

Definition 23.19:

- (a) A *linear form* on \mathbb{R}^n is a function of the form $L(x) = \sum_{i=1}^n a_i x_i$ (for some constants a_1, \dots, a_n).
- (b) A *quadratic form* on \mathbb{R}^n is a function of the form $Q(x) = \sum_{i=1}^n \sum_{j=1}^n b_{ij} x_i x_j$ (for some constants b_{ij}).

Example 23.20:

- (a) We can define a linear form on \mathbb{R}^3 by $L(x) = 7x_1 - 8x_2 + 9x_3$.
- (b) We can define a quadratic form on \mathbb{R}^4 by $Q(x) = 10x_1x_2 + 12x_3x_4 - 14x_1x_4 - 16x_2x_3$.

Given a linear form $L(x) = \sum_i a_i x_i$, we can form the vector $a = [a_1 \ \dots \ a_n]^T$, and clearly $L(x) = a \cdot x = a^T x$.

Symmetric expressions for quadratic forms

Consider a quadratic form $Q(x) = \sum_{i,j} b_{ij} x_i x_j$.

Form the matrix B with entries b_{ij} : we find that $Q(x) = x^T Bx$.

For example, if $n = 2$ and $Q(x) = 2x_1^2 + 4x_1x_2 + 7x_2^2$ then $B = \begin{bmatrix} 2 & 4 \\ 0 & 7 \end{bmatrix}$ and

$$x^T Bx = [x_1 \ x_2] \begin{bmatrix} 2 & 4 \\ 0 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = [x_1 \ x_2] \begin{bmatrix} 2x_1 + 4x_2 \\ 7x_2 \end{bmatrix} = 2x_1^2 + 4x_1x_2 + 7x_2^2 = Q(x).$$

Alternatively, note that $x_1x_2 = x_2x_1$:

- (a) Rewriting the same $Q(x)$ as $2x_1^2 + 3x_1x_2 + 1x_2x_1 + 7x_2^2$ gives $B = \begin{bmatrix} 2 & 3 \\ 1 & 7 \end{bmatrix}$.
- (b) Rewriting the same $Q(x)$ as $2x_1^2 + 2x_1x_2 + 2x_2x_1 + 7x_2^2$ gives $B = \begin{bmatrix} 2 & 2 \\ 2 & 7 \end{bmatrix}$.
- (c) Rewriting the same $Q(x)$ as $2x_1^2 + 1x_1x_2 + 3x_2x_1 + 7x_2^2$ gives $B = \begin{bmatrix} 2 & 1 \\ 3 & 7 \end{bmatrix}$.

In option (b) we “share the coefficient equally” between x_1x_2 and x_2x_1 , so the matrix B is symmetric. This is the preferred option.

We can do the same for any quadratic form.

Symmetric expressions for quadratic forms

For example, we considered above the quadratic form

$$Q(x) = 10x_1x_2 + 12x_3x_4 - 14x_1x_4 - 16x_2x_3.$$

This can be rewritten symmetrically as

$$Q(x) = 5x_1x_2 + 5x_2x_1 + 6x_3x_4 + 6x_4x_3 - 7x_1x_4 - 7x_4x_1 - 8x_2x_3 - 8x_3x_2,$$

which corresponds to the symmetric matrix

$$B = \begin{bmatrix} 0 & 5 & 0 & -7 \\ 5 & 0 & -8 & 0 \\ 0 & -8 & 0 & 6 \\ -7 & 0 & 6 & 0 \end{bmatrix}$$

$$\begin{aligned} x^T Bx &= [x_1 \ x_2 \ x_3 \ x_4] \begin{bmatrix} 0 & 5 & 0 & -7 \\ 5 & 0 & -8 & 0 \\ 0 & -8 & 0 & 6 \\ -7 & 0 & 6 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} \\ &= [x_1 \ x_2 \ x_3 \ x_4] \begin{bmatrix} 5x_2 - 7x_4 \\ 5x_1 - 8x_3 \\ -8x_2 + 6x_4 \\ -7x_1 + 6x_3 \end{bmatrix} = Q(x). \end{aligned}$$

Diagonalisation of quadratic forms

Proposition 23.23: Let $Q(x)$ be a quadratic form on \mathbb{R}^n . Then there are integers $r, s \geq 0$ and nonzero vectors $v_1, \dots, v_r, w_1, \dots, w_s$ such that all the v 's and w 's are orthogonal to each other, and

$$Q(x) = (x \cdot v_1)^2 + \dots + (x \cdot v_r)^2 - (x \cdot w_1)^2 - \dots - (x \cdot w_s)^2.$$

Or in terms of linear forms $L_i(x) = x \cdot v_i$ and $M_j(x) = x \cdot w_j$:

$$Q = L_1^2 + \dots + L_r^2 - M_1^2 - \dots - M_s^2.$$

The *rank* of Q is defined to be $r + s$, and the *signature* is defined to be $r - s$.

