

Exercise 3 Let $\varphi : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ be a rotation around the axis $x_1 = x_2, x_3 = 0$ by the angle $\frac{\pi}{2}$ in such a way, that $\varphi(1,0,0)$ has all three components positive. Find a matrix A such that in standard coordinates

$$\varphi(x) = Ax.$$

Solution: ① We will find the matrix $(\varphi)_{\alpha, \alpha}$ in a suitable orthonormal basis and then we can compute

$$A = (\varphi)_{E,E} = (\text{id})_{E,\alpha} (\varphi)_{\alpha,\alpha} (\text{id})_{\alpha,E} = P (\varphi)_{\alpha,\alpha} P^{-1}$$

where E is the standard basis $e_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, e_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, e_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$.

Basis α The first vector is the generator of the axis, i.e.

$$v_1 = \left(\frac{1}{\sqrt{2}}, \frac{1}{\sqrt{2}}, 0 \right)$$

Other two vectors are perpendicular to it and also mutually orthogonal.

$$v_2 = \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 0 \right) \quad v_3 = (0, 0, 1)$$

$\alpha = (v_1, v_2, v_3)$ is an orthonormal basis and $\varphi(v_1) = v_1, \varphi(v_3) = -v_2$ (since $\varphi(1,0,0) = (x,y,z)$ gives $x > 0, y > 0, z > 0$ gives the direction)

$\varphi(v_2) = v_3$. Hence

$$(\varphi)_{\alpha, \alpha} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \frac{\pi}{2} & -\sin \frac{\pi}{2} \\ 0 & \sin \frac{\pi}{2} & \cos \frac{\pi}{2} \end{pmatrix}$$

$$P = (\text{id})_{\epsilon, \alpha} = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

is an orthogonal matrix, since both bases

ϵ and α are orthonormal. Hence

$$(\text{id})_{\alpha, \epsilon} = (\text{id})_{\epsilon, \alpha}^{-1} = P^{-1} = P^T = \begin{pmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

Now

$$A = P(\varphi)_{\alpha, \alpha} P^{-1} = P(\varphi)_{\alpha, \alpha} P^T = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & -\frac{\sqrt{2}}{2} \\ \frac{1}{2} & \frac{1}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} & 0 \end{pmatrix}.$$

② Another solution uses the fact that the columns of the matrix A are Ae_1, Ae_2, Ae_3 . So we compute $\varphi(e_1), \varphi(e_2), \varphi(e_3)$ using geometric considerations.

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SELFADJOINT OPERATORS

Let U and V be spaces with scalar product over \mathbb{R} or \mathbb{C} . Let $\varphi: U \rightarrow V$ be a linear mapping. A linear mapping

$$\varphi^*: V \rightarrow U$$

is called adjoint to φ , if for all $u \in U$ and $v \in V$

$$\langle \varphi(u), v \rangle_V = \langle u, \varphi^*(v) \rangle_U.$$

Example 1 $U = \mathbb{R}^n$, $V = \mathbb{R}^k$, $\varphi(x) = Ax$, where A is a matrix $k \times n$. Then $\varphi^*: \mathbb{R}^k \rightarrow \mathbb{R}^n$ is the mapping $\varphi^*(y) = A^T y$, since for $x \in \mathbb{R}^n$ and $y \in \mathbb{R}^k$ we have

$$\langle \varphi(x), y \rangle = \langle Ax, y \rangle = (Ax)^T \cdot y = x^T A^T y$$

$$\langle x, \varphi^*(y) \rangle = \langle x, A^T y \rangle = x^T (A^T y) = x^T A^T y.$$

Example 2 $U = \mathbb{C}^n$, $V = \mathbb{C}^k$, $\varphi(x) = Ax$, $\varphi^*(y) = \bar{A}^T y$, where $\bar{}$ stands for complex conjugation. The proof is very similar to the one above.

