbb{R}^n$  are congruent to each other, and so by Theorem 9.4.3, they all have the same signature.

## Sylvester's law of inertia

Let  $q$  be a quadratic form on  $\mathbb{R}^n$ , and let  $(n_+, n_-, n_0)$  be the signature of  $q$ . Then  $\mathbb{R}^n$  has a polar basis  $\mathcal{B}$  associated with  $q$ . Moreover, for any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  s.t. the matrix  $C$  of  $q$  with respect to  $\mathcal{C}$  is diagonal, with only 1's,  $-1$ 's, and 0's on the main diagonal, the following holds: the number of 1's,  $-1$ 's, and 0's on the main diagonal of  $C$  is  $n_+$ ,  $n_-$ , and  $n_0$ , respectively.

- **Remark:** The basis  $\mathcal{C}$  from the second sentence of Sylvester's law of inertia is not assumed to be polar, i.e. it is possible that it is not orthogonal.

## Sylvester's law of inertia

Let  $q$  be a quadratic form on  $\mathbb{R}^n$ , and let  $(n_+, n_-, n_0)$  be the signature of  $q$ . Then  $\mathbb{R}^n$  has a polar basis  $\mathcal{B}$  associated with  $q$ . Moreover, for any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  s.t. the matrix  $C$  of  $q$  with respect to  $\mathcal{C}$  is diagonal, with only 1's,  $-1$ 's, and 0's on the main diagonal, the following holds: the number of 1's,  $-1$ 's, and 0's on the main diagonal of  $C$  is  $n_+$ ,  $n_-$ , and  $n_0$ , respectively.

- **Remark:** The basis  $\mathcal{C}$  from the second sentence of Sylvester's law of inertia is not assumed to be polar, i.e. it is possible that it is not orthogonal.
- Let's prove the theorem!



*Proof.* Let  $A$  be the matrix of the quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{R}^n$ ; then the signature of  $A$  is  $(n_+, n_-, n_0)$ .

*Proof.* Let  $A$  be the matrix of the quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{R}^n$ ; then the signature of  $A$  is  $(n_+, n_-, n_0)$ .

We first prove the existence of the polar basis  $\mathcal{B}$ .

*Proof.* Let  $A$  be the matrix of the quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{R}^n$ ; then the signature of  $A$  is  $(n_+, n_-, n_0)$ .

We first prove the existence of the polar basis  $\mathcal{B}$ . Set

$$D := D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

*Proof.* Let  $A$  be the matrix of the quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{R}^n$ ; then the signature of  $A$  is  $(n_+, n_-, n_0)$ .

We first prove the existence of the polar basis  $\mathcal{B}$ . Set

$$D := D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

By Proposition 9.4.1, there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.  $D = R^T A R$ . Since  $R$  is invertible, the Invertible Matrix Theorem guarantees that its columns form a basis  $\mathcal{B}$  of  $\mathbb{R}^n$ ;

*Proof.* Let  $A$  be the matrix of the quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{R}^n$ ; then the signature of  $A$  is  $(n_+, n_-, n_0)$ .

We first prove the existence of the polar basis  $\mathcal{B}$ . Set

$$D := D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

By Proposition 9.4.1, there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.  $D = R^T A R$ . Since  $R$  is invertible, the Invertible Matrix Theorem guarantees that its columns form a basis  $\mathcal{B}$  of  $\mathbb{R}^n$ ; since the columns of  $R$  are pairwise orthogonal, the basis  $\mathcal{B}$  is orthogonal.

*Proof.* Let  $A$  be the matrix of the quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{R}^n$ ; then the signature of  $A$  is  $(n_+, n_-, n_0)$ .

We first prove the existence of the polar basis  $\mathcal{B}$ . Set

$$D := D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

By Proposition 9.4.1, there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.  $D = R^T A R$ . Since  $R$  is invertible, the Invertible Matrix Theorem guarantees that its columns form a basis  $\mathcal{B}$  of  $\mathbb{R}^n$ ; since the columns of  $R$  are pairwise orthogonal, the basis  $\mathcal{B}$  is orthogonal. Moreover, by Theorem 4.5.1 (or alternatively, by Lemma 4.5.8), we have that  $R = \varepsilon_n [\text{Id}_V]_{\mathcal{B}}$ , so that

$$D = \varepsilon_n [\text{Id}_V]_{\mathcal{B}}^T A \varepsilon_n [\text{Id}_V]_{\mathcal{B}}.$$

But now Theorem 9.3.3 guarantees that  $D$  is the matrix of  $q$  with respect to  $\mathcal{B}$ .

