

Lesson 13

Chapter 7

Symmetric Matrices and Quadratic Forms

► Diagonalization of Symmetric Matrices

► Quadratic Forms

► Constrained Optimization

► The Singular Value Decomposition

► Applications to Image Processing and Statistics

7.1 Diagonalization of Symmetric Matrices

Definition:

Symmetric matrix: $A^T = A \rightarrow$ Symmetric around the main diagonal: $a_{ij} = a_{ji}$

Ex 1

$$\begin{bmatrix} 1 & 0 \\ 0 & -3 \end{bmatrix}$$

$$\begin{bmatrix} 0 & -1 & 0 \\ -1 & 5 & 8 \\ 0 & 8 & -7 \end{bmatrix}$$

$$\begin{bmatrix} a & b & c \\ b & d & e \\ c & e & f \end{bmatrix}$$

← Symmetric

$$\begin{bmatrix} 1 & -3 \\ 3 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & -4 & 0 \\ -6 & 1 & -4 \\ 0 & -6 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 5 & 4 & 3 & 2 \\ 4 & 3 & 2 & 1 \\ 3 & 2 & 1 & 0 \end{bmatrix}$$

← Nonsymmetric

En matrix A er symmetrisk, hvis $A = A^T$

Lad nu en matrix A være defineret ved det *ydre produkt* af en vektor a : $A = \mathbf{a}\mathbf{a}^T$

Spørgsmål: Vil A (altid) være symmetrisk?

Beregning:

Definer fire vektorer \mathbf{a} af dimension 4×1 med tilfældige talværdier, beregn A som ovenfor og kvalificer svaret.

Theorem 5.5, The Diagonalization Theorem

An $n \times n$ matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.

In fact, $A = PDP^{-1}$, with D a diagonal matrix, if and only if the columns of P are n linearly independent eigenvectors of A . In this case, the diagonal entries of D are eigenvalues of A that correspond, respectively, to the eigenvectors of A .

Step 1: Eigenvalues: $\det(A - \lambda_i I) = 0$ (Characteristic Equation)

Step 2: Eigenvectors: $(A - \lambda_i I)\mathbf{v}_i = \mathbf{0}; \quad \mathbf{v}_i \neq \mathbf{0}$

Step 3: Diagonalization matrix: $P = [\mathbf{v}_1 \ \mathbf{v}_2 \ \dots \ \mathbf{v}_n]$

Step 4: Diagonal matrix: $D = \begin{bmatrix} \lambda_1 & \cdots & 0 \\ 0 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \lambda_n \end{bmatrix}$

Ex 2 $A = \begin{bmatrix} 6 & -2 & -1 \\ -2 & 6 & -1 \\ -1 & -1 & 5 \end{bmatrix} = PDP^{-1} \quad ???$

Step 1: Eigenvalues: $\det(A - \lambda I) = -\lambda^3 + 17\lambda^2 - 90\lambda + 144 = -(\lambda - 8)(\lambda - 6)(\lambda - 3) = 0$
 $\Rightarrow \lambda_1 = 8; \lambda_2 = 6; \lambda_3 = 3$

Step 2: Eigenvectors: $(A - \lambda_i I)\mathbf{v}_i = \mathbf{0} \Rightarrow \lambda_1 = 8: \mathbf{v}_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}; \quad \lambda_2 = 6: \mathbf{v}_2 = \begin{bmatrix} -1 \\ -1 \\ 2 \end{bmatrix}; \quad \lambda_3 = 3: \mathbf{v}_3 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

$\mathbf{v}_1 \cdot \mathbf{v}_2 = \mathbf{v}_1 \cdot \mathbf{v}_3 = \mathbf{v}_2 \cdot \mathbf{v}_3 = 0 \rightarrow$ **Orthogonal basis** for \mathbb{R}^3

$\mathbf{u}_1 = \begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix}; \quad \mathbf{u}_2 = \begin{bmatrix} -1/\sqrt{6} \\ -1/\sqrt{6} \\ 2/\sqrt{6} \end{bmatrix}; \quad \mathbf{u}_3 = \begin{bmatrix} 1/\sqrt{3} \\ 1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix} \rightarrow$ **Orthonormal basis of eigenvectors**

Ex 2

Step 3: Diagonalization matrix:

$$P = [\mathbf{u}_1 \ \mathbf{u}_2 \ \mathbf{u}_3] = \begin{bmatrix} -1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{6} & 1/\sqrt{3} \\ 0 & 2/\sqrt{6} & 1/\sqrt{3} \end{bmatrix} \rightarrow \text{Orthogonal matrix} \rightarrow P^{-1} = P^T = \begin{bmatrix} -1/\sqrt{2} & 1/\sqrt{2} & 0 \\ -1/\sqrt{6} & -1/\sqrt{6} & 2/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \end{bmatrix}$$

Theorem 6.6
↓

Step 4: Diagonal matrix: $D = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} = \begin{bmatrix} 8 & 0 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 3 \end{bmatrix}$

$$A = \begin{bmatrix} 6 & -2 & -1 \\ -2 & 6 & -1 \\ -1 & -1 & 5 \end{bmatrix} = PDP^{-1} = PDP^T \rightarrow \text{Orthogonally diagonalizable} \quad \checkmark$$

Ex 3 $A = \begin{bmatrix} 3 & -2 & 4 \\ -2 & 6 & 2 \\ 4 & 2 & 3 \end{bmatrix} = PDP^{-1} \quad ???$

