

# Video Lecture F9: Quadratic Forms & Principal Axes Theorem

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## Outline & Objectives

- Quadratic forms generalize taking the inner (dot, scalar) product of a vector with itself. They come up in engineering applications involving optimization and signal processing, utility functions in economics, confidence ellipsoids in statistics, etc.
- Analyze how a symmetric matrix  $A$  defines a *quadratic form*  $Q(\vec{x}) = \vec{x}^T A \vec{x}$ .
- Leverage the *orthogonal diagonalization*,  $A = P D P^T$ , to compute a *change of basis* matrix  $P$  so that  $\vec{x} = P \vec{y}$  transforms  $Q(\vec{x}) = \vec{x}^T A \vec{x}$  to  $Q(\vec{y}) = \vec{y}^T D \vec{y}$ , with no cross terms.

## Definition

A *quadratic form*  $Q$  on  $\mathbb{R}^n$  is a function  $Q : \mathbb{R}^n \rightarrow \mathbb{R}$  of the form  $Q(\vec{x}) = \vec{x}^\top A \vec{x}$ , where  $A \in \mathbb{R}^{n \times n}$  is symmetric and called the *matrix of the quadratic form*.

$$I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad A = \begin{bmatrix} 3 & 0 \\ 0 & -5 \end{bmatrix}, \quad B = \begin{bmatrix} 7 & -3 \\ -3 & 4 \end{bmatrix}, \quad C = \begin{bmatrix} 4 & -3 \\ -5 & 4 \end{bmatrix}.$$

Find  $A$  for  $Q(\vec{x}) = -4x_1^2 + 7x_2^2 - 5x_3^2 - 6x_1x_2 + 3x_2x_3$ .

Any *invertible* matrix  $P \in \mathbb{R}^{n \times n}$  represents a change of basis from  $\mathcal{E}$ -coords to  $\mathcal{B}$ -coords:  $\vec{x} = P\vec{y} \iff \vec{y} = P^{-1}\vec{x} \iff [\vec{y}]_{\mathcal{B}} = [\vec{x}]_{\mathcal{E}}$

**KEY IDEA:** We can change variables to make QF simpler!

$$\vec{x}^\top A \vec{x} = (P\vec{y})^\top A (P\vec{y}) = \vec{y}^\top P^\top A P \vec{y} = \vec{y}^\top (P^\top A P) \vec{y} = \vec{y}^\top (D) \vec{y}$$

## Principal Axes Theorem

Use change of variable to simplify:  $Q(\vec{x}) = 2x_1^2 - 4x_1x_2 + 5x_2^2$ .

### Theorem (Principal Axes Theorem)

Let  $Q$  be a QF on  $\mathbb{R}^n$  corr. to (symm)  $A \in \mathbb{R}^{n \times n}$ . Then we can find an *orthogonal*  $P \in \mathbb{R}^{n \times n}$  such that  $\vec{x} = P\vec{y}$ , transforming  $Q(\vec{x}) = \vec{x}^T A \vec{x}$  to  $Q(\vec{y}) = \vec{y}^T D \vec{y}$ , with no cross term. ( $D$  is diag.)

The columns of  $P$  are called the *principal axes* of the quadratic form  $Q$ .

We'll see soon how this helps us *classify* QF and handle optimization problems.