Proof: There is a symmetric matrix B such that $Q(x) = x^T B x$. By Proposition 23.12, we can find an orthonormal basis u_1, \dots, u_n for \mathbb{R}^n such that each u_i is an eigenvector for B , with eigenvalue $\lambda_i \in \mathbb{R}$ say. Let r be the number of indices i for which $\lambda_i > 0$, and let s be the number of indices i for which $\lambda_i < 0$. We can assume that things have been ordered such that $\lambda_1, \dots, \lambda_r > 0$ and $\lambda_{r+1}, \dots, \lambda_{r+s} < 0$ and any eigenvalues after λ_{r+s} are zero. Now put $U = [u_1 | \dots | u_n]$ and $D = \text{diag}(\lambda_1, \dots, \lambda_n)$. We have seen that $B = UDU^T$, so

$$Q(x) = x^T B x = x^T UDU^T x = (U^T x)^T (DU^T x) = (U^T x) \cdot (DU^T x).$$

Diagonalisation of quadratic forms

$$U = [u_1 | \dots | u_n] \quad D = \text{diag}(\lambda_1, \dots, \lambda_n) \quad Q(x) = (U^T x) \cdot (DU^T x)$$

$$U^T x = \begin{bmatrix} u_1^T \\ \vdots \\ u_n^T \end{bmatrix} x = \begin{bmatrix} u_1 \cdot x \\ \vdots \\ u_n \cdot x \end{bmatrix}$$

$$DU^T x = \text{diag}(\lambda_1, \dots, \lambda_n) \begin{bmatrix} u_1 \cdot x \\ \vdots \\ u_n \cdot x \end{bmatrix} = \begin{bmatrix} \lambda_1 u_1 \cdot x \\ \vdots \\ \lambda_n u_n \cdot x \end{bmatrix}$$

$$Q(x) = (U^T x) \cdot (DU^T x) = \begin{bmatrix} u_1 \cdot x \\ \vdots \\ u_n \cdot x \end{bmatrix} \cdot \begin{bmatrix} \lambda_1 u_1 \cdot x \\ \vdots \\ \lambda_n u_n \cdot x \end{bmatrix} = \lambda_1 (u_1 \cdot x)^2 + \dots + \lambda_n (u_n \cdot x)^2.$$

Diagonalisation of quadratic forms

$$U = [u_1 | \dots | u_n] \quad D = \text{diag}(\lambda_1, \dots, \lambda_n) \quad Q(x) = \lambda_1 (u_1 \cdot x)^2 + \dots + \lambda_n (u_n \cdot x)^2.$$

$$\lambda_1, \dots, \lambda_r > 0 \quad \lambda_{r+1}, \dots, \lambda_{r+s} < 0 \quad \lambda_{r+s+1}, \dots, \lambda_n = 0$$

- ▶ For $1 \leq i \leq r$ we have $\lambda_i > 0$ and we put $v_i = \sqrt{\lambda_i} u_i$ so $\lambda_i (u_i \cdot x)^2 = (v_i \cdot x)^2$.
- ▶ For $r+1 \leq i \leq r+s$ we have $\lambda_i < 0$ and we put $w_{i-r} = \sqrt{|\lambda_i|} u_i$ so $\lambda_i (u_i \cdot x)^2 = -(w_{i-r} \cdot x)^2$.
- ▶ For $i > r+s$ we have $\lambda_i = 0$ and $\lambda_i (u_i \cdot x)^2 = 0$.

We thus have

$$Q(x) = (x \cdot v_1)^2 + \dots + (x \cdot v_r)^2 - (x \cdot w_1)^2 - \dots - (x \cdot w_s)^2$$

as required.

Example of diagonalising a quadratic form

Consider the quadratic form $Q(x) = x_1 x_2 - x_3 x_4$ on \mathbb{R}^4 . It is elementary that for all $a, b \in \mathbb{R}$ we have

$$\left(\frac{a+b}{2}\right)^2 - \left(\frac{a-b}{2}\right)^2 = \frac{a^2 + 2ab + b^2 - a^2 + 2ab - b^2}{4} = ab.$$

Using this, we can rewrite $Q(x)$ as

$$Q(x) = \left(\frac{x_1 + x_2}{2}\right)^2 - \left(\frac{x_1 - x_2}{2}\right)^2 - \left(\frac{x_3 + x_4}{2}\right)^2 + \left(\frac{x_3 - x_4}{2}\right)^2.$$

Now put

$$v_1 = \begin{bmatrix} 1/2 \\ 1/2 \\ 0 \\ 0 \end{bmatrix} \quad v_2 = \begin{bmatrix} 0 \\ 0 \\ 1/2 \\ -1/2 \end{bmatrix} \quad w_1 = \begin{bmatrix} 1/2 \\ -1/2 \\ 0 \\ 0 \end{bmatrix} \quad w_2 = \begin{bmatrix} 0 \\ 0 \\ 1/2 \\ 1/2 \end{bmatrix}.$$

We then have

$$Q(x) = (x \cdot v_1)^2 + (x \cdot v_2)^2 - (x \cdot w_1)^2 - (x \cdot w_2)^2$$

and it is easy to see that the v 's and w 's are all orthogonal.