Step 1: Eigenvalues: $\det(A - \lambda I) = -\lambda^3 + 12\lambda^2 - 21\lambda - 98 = -(\lambda - 7)^2(\lambda + 2) = 0$
 $\Rightarrow \lambda_1 = \lambda_2 = 7; \quad \lambda_3 = -2$

*Two-dimensional
eigenspace*

Step 2: Eigenvectors: $(A - \lambda_i I)\mathbf{v}_i = \mathbf{0} \Rightarrow \lambda_1 = \lambda_2 = 7: \mathbf{v}_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}; \quad \mathbf{v}_2 = \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix}; \quad \lambda_3 = -2: \mathbf{v}_3 = \begin{bmatrix} -2 \\ -1 \\ 2 \end{bmatrix}$

$\mathbf{v}_1 \cdot \mathbf{v}_3 = \mathbf{v}_2 \cdot \mathbf{v}_3 = 0; \quad \mathbf{v}_1 \cdot \mathbf{v}_2 = -1 \rightarrow \text{Gram-Schmidt: } \mathbf{z}_2 = \mathbf{v}_2 - \frac{\mathbf{v}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 = \begin{bmatrix} -1/2 \\ 2 \\ 1/2 \end{bmatrix}$

$\mathbf{u}_1 = \begin{bmatrix} 1/\sqrt{2} \\ 0 \\ 1/\sqrt{2} \end{bmatrix}; \quad \mathbf{u}_2 = \begin{bmatrix} -1/\sqrt{18} \\ 4/\sqrt{18} \\ 1/\sqrt{18} \end{bmatrix}; \quad \mathbf{u}_3 = \begin{bmatrix} -2/3 \\ -1/3 \\ 2/3 \end{bmatrix} \rightarrow \text{Orthonormal basis of eigenvectors}$

Ex 3

Step 3: Diagonalization matrix:

$$P = [\mathbf{u}_1 \ \mathbf{u}_2 \ \mathbf{u}_3] = \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{18} & -2/3 \\ 0 & 4/\sqrt{18} & -1/3 \\ 1/\sqrt{2} & 1/\sqrt{18} & 2/3 \end{bmatrix} \rightarrow \text{Orthogonal matrix} \rightarrow P^{-1} = P^T = \begin{bmatrix} 1/\sqrt{2} & 0 & 1/\sqrt{2} \\ -1/\sqrt{18} & 4/\sqrt{18} & 1/\sqrt{18} \\ -2/3 & -1/3 & 2/3 \end{bmatrix}$$

Theorem 6.6
↓

Step 4: Diagonal matrix: $D = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} = \begin{bmatrix} 7 & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & -2 \end{bmatrix}$

$$A = \begin{bmatrix} 3 & -2 & 4 \\ -2 & 6 & 2 \\ 4 & 2 & 3 \end{bmatrix} = PDP^{-1} = PDP^T \rightarrow \text{Orthogonally diagonalizable} \quad \checkmark$$

Definition:

Orthogonal matrix

Diagonal matrix

A is orthogonally diagonalizable: $A = PDP^{-1}$

Theorem 7.1:

A is symmetric \Rightarrow any two eigenvectors from different eigenspaces are orthogonal

Theorem 7.2:

An $n \times n$ matrix A is orthogonal diagonalizable $\Leftrightarrow A$ is symmetric

Numerical note: Using orthogonal matrices generally reduces numerical errors in calculations

Theorem 7.3: The Spectral Theorem for Symmetric Matrices

A $n \times n$ symmetric matrix A has the following properties

- ▶ A has n real eigenvalues, counting multiplicities.
- ▶ The dimension of the eigenspace for each eigenvalue λ equals the multiplicity of λ as a root of the characteristic equation.
- ▶ The eigenspaces are mutually orthogonal, in the sense that eigenvectors corresponding to different eigenvalues are orthogonal.
- ▶ A is orthogonally diagonalizable.

Spectral decomposition:

Spectrum of A

Let: A a symmetric $n \times n$ matrix with eigenvalues: $\lambda_1, \dots, \lambda_n$ and corresponding orthonormal eigenvectors: $\mathbf{u}_1, \dots, \mathbf{u}_n$.

Then:

$$A = PDP^{-1} = PDP^T = [\mathbf{u}_1 \dots \mathbf{u}_n] \begin{bmatrix} \lambda_1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix} \begin{bmatrix} \mathbf{u}_1^T \\ \vdots \\ \mathbf{u}_n^T \end{bmatrix} = [\lambda_1 \mathbf{u}_1 \dots \lambda_n \mathbf{u}_n] \begin{bmatrix} \mathbf{u}_1^T \\ \vdots \\ \mathbf{u}_n^T \end{bmatrix}$$

$$= \lambda_1 \mathbf{u}_1 \mathbf{u}_1^T + \dots + \lambda_i \mathbf{u}_i \mathbf{u}_i^T + \dots + \lambda_n \mathbf{u}_n \mathbf{u}_n^T \quad \leftarrow \text{Spectral decomposition of } A$$

$n \times n$ matrix
with rank 1

Projection matrix on
subspace spanned by \mathbf{u}_i :
 $(\mathbf{u}_i \mathbf{u}_i^T) \mathbf{x} = \text{proj}_{\mathbf{u}_i} \mathbf{x}$