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Theorem Let α be an orthonormal basis in U , β an orthonormal basis in V , $\varphi: U \rightarrow V$ and $\varphi^*: V \rightarrow U$. Then

$$(\varphi^*)_{\alpha, \beta} = \overline{(\varphi)_{\beta, \alpha}}^T.$$

Proof: Let $(\varphi)_{\beta, \alpha} = A$, $(\varphi^*)_{\alpha, \beta} = B$. Then for $u \in U$ and $v \in V$ we get

$$\langle \varphi(u), v \rangle = (\varphi(u))_{\beta}^T \cdot \overline{(v)}_{\beta} = (A(u)_{\alpha})^T \overline{(v)}_{\beta} = (u_{\alpha})_{\alpha}^T A^T \overline{(v)}_{\beta}$$

$$\langle u, \varphi^*(v) \rangle = (u)_{\alpha}^T \overline{(\varphi^*(v))}_{\alpha} = (u)_{\alpha}^T \cdot \overline{B(v)}_{\beta} = (u)_{\alpha}^T \overline{B} \overline{(v)}_{\beta}$$

Hence $A^T = \overline{B}$, and consequently

$$B = \overline{A}^T.$$

Corollary To every linear mapping $\varphi: U \rightarrow V$ there is an adjoint mapping φ^* .

Definition of selfadjoint mapping

Let U be a vector space with a scalar product. An operator $\varphi: U \rightarrow U$ is called selfadjoint if $\varphi = \varphi^*$, i.e. for all $u, v \in U$

$$\langle \varphi(u), v \rangle = \langle u, \varphi(v) \rangle.$$

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Example 1 $U = \mathbb{R}^n$, $A = A^T$ is a symmetric matrix. Then $\varphi: \mathbb{R}^n \rightarrow \mathbb{R}^n$, $\varphi(x) = Ax$ is a selfadjoint operator.

Example 2 $U = \mathbb{C}^n$, $A = \bar{A}^T$ is a hermitian matrix. Then $\varphi: \mathbb{C}^n \rightarrow \mathbb{C}^n$, $\varphi(x) = Ax$ is a selfadjoint operator.

Hermitian matrix $A = \begin{pmatrix} 3 & i & 2-i \\ -i & 4 & 3+4i \\ 2+i & 3-4i & -8 \end{pmatrix}$

Example 3 $\varphi: U \rightarrow U$ is an orthogonal projection onto the subspace $V \subseteq U$.

The φ is a selfadjoint operator.

Let $u = u_1 + u_2$, $v = v_1 + v_2$, $u_1, v_1 \in V$ and $u_2, v_2 \in V^\perp$. Then

$$\begin{aligned} \langle \varphi(u), v \rangle &= \langle u_1, v \rangle = \langle u_1, v_1 + v_2 \rangle = \langle u_1, v_1 \rangle \\ \langle u, \varphi(v) \rangle &= \langle u, v_1 \rangle = \langle u_1 + u_2, v_1 \rangle = \langle u_1, v_1 \rangle. \end{aligned}$$

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Eigenvalues and eigenvectors of selfadjoint operators

- ① Every eigenvalue of a selfadjoint operator is real (even if we work over \mathbb{C}).
- ② Eigenvectors to different eigenvalues are perpendicular each to another.

Proof ① Let $\varphi(u) = \lambda u$, $u \neq \vec{0}$. Then

$$\begin{aligned} \lambda \langle u, u \rangle &= \langle \lambda u, u \rangle = \langle \varphi(u), u \rangle = \langle u, \varphi(u) \rangle = \langle u, \lambda u \rangle \\ &= \bar{\lambda} \langle u, u \rangle \end{aligned}$$

which gives $\lambda = \bar{\lambda}$, and so λ is real.

