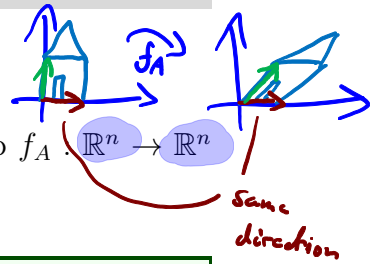


German: Eigenwert (Hilbert 1904)
 (proper value, own value, characteristic value)

6

Eigenvalues and similar things



Consider again a square matrix $A \in \mathbb{R}^{n \times n}$ and the associated linear map $f_A: \mathbb{R}^n \rightarrow \mathbb{R}^n$ which maps \mathbb{R}^n into itself.

Question:

Are there vectors \mathbf{v} which are only scaled by f_A ? This means that they satisfy:

$$\underline{A\mathbf{v} = \lambda\mathbf{v}} \text{ or equivalently } \underline{(A - \lambda\mathbb{1})\mathbf{v} = \mathbf{0}}$$

- λ is called eigenvalue of A ,
- \mathbf{v} is called eigenvector of A (if $\mathbf{v} \neq \mathbf{0}$).

Kernel of $A - \lambda\mathbb{1}$

First conclusions:

- Not very interesting (trivial): $\mathbf{v} = \mathbf{0}$.
- $\mathbf{v} \in \text{Ker}(A) \setminus \{\mathbf{0}\} \Rightarrow A\mathbf{v} = 0\mathbf{v}$, so $\lambda = 0$.
- $\mathbf{v} \in \text{Ker}(A - \lambda\mathbb{1}) \setminus \{\mathbf{0}\} \Rightarrow A\mathbf{v} = \lambda\mathbf{v}$, so λ is an eigenvalue.
- \mathbf{v} eigenvector $\Rightarrow \alpha\mathbf{v}$ is also an eigenvector (for $\alpha \neq 0$).

Example: (a) $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$, $A\mathbf{v} = \lambda\mathbf{v}$ For $\mathbf{v} \neq \mathbf{0}$

$$\Leftrightarrow \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \lambda \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$$

$$\Leftrightarrow \begin{matrix} \rightarrow \\ \rightarrow \end{matrix} \begin{pmatrix} v_1 + v_2 \\ v_2 \end{pmatrix} = \lambda \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} \quad \begin{matrix} 3 \text{ unknowns} \\ 2 \text{ equations} \\ \text{not a linear system} \end{matrix}$$

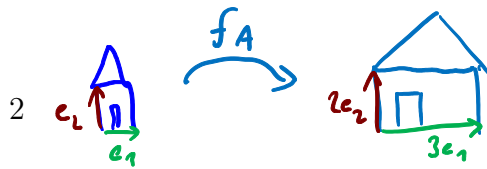
$E_2 \Rightarrow \lambda = 1$ or $v_2 = 0$

$E_1 \Rightarrow \lambda = 1$ or $v_1 = 0 \Rightarrow \mathbf{v} = \mathbf{0}$ (excluded!)

$E_1 \Rightarrow \lambda = 1$ and $v_2 = 0 \Rightarrow \lambda = 1, \mathbf{v} = \begin{pmatrix} v_1 \\ 0 \end{pmatrix}$ for all $v_1 \in \mathbb{R} \setminus \{0\}$

(b) $B = \begin{pmatrix} 3 & 0 \\ 0 & 2 \end{pmatrix}$, $B\mathbf{v} = \lambda\mathbf{v} \Leftrightarrow \begin{pmatrix} 3 & 0 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \lambda \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}$

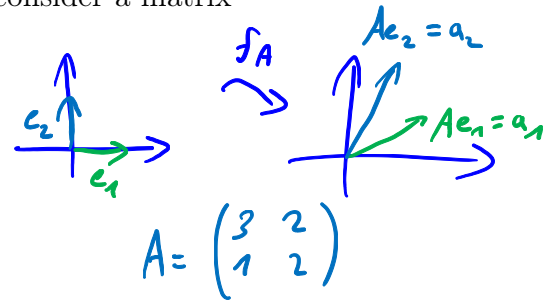
$$\Leftrightarrow \begin{pmatrix} 3v_1 \\ 2v_2 \end{pmatrix} = \lambda \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} \Rightarrow \lambda = 2, \mathbf{v} = \begin{pmatrix} 0 \\ v_2 \end{pmatrix} \text{ or } \lambda = 3, \mathbf{v} = \begin{pmatrix} v_1 \\ 0 \end{pmatrix}$$