Også kaldet:
egenværdi-dekomposition

Ex 4

$$A = \begin{bmatrix} 7 & 2 \\ 2 & 4 \end{bmatrix} = \begin{bmatrix} 2/\sqrt{5} & -1/\sqrt{5} \\ 1/\sqrt{5} & 2/\sqrt{5} \end{bmatrix} \begin{bmatrix} 8 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 2/\sqrt{5} & 1/\sqrt{5} \\ -1/\sqrt{5} & 2/\sqrt{5} \end{bmatrix}$$

$$\mathbf{u}_1 = \begin{bmatrix} 2/\sqrt{5} \\ 1/\sqrt{5} \end{bmatrix} \Rightarrow \mathbf{u}_1 \mathbf{u}_1^T = \begin{bmatrix} 4/5 & 2/5 \\ 2/5 & 1/5 \end{bmatrix} \quad \mathbf{u}_2 = \begin{bmatrix} -1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix} \Rightarrow \mathbf{u}_2 \mathbf{u}_2^T = \begin{bmatrix} 1/5 & -2/5 \\ -2/5 & 4/5 \end{bmatrix}$$

Spectral decomposition:

$$A = 8\mathbf{u}_1 \mathbf{u}_1^T + 3\mathbf{u}_2 \mathbf{u}_2^T = \begin{bmatrix} 32/5 & 16/5 \\ 16/5 & 8/5 \end{bmatrix} + \begin{bmatrix} 3/5 & -6/5 \\ -6/5 & 12/5 \end{bmatrix} = \begin{bmatrix} 35/5 & 10/5 \\ 10/5 & 20/5 \end{bmatrix} = \begin{bmatrix} 7 & 2 \\ 2 & 4 \end{bmatrix} \checkmark$$

OPGAVE 1

For den symmetriske matrix A:

$$A = \begin{bmatrix} 1 & 2 & -5 \\ 2 & -3 & 0 \\ -5 & 0 & -1 \end{bmatrix}$$

- Bestem egenverdierne og de tilhørende *ortonormale* egenvektorer
- Udregn de tre matricer i den spektrale dekomposition og opskriv dekompositionen
- Gør prøve: Summerer de tre matricer til A?
- Er dekompositionen PDP^{-1} i øvrigt lig med PDP^T ??

7.2 Quadratic Forms

Definition:

Q is a **quadratic form** on \mathbb{R}^n if:

$$Q(\boldsymbol{x}) = \boldsymbol{x}^T A \boldsymbol{x}$$

where \boldsymbol{x} is a vector in \mathbb{R}^n and A – the **matrix of the quadratic form** – is an $n \times n$ symmetric matrix .

Theorem 7.2

A symmetric $\Rightarrow \exists$ orthogonal matrix P : $P^T A P = D$, where D is a diagonal matrix.

Change of variable:

See Chap. 4.4

$\mathbf{x} = P\mathbf{y}$ (or $\mathbf{y} = P^{-1}\mathbf{x}$) \rightarrow \mathbf{y} is the coordinate vector of \mathbf{x} in the orthonormal basis of \mathbb{R}^n determined by the columns of P .

”Hovedakse”

The Principal Axes

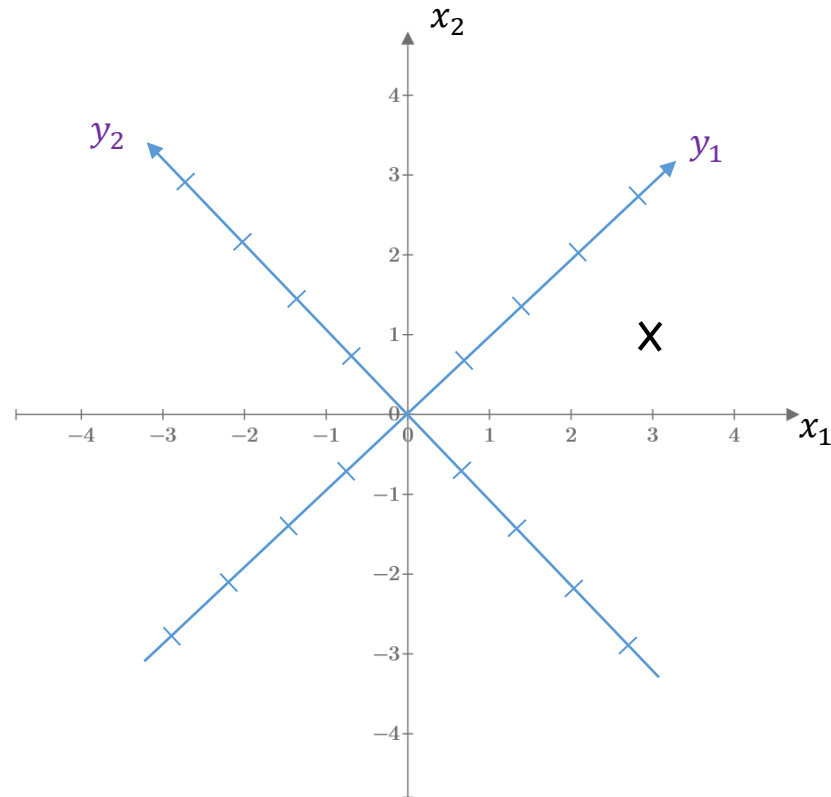
Theorem 7.4: The Principal Axes Theorem:

$$Q(\mathbf{x}) = \mathbf{x}^T A \mathbf{x} = (P\mathbf{y})^T A (P\mathbf{y}) = \mathbf{y}^T (P^T A P) \mathbf{y} = \mathbf{y}^T D \mathbf{y} = \lambda_1 y_1^2 + \cdots + \lambda_n y_n^2$$

OBS: No cross-product terms!