*Proof.* Let  $A$  be the matrix of the quadratic form  $q$  with respect to the standard basis  $\mathcal{E}_n$  of  $\mathbb{R}^n$ ; then the signature of  $A$  is  $(n_+, n_-, n_0)$ .

We first prove the existence of the polar basis  $\mathcal{B}$ . Set

$$D := D(\underbrace{1, \dots, 1}_{n_+}, \underbrace{-1, \dots, -1}_{n_-}, \underbrace{0, \dots, 0}_{n_0}).$$

By Proposition 9.4.1, there exists an invertible matrix  $R \in \mathbb{R}^{n \times n}$  with pairwise orthogonal columns s.t.  $D = R^T A R$ . Since  $R$  is invertible, the Invertible Matrix Theorem guarantees that its columns form a basis  $\mathcal{B}$  of  $\mathbb{R}^n$ ; since the columns of  $R$  are pairwise orthogonal, the basis  $\mathcal{B}$  is orthogonal. Moreover, by Theorem 4.5.1 (or alternatively, by Lemma 4.5.8), we have that  $R = \varepsilon_n [\text{Id}_V]_{\mathcal{B}}$ , so that

$$D = \varepsilon_n [\text{Id}_V]_{\mathcal{B}}^T A \varepsilon_n [\text{Id}_V]_{\mathcal{B}}.$$

But now Theorem 9.3.3 guarantees that  $D$  is the matrix of  $q$  with respect to  $\mathcal{B}$ . We have already seen that the basis  $\mathcal{B}$  is orthogonal, and we deduce that  $\mathcal{B}$  is a polar basis of  $\mathbb{R}^n$  associated with  $q$ .

## Sylvester's law of inertia

Let  $q$  be a quadratic form on  $\mathbb{R}^n$ , and let  $(n_+, n_-, n_0)$  be the signature of  $q$ . Then  $\mathbb{R}^n$  has a polar basis  $\mathcal{B}$  associated with  $q$ . Moreover, for any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  s.t. the matrix  $C$  of  $q$  with respect to  $\mathcal{C}$  is diagonal, with only 1's,  $-1$ 's, and 0's on the main diagonal, the following holds: the number of 1's,  $-1$ 's, and 0's on the main diagonal of  $C$  is  $n_+$ ,  $n_-$ , and  $n_0$ , respectively.

*Proof (continued).* Now, fix any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  such that the matrix of  $q$  with respect to  $\mathcal{C}$  is a diagonal matrix  $C$  with only 1's,  $-1$ 's, and 0's on the main diagonal.



## Sylvester's law of inertia

Let  $q$  be a quadratic form on  $\mathbb{R}^n$ , and let  $(n_+, n_-, n_0)$  be the signature of  $q$ . Then  $\mathbb{R}^n$  has a polar basis  $\mathcal{B}$  associated with  $q$ . Moreover, for any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  s.t. the matrix  $C$  of  $q$  with respect to  $\mathcal{C}$  is diagonal, with only 1's,  $-1$ 's, and 0's on the main diagonal, the following holds: the number of 1's,  $-1$ 's, and 0's on the main diagonal of  $C$  is  $n_+$ ,  $n_-$ , and  $n_0$ , respectively.

*Proof (continued).* Now, fix any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  such that the matrix of  $q$  with respect to  $\mathcal{C}$  is a diagonal matrix  $C$  with only 1's,  $-1$ 's, and 0's on the main diagonal. By Theorem 9.3.4, matrices  $A$  and  $C$  are congruent, and so by Theorem 9.4.3, they have the same signature, which is  $(n_+, n_-, n_0)$ .

## Sylvester's law of inertia

Let  $q$  be a quadratic form on  $\mathbb{R}^n$ , and let  $(n_+, n_-, n_0)$  be the signature of  $q$ . Then  $\mathbb{R}^n$  has a polar basis  $\mathcal{B}$  associated with  $q$ . Moreover, for any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  s.t. the matrix  $C$  of  $q$  with respect to  $\mathcal{C}$  is diagonal, with only 1's,  $-1$ 's, and 0's on the main diagonal, the following holds: the number of 1's,  $-1$ 's, and 0's on the main diagonal of  $C$  is  $n_+$ ,  $n_-$ , and  $n_0$ , respectively.