Example of diagonalising a quadratic form

Consider the quadratic form $Q(x) = 4x_1x_2 + 6x_2x_3 + 4x_3x_4$ on \mathbb{R}^4 .

Rewritten symmetrically: $Q(x) = 2x_1x_2 + 2x_2x_1 + 3x_2x_3 + 3x_3x_2 + 2x_3x_4 + 2x_4x_3$.

Corresponding matrix: $B = \begin{bmatrix} 0 & 2 & 0 & 0 \\ 2 & 0 & 3 & 0 \\ 0 & 3 & 0 & 2 \\ 0 & 0 & 2 & 0 \end{bmatrix}$. Characteristic polynomial:

$$\det \begin{bmatrix} -t & 2 & 0 & 0 \\ 2 & -t & 3 & 0 \\ 0 & 3 & -t & 2 \\ 0 & 0 & 2 & -t \end{bmatrix} = -t \det \begin{bmatrix} -t & 3 & 0 \\ 3 & -t & 2 \\ 0 & 2 & -t \end{bmatrix} - 2 \det \begin{bmatrix} 2 & 3 & 0 \\ 0 & -t & 2 \\ 0 & 2 & -t \end{bmatrix}$$

$$\det \begin{bmatrix} -t & 3 & 0 \\ 3 & -t & 2 \\ 0 & 2 & -t \end{bmatrix} = -t(t^2 - 4) - 3(-3t) = 13t - t^3$$

$$\det \begin{bmatrix} 2 & 3 & 0 \\ 0 & -t & 2 \\ 0 & t & -2 \end{bmatrix} = 2(t^2 - 4) - 3(0 - 0) = 2t^2 - 8$$

$$\begin{aligned} \chi_B(t) &= -t(13t - t^3) - 2(2t^2 - 8) = t^4 - 17t^2 + 16 \\ &= (t^2 - 1)(t^2 - 16) = (t - 1)(t + 1)(t - 4)(t + 4). \end{aligned}$$

Example of diagonalising a quadratic form

$$Q(x) = x^T B x \quad B = \begin{bmatrix} 0 & 2 & 0 & 0 \\ 2 & 0 & 3 & 0 \\ 0 & 3 & 0 & 2 \\ 0 & 0 & 2 & 0 \end{bmatrix} \quad \chi_B(t) = (t-1)(t+1)(t-4)(t+4)$$

Eigenvalues: $\lambda_1 = 1, \lambda_2 = 4, \lambda_3 = -1$ and $\lambda_4 = -4$.

Row-reduce $B - \lambda_i I$ to find the eigenvectors:

$$t_1 = \begin{bmatrix} 2 \\ 1 \\ -1 \\ -2 \end{bmatrix} \quad t_2 = \begin{bmatrix} 1 \\ 2 \\ 2 \\ 1 \end{bmatrix} \quad t_3 = \begin{bmatrix} 2 \\ -1 \\ -1 \\ 2 \end{bmatrix} \quad t_4 = \begin{bmatrix} 1 \\ -2 \\ 2 \\ -1 \end{bmatrix}$$

In each case we see that $t_i \cdot t_i = 10$ so the corresponding orthonormal basis consists of the vectors $u_i = t_i / \sqrt{10}$. Following Proposition 23.23:

$$\begin{aligned} v_1 &= \sqrt{|\lambda_1|} u_1 = t_1 / \sqrt{10} = \sqrt{1/10} [2 \ 1 \ -1 \ -2]^T \\ v_2 &= \sqrt{|\lambda_2|} u_2 = \sqrt{4} t_2 / \sqrt{10} = \sqrt{2/5} [1 \ 2 \ 2 \ 1]^T \\ w_1 &= \sqrt{|\lambda_3|} u_3 = t_3 / \sqrt{10} = \sqrt{1/10} [2 \ -1 \ -1 \ 2]^T \\ w_2 &= \sqrt{|\lambda_4|} u_4 = \sqrt{4} t_4 / \sqrt{10} = \sqrt{2/5} [1 \ -2 \ 2 \ -1]^T \end{aligned}$$

Example of diagonalising a quadratic form

$$Q(x) = 4x_1x_2 + 6x_2x_3 + 4x_3x_4$$

$$v_1 = \sqrt{\frac{1}{10}} \begin{bmatrix} 2 \\ 1 \\ -1 \\ -2 \end{bmatrix} \quad v_2 = \sqrt{\frac{2}{5}} \begin{bmatrix} 1 \\ 2 \\ 2 \\ 1 \end{bmatrix} \quad w_1 = \sqrt{\frac{1}{10}} \begin{bmatrix} 2 \\ -1 \\ -1 \\ 2 \end{bmatrix} \quad w_2 = \sqrt{\frac{2}{5}} \begin{bmatrix} 1 \\ -2 \\ 2 \\ -1 \end{bmatrix}$$

Conclusion: $Q(x) = (x \cdot v_1)^2 + (x \cdot v_2)^2 - (x \cdot w_1)^2 - (x \cdot w_2)^2$.