Ex 6 $Q(\mathbf{x}) = 5x_1^2 - 4x_1x_2 + 5x_2^2$

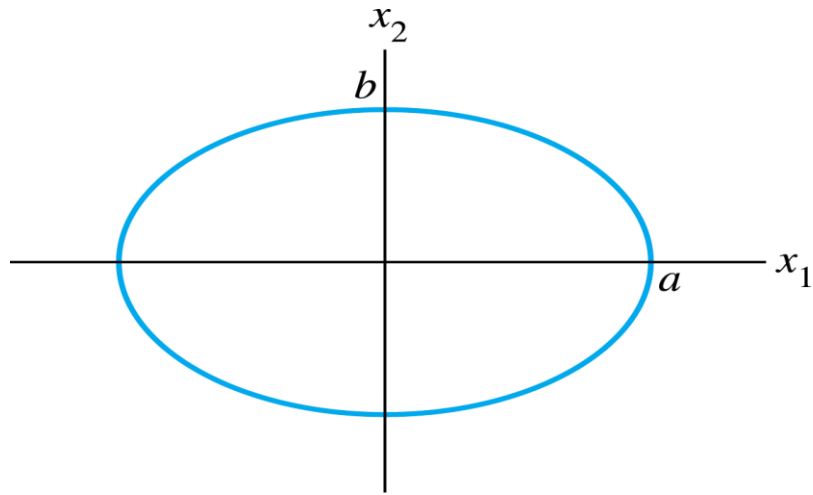
$$\rightarrow P = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}; D = \begin{bmatrix} 3 & 0 \\ 0 & 7 \end{bmatrix}; P^{-1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$



Geometric interpretation of Principal Axes

$\mathbf{x}^T A \mathbf{x} = 1$; $\mathbf{x} \in \mathbb{R}^2$, A diagonal 2×2 matrix

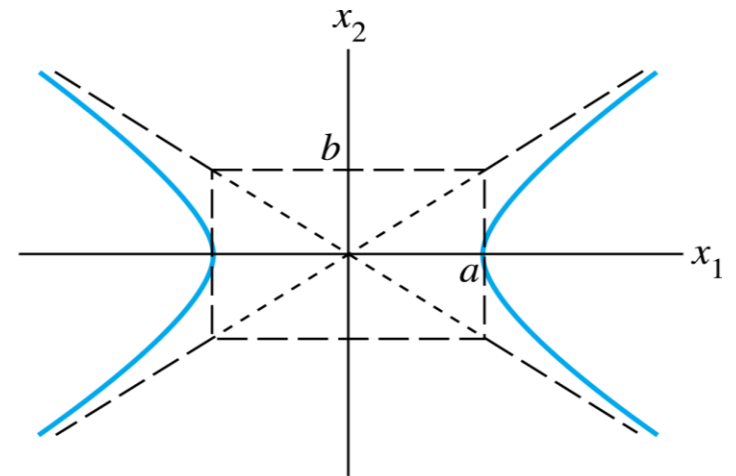
$$A = \begin{bmatrix} 1/a^2 & 0 \\ 0 & 1/b^2 \end{bmatrix}$$



$$\frac{x_1^2}{a^2} + \frac{x_2^2}{b^2} = 1, \quad a > b > 0$$

ellipse

$$A = \begin{bmatrix} 1/a^2 & 0 \\ 0 & -1/b^2 \end{bmatrix}$$



$$\frac{x_1^2}{a^2} - \frac{x_2^2}{b^2} = 1, \quad a > b > 0$$

hyperbola

Geometric interpretation of Principal Axes (Ex 6)

$\mathbf{x}^T A \mathbf{x} = c$; $\mathbf{x} \in \mathbb{R}^2$, $c \in \mathbb{R}$, A symmetric 2×2 matrix

$$A = \begin{bmatrix} a & d \\ d & b \end{bmatrix} = \begin{bmatrix} 5 & -2 \\ -2 & 5 \end{bmatrix}; \quad c = 38$$