6.1 What is an eigenvalue and an eigenvector?

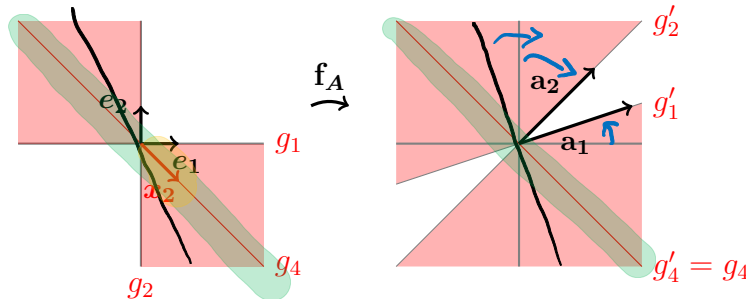
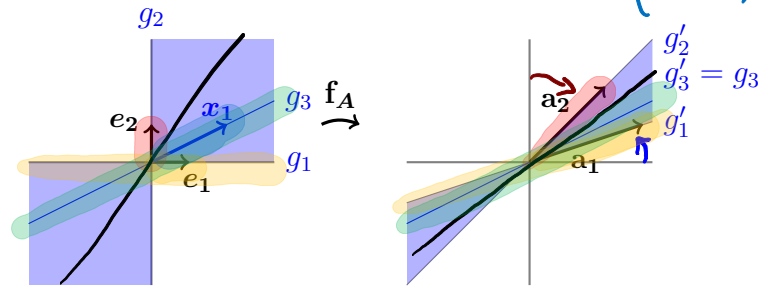
We start with an illustration in two-dimensional cases and consider a matrix

$$A = \begin{pmatrix} | & | \\ a_1 & a_2 \\ | & | \end{pmatrix} \in \mathbb{R}^{2 \times 2}$$



and the associated linear map $f_A : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ with $x \mapsto Ax$.

Images of the lines:
 $f_A(g_1) = g_1'$
 $f_A(g_2) = g_2'$
 Stable line: $f_A(g_3) = g_3$
 (Eigenvector direction!)



In the same sense, we can look at the other quadrants of our coordinate system. There we also find such a special line:

$$g_4' := f_A(g_4) = g_4.$$

For points $0 \neq x_1 \in g_3$, $0 \neq x_2 \in g_4$, $Ax_1 \in g_3$, $Ax_2 \in g_4$ $\begin{pmatrix} g_3 = \text{span}(x_1) \\ g_4 = \text{span}(x_2) \end{pmatrix}$

$\Rightarrow Ax_1 = \lambda_1 \cdot x_1$ for some $\lambda_1 \in \mathbb{R}$

$\Rightarrow Ax_2 = \lambda_2 \cdot x_2$ for some $\lambda_2 \in \mathbb{R}$

Definition 6.1. Eigenvalue, Eigenvector, spectrum

Let A be a square matrix. A vector $x \neq 0$ is called an **eigenvector** of A , if Ax is a multiple of x . This scalar λ , which means $Ax = \lambda x$, is called **eigenvalue** of A . The set of all **eigenvalues** of A is called the **spectrum** of A and denoted by $\text{spec}(A)$.

\leadsto Later: eigenvalues $\in \mathbb{C}$
 \leadsto matrices $A \in \mathbb{C}^{n \times n}$

In our example above: $A = \begin{pmatrix} 3 & 2 \\ 1 & 2 \end{pmatrix}$

$$\underline{A} \begin{pmatrix} 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 6+2 \\ 2+2 \end{pmatrix} = \begin{pmatrix} 8 \\ 4 \end{pmatrix} = \underline{4} \cdot \begin{pmatrix} 2 \\ 1 \end{pmatrix} \Rightarrow \underline{\lambda=4} \text{ eigenvalue, } \begin{pmatrix} 2 \\ 1 \end{pmatrix} \text{