*Proof (continued).* Now, fix any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  such that the matrix of  $q$  with respect to  $\mathcal{C}$  is a diagonal matrix  $C$  with only 1's,  $-1$ 's, and 0's on the main diagonal. By Theorem 9.3.4, matrices  $A$  and  $C$  are congruent, and so by Theorem 9.4.3, they have the same signature, which is  $(n_+, n_-, n_0)$ . Since the matrix  $C$  is diagonal, we know its entries on the main diagonal form its spectrum (this follows from Proposition 8.2.7);

## Sylvester's law of inertia

Let  $q$  be a quadratic form on  $\mathbb{R}^n$ , and let  $(n_+, n_-, n_0)$  be the signature of  $q$ . Then  $\mathbb{R}^n$  has a polar basis  $\mathcal{B}$  associated with  $q$ . Moreover, for any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  s.t. the matrix  $C$  of  $q$  with respect to  $\mathcal{C}$  is diagonal, with only 1's,  $-1$ 's, and 0's on the main diagonal, the following holds: the number of 1's,  $-1$ 's, and 0's on the main diagonal of  $C$  is  $n_+$ ,  $n_-$ , and  $n_0$ , respectively.

*Proof (continued).* Now, fix any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  such that the matrix of  $q$  with respect to  $\mathcal{C}$  is a diagonal matrix  $C$  with only 1's,  $-1$ 's, and 0's on the main diagonal. By Theorem 9.3.4, matrices  $A$  and  $C$  are congruent, and so by Theorem 9.4.3, they have the same signature, which is  $(n_+, n_-, n_0)$ . Since the matrix  $C$  is diagonal, we know its entries on the main diagonal form its spectrum (this follows from Proposition 8.2.7); so, the number of 1's,  $-1$ 's, and 0's on the main diagonal of  $C$  is  $n_+$ ,  $n_-$ , and  $n_0$ , respectively.  $\square$

## Sylvester's law of inertia

Let  $q$  be a quadratic form on  $\mathbb{R}^n$ , and let  $(n_+, n_-, n_0)$  be the signature of  $q$ . Then  $\mathbb{R}^n$  has a polar basis  $\mathcal{B}$  associated with  $q$ . Moreover, for any basis  $\mathcal{C}$  of  $\mathbb{R}^n$  s.t. the matrix  $C$  of  $q$  with respect to  $\mathcal{C}$  is diagonal, with only 1's,  $-1$ 's, and 0's on the main diagonal, the following holds: the number of 1's,  $-1$ 's, and 0's on the main diagonal of  $C$  is  $n_+$ ,  $n_-$ , and  $n_0$ , respectively.

- For a numerical example, see the Lecture Notes.

- For quadratic forms on  $\mathbb{R}^2$ , there exist only six possible signatures  $(n_+, n_-, n_0)$ , namely, the following:

- $(2, 0, 0)$ ;

- $(1, 0, 1)$ ;

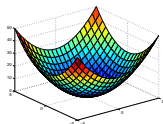
- $(1, 1, 0)$ ;

- $(0, 2, 0)$ ;

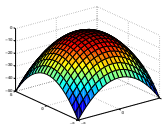
- $(0, 1, 1)$ ;

- $(0, 0, 2)$ .

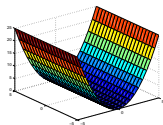
- For quadratic forms on  $\mathbb{R}^2$ , there exist only six possible signatures  $(n_+, n_-, n_0)$ , namely, the following:
  - $(2, 0, 0)$ ;
  - $(1, 0, 1)$ ;
  - $(1, 1, 0)$ ;
  - $(0, 2, 0)$ ;
  - $(0, 1, 1)$ ;
  - $(0, 0, 2)$ .
- Thus, the graph of any quadratic form  $q$  on  $\mathbb{R}^2$  has the same general shape as one of the six graphs shown on the next slide (the one that has the same signature as  $q$ ).



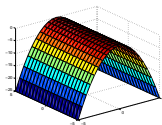
$$x_1^2 + x_2^2$$



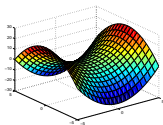
$$-x_1^2 - x_2^2$$



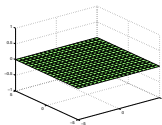
$$x_1^2$$



$$-x_1^2$$



$$x_1^2 - x_2^2$$



$$0$$

- The graphs were generated by Milan Hladík, who kindly shared them with me.

- The actual graph of the quadratic form  $q$  would be obtained by starting with one of the six graphs from the previous slide (the one that has the same signature as  $q$ ), and then possibly stretching or shrinking the graph along the  $x_1$ - and  $x_2$ -axes (the coordinate axes of the domain), and then possibly rotating it about the vertical axis  $x_3$ .
  - This to account for the fact that a polar basis  $\mathcal{B}$  of  $\mathbb{R}^2$  associated with  $q$  is not necessarily equal to the standard basis  $\mathcal{E}_2 = \{\mathbf{e}_1, \mathbf{e}_2\}$ , but the vectors of  $\mathcal{B}$  are indeed orthogonal to each other (by the definition of a polar basis).