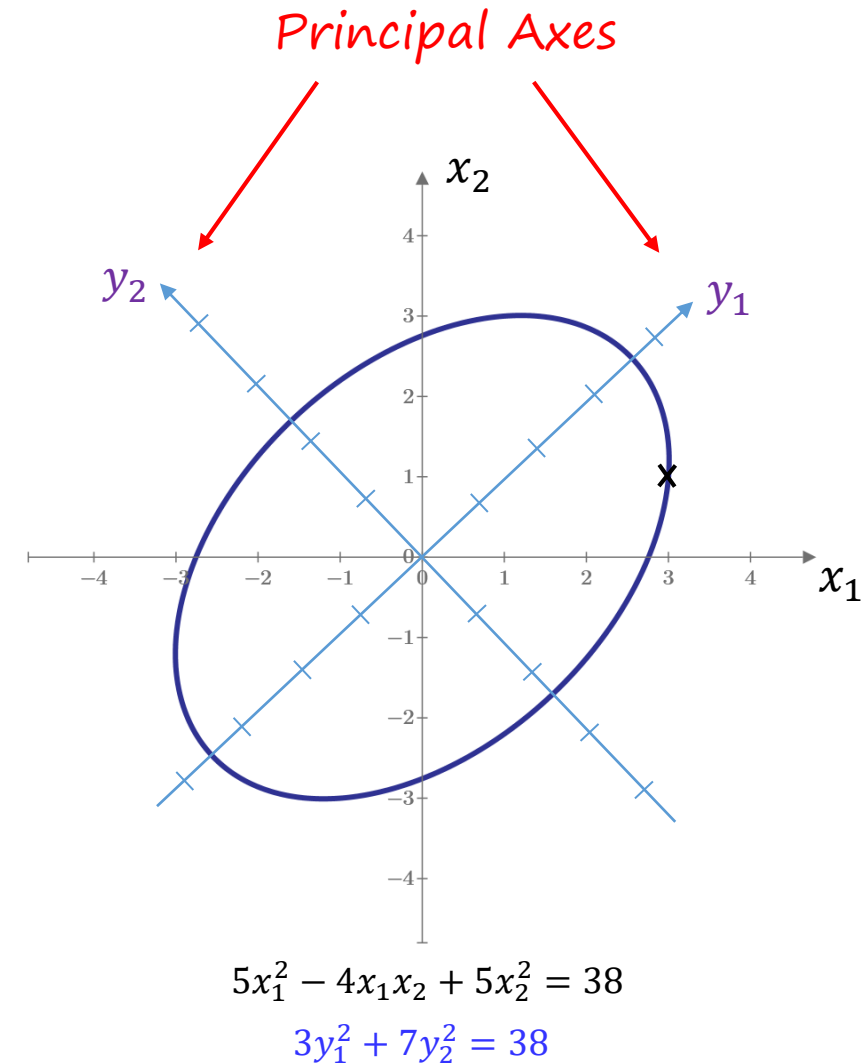
$$\mathbf{x}^T A \mathbf{x} = c \Leftrightarrow 5x_1^2 - 4x_1x_2 + 5x_2^2 = 38$$

$$\lambda_1 = 3: \mathbf{u}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \lambda_2 = 7: \mathbf{u}_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$P = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \quad D = \begin{bmatrix} 3 & 0 \\ 0 & 7 \end{bmatrix}$$

$$\mathbf{y} = P^{-1} \mathbf{x} = \frac{1}{\sqrt{2}} \begin{bmatrix} x_1 + x_2 \\ -x_1 + x_2 \end{bmatrix}$$

$$\mathbf{x}^T A \mathbf{x} = \mathbf{y}^T D \mathbf{y} = c \Leftrightarrow 3y_1^2 + 7y_2^2 = 38 \Leftrightarrow \frac{y_1^2}{(\sqrt{38/3})^2} + \frac{y_2^2}{(\sqrt{38/7})^2} = 1 \quad \leftarrow \text{Ellipse}$$



Geometric interpretation of Principal Axes

$$\mathbf{x}^T A \mathbf{x} = c; \quad \mathbf{x} \in \mathbb{R}^2, \quad c \in \mathbb{R}, \quad A \text{ symmetric } 2 \times 2 \text{ matrix}$$

$$A = \begin{bmatrix} a & d \\ d & -b \end{bmatrix} = \begin{bmatrix} 1 & -4 \\ -4 & -5 \end{bmatrix}; \quad c = 16$$

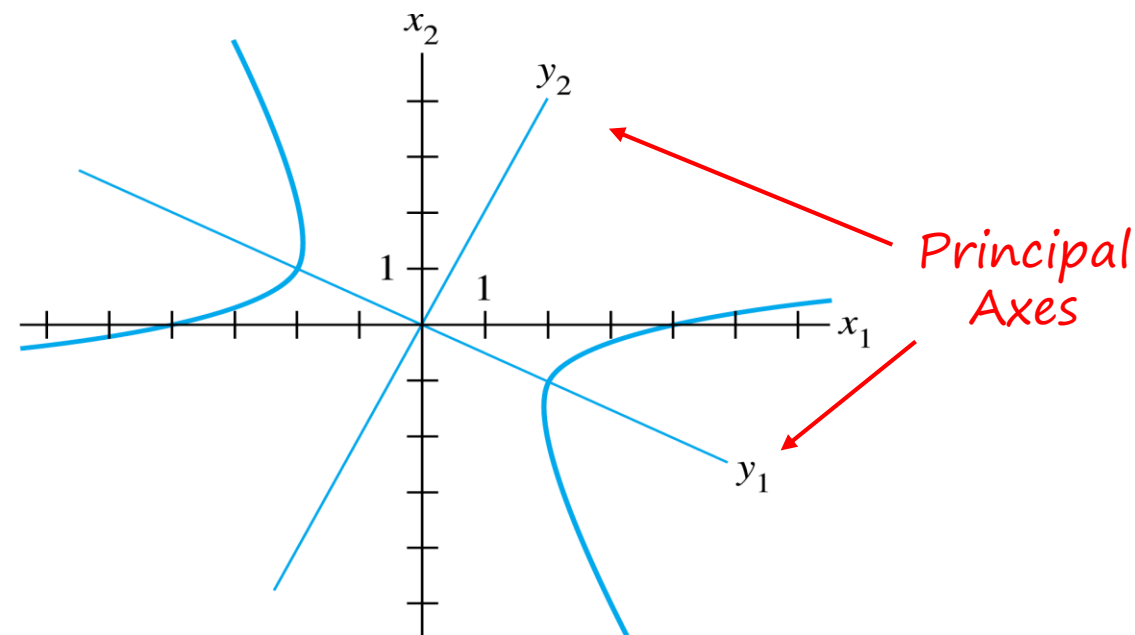
$$\mathbf{x}^T A \mathbf{x} = c \Leftrightarrow x_1^2 - 8x_1x_2 - 5x_2^2 = 16$$