② Let $\varphi(u_1) = \lambda_1 u_1$, $\varphi(u_2) = \lambda_2 u_2$, $\lambda_1 \neq \lambda_2$, $u_1 \neq \vec{0}$, $u_2 \neq \vec{0}$. Then

$$\begin{aligned} \lambda_1 \langle u_1, u_2 \rangle &= \langle \lambda_1 u_1, u_2 \rangle = \langle \varphi(u_1), u_2 \rangle = \langle u_1, \varphi(u_2) \rangle = \\ &= \langle u_1, \lambda_2 u_2 \rangle = \bar{\lambda}_2 \langle u_1, u_2 \rangle = \\ &= \lambda_2 \langle u_1, u_2 \rangle. \end{aligned}$$

Hence

$$(\lambda_1 - \lambda_2) \langle u_1, u_2 \rangle = 0.$$

Since $\lambda_1 - \lambda_2 \neq 0$, $\langle u_1, u_2 \rangle = 0$.

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Theorem Let $\varphi : \mathcal{H} \rightarrow \mathcal{H}$ be a selfadjoint operator. Then in \mathcal{H} there is an orthonormal basis $\alpha = (u_1, u_2, \dots, u_n)$ formed by eigenvectors. In this basis

$$(\varphi)_{\alpha, \alpha} = \begin{pmatrix} \lambda_1 & & & \\ & \lambda_2 & & 0 \\ & & \ddots & \\ 0 & & & \lambda_n \end{pmatrix}$$

where $\lambda_1, \lambda_2, \dots, \lambda_n$ are eigenvalues.

(This is an analogue to similar theorem for unitary operators!)

Proof by induction on $\dim \mathcal{H}$. I will not give it.

Corollary - spectral decomposition of self-adjoint operators.

Every selfadjoint operator is a linear combination of orthogonal projection on (mutually orthogonal) eigenspaces:

$$\varphi = \lambda_1 P_1 + \dots + \lambda_k P_k$$

where λ_i are different eigenvalues and P_i is a projection on $\ker(\varphi - \lambda_i \text{id})$.

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EXERCISE 1 Find eigenvalues and an orthonormal basis in \mathbb{R}^3 formed by eigenvectors of the linear operator $\varphi(x) = Ax$, where

$$A = \begin{pmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{pmatrix}.$$

APPLICATION TO QUADRATIC FORMS

Every quadratic form on \mathbb{R}^n

$$f(x) = \sum_{i,j=1}^n a_{ij} x_i x_j = x^T A x$$

is given by a symmetric matrix A . This matrix determines a selfadjoint operator

$$\varphi : \mathbb{R}^n \rightarrow \mathbb{R}^n \text{ given by } \varphi(x) = Ax.$$

Now we can take the basis $\alpha = (u_1, u_2, \dots, u_n)$ formed by eigenvectors of the operator φ (of the matrix A). In this basis the matrix of the quadratic form is given by the corresponding symmetric bilinear form $g : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$

$$g(u_i, u_j) = u_i^T \cdot A u_j = u_i^T \lambda_j u_j = \lambda_j \langle u_i, u_j \rangle$$

$$= \begin{cases} 0 & \text{if } i \neq j \\ \lambda_i & \text{if } i = j \end{cases}$$

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So in the coordinates of the orthonormal basis α we have

$$f(y) = \lambda_1 y_1^2 + \lambda_2 y_2^2 + \dots + \lambda_n y_n^2$$

where $\lambda_1, \lambda_2, \dots, \lambda_n$ are the eigenvalues of A .

EXERCISE 2 Find an orthonormal basis in which the quadratic form

$$f: \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

$$f(x) = 4x_1^2 + 4x_2^2 + 4x_3^2 + 4x_1x_2 + 4x_1x_3 + 4x_2x_3$$

has a diagonal form.

HOMEWORK 10 Find an orthogonal basis in which the quadratic form $g: \mathbb{R}^3 \rightarrow \mathbb{R}^3$

$$g(x) = 17x_1^2 + 4x_1x_2 - 4x_1x_3 + 14x_2^2 + 8x_2x_3 + 14x_3^2$$

has a diagonal form. (Hint: One of the eigenvalues is 18.)