$$\lambda_1 = 3: \mathbf{u}_1 = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ -1 \end{bmatrix} \quad \lambda_2 = -7: \mathbf{u}_2 = \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$P = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & 1 \\ -1 & 2 \end{bmatrix} \quad D = \begin{bmatrix} 3 & 0 \\ 0 & -7 \end{bmatrix}$$

$$\mathbf{y} = P^{-1} \mathbf{x} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2x_1 - x_2 \\ x_1 + 2x_2 \end{bmatrix}$$

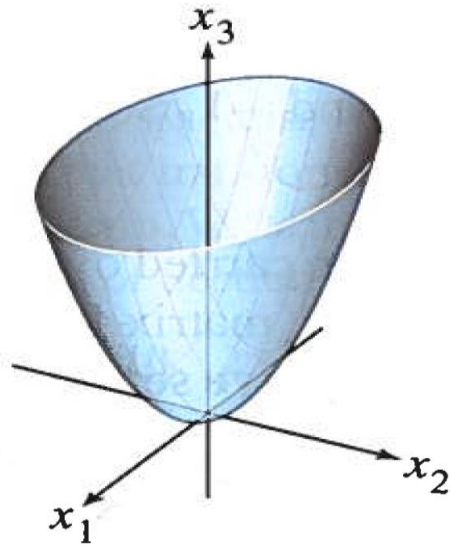
$$\mathbf{x}^T A \mathbf{x} = \mathbf{y}^T D \mathbf{y} = c \Leftrightarrow 3y_1^2 - 7y_2^2 = 16 \Leftrightarrow \frac{y_1^2}{(4/\sqrt{3})^2} - \frac{y_2^2}{(4/\sqrt{7})^2} = 1$$



$$(b) \quad x_1^2 - 8x_1x_2 - 5x_2^2 = 16$$

$$3y_1^2 - 7y_2^2 = 16$$

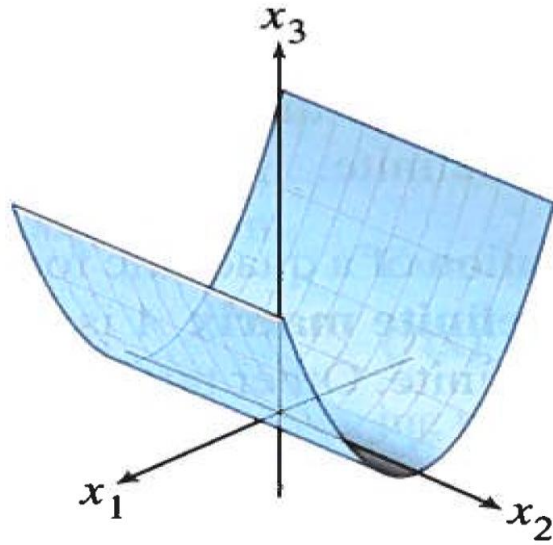
Classifying Quadratic Forms



(a) $z = 3x_1^2 + 7x_2^2$



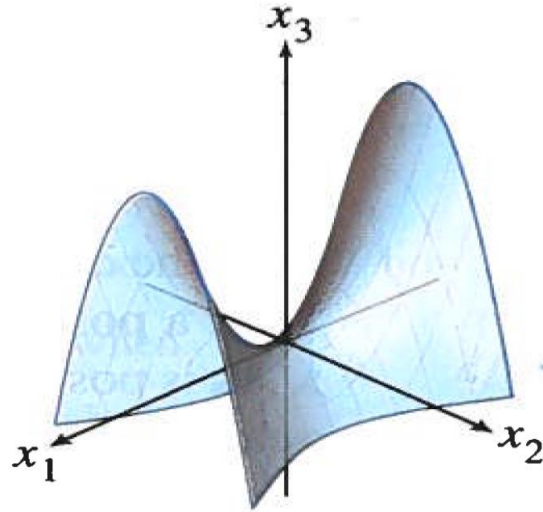
Positive definite:
 $Q(x) > 0$ for all $x \neq 0$



(b) $z = 3x_1^2$



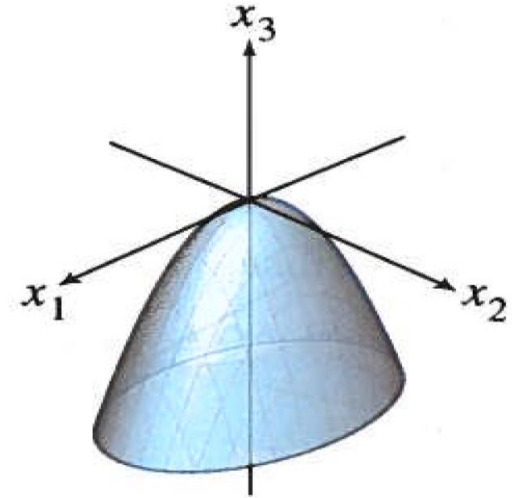
Positive semidefinite:
 $Q(x) \geq 0$ for all x



(c) $z = 3x_1^2 - 7x_2^2$



Indefinite:
 $Q(x)$ both positive
and negative




(d) $z = -3x_1^2 - 7x_2^2$



Negative definite:
 $Q(x) < 0$ for all $x \neq 0$

Theorem 7.5, Quadratic forms and eigenvalues

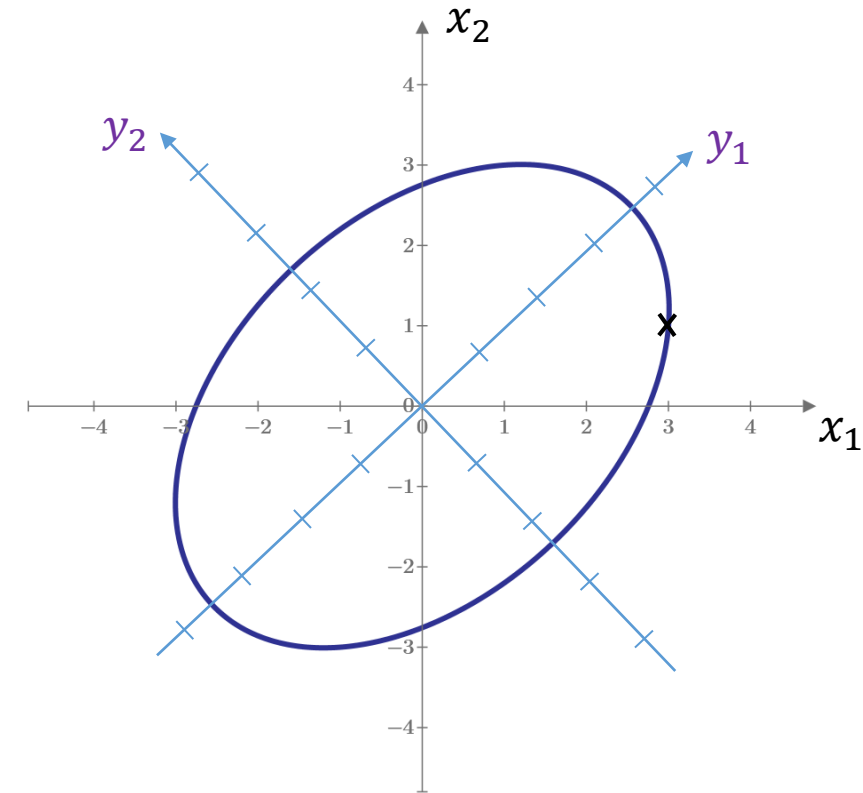
Let A be an $n \times n$ symmetric matrix. Then a quadratic form $\mathbf{x}^T A \mathbf{x}$ is:

-  **a:** Positive definite if and only if the eigenvalues of A are all positive.
- b:** Negative definite if and only if the eigenvalues of A are all negative.
- c:** Indefinite if and only if A has both positive and negative eigenvalues.

Note: some times the term *nonnegative definite* is also used for *positive semidefinite*

Diskussion:

I hvilke tilfælde af beskrivelse af datasæt kunne det være smart at lave koordinattransformation og dermed danne nye hovedakser i koordinatsystemet?



7.3 Constrained Optimization

Constrained optimization

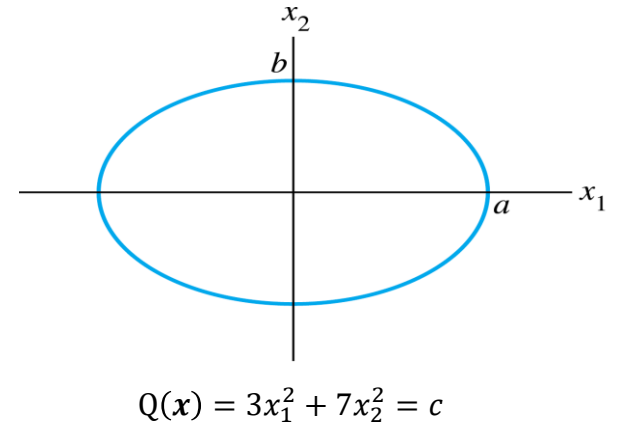
Finding max/min of a quadratic form $Q(\mathbf{x})$ for \mathbf{x} in some specific (constrained) set. Typically with constraint:

$$\|\mathbf{x}\|^2 = \|\mathbf{x}\| = \mathbf{x}^T \mathbf{x} = x_1^2 + \cdots + x_n^2 = 1$$

Ex 7

$$Q(\mathbf{x}) = 3x_1^2 + 7x_2^2 = \mathbf{x}^T A \mathbf{x}; \quad A = \begin{bmatrix} 3 & 0 \\ 0 & 7 \end{bmatrix} \Rightarrow \begin{cases} \lambda_1 = 7: \mathbf{u}_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \\ \lambda_2 = 3: \mathbf{u}_2 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \end{cases}$$

$$\|\mathbf{x}\|^2 = \mathbf{x}^T \mathbf{x} = x_1^2 + x_2^2 = 1$$



$$\left. \begin{aligned} Q(\mathbf{x}) = 3x_1^2 + 7x_2^2 &\leq 7x_1^2 + 7x_2^2 = 7\|\mathbf{x}\|^2 = 7 \\ Q(\pm \mathbf{u}_1) &= Q(0, \pm 1) = 7 \end{aligned} \right\} \Rightarrow Q_{\max} = 7 = \lambda_{\max}$$

$$\left. \begin{aligned} Q(\mathbf{x}) = 3x_1^2 + 7x_2^2 &\geq 3x_1^2 + 3x_2^2 = 3\|\mathbf{x}\|^2 = 3 \\ Q(\pm \mathbf{u}_2) &= Q(\pm 1, 0) = 3 \end{aligned} \right\} \Rightarrow Q_{\min} = 3 = \lambda_{\min}$$

$$\|\mathbf{x}\|^2 = \mathbf{x}^T \mathbf{x} = 1 \Rightarrow 3 \leq Q(\mathbf{x}) \leq 7 \text{ (continuous)}$$

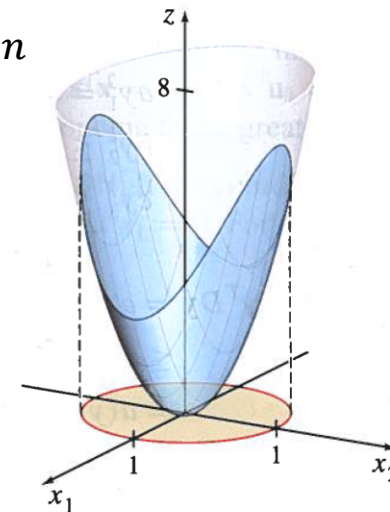


FIGURE 1 $z = 3x_1^2 + 7x_2^2$.

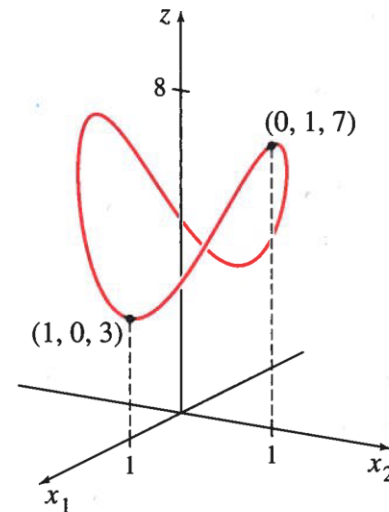


FIGURE 2 The intersection of $z = 3x_1^2 + 7x_2^2$ and the cylinder $x_1^2 + x_2^2 = 1$.

Ex 8

Symmetric matrix

Eigenvalues/-vectors

$$A = \begin{bmatrix} 3 & 2 & 1 \\ 2 & 3 & 1 \\ 1 & 1 & 4 \end{bmatrix} \rightarrow \lambda_1 = 6: \mathbf{u}_1 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}; \quad \lambda_2 = 3: \mathbf{u}_2 = \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ 1 \\ -2 \end{bmatrix}; \quad \lambda_3 = 1: \mathbf{u}_3 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

Change of coordinates: $\mathbf{x} = P\mathbf{y} \Rightarrow \|\mathbf{x}\|^2 = \|P\mathbf{y}\|^2 = (P\mathbf{y})^T(P\mathbf{y}) = \mathbf{y}^T P^T P \mathbf{y} = \mathbf{y}^T I \mathbf{y} = \mathbf{y}^T \mathbf{y} = \|\mathbf{y}\|^2$

Constrain: $\|\mathbf{x}\|^2 = 1 \Leftrightarrow \|\mathbf{y}\|^2 = 1$

Diagonalization: $\mathbf{x} = P\mathbf{y} \Rightarrow \mathbf{x}^T A \mathbf{x} = \mathbf{y}^T D \mathbf{y}$, where $D = \begin{bmatrix} 6 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{bmatrix} \Rightarrow 1 \leq \mathbf{x}^T A \mathbf{x} = \mathbf{y}^T D \mathbf{y} \leq 6$

Constrained limits: $\Rightarrow 1 \leq Q(\mathbf{x}) = \mathbf{x}^T A \mathbf{x} = 3x_1^2 + 3x_2^2 + 4x_3^2 + 4x_1x_2 + 2x_1x_3 + 2x_2x_3 \leq 6$

Maximum value/vector: $\Rightarrow Q(\mathbf{u}_1) = Q\left(\frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}, \frac{1}{\sqrt{3}}\right) = 6$

Minimum value/vector: $\Rightarrow Q(\mathbf{u}_3) = Q\left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 0\right) = 1$

Theorem 7.6

Let A be a symmetric matrix and let m and M be defined as

$$m = \min\{\mathbf{x}^T A \mathbf{x} : \|\mathbf{x}\| = 1\}, \quad M = \max\{\mathbf{x}^T A \mathbf{x} : \|\mathbf{x}\| = 1\},$$

Then M is the greatest eigenvalue λ_1 of A and m is the least eigenvalue of A . The value of $\mathbf{x}^T A \mathbf{x}$ is M when \mathbf{x} is a unit eigenvector \mathbf{u}_1 corresponding to λ_1 . The value of $\mathbf{x}^T A \mathbf{x}$ is m when \mathbf{x} is a unit eigenvector corresponding to m .

For $\|\mathbf{x}\| = 1$: $Q(\mathbf{x}) = \mathbf{x}^T A \mathbf{x}$ will take all values between m and M .

Today's words and concepts

Symmetric matrix

Orthogonal diagonalizable

Matrix of quadratic form

Indefinite

Negative definite

Constrained optimization

The Spectral Theorem

The Principal Axes Theorem

Positive definite

Positive semidefinite

Principal Axes

Spectral decomposition

Quadratic forms

Negative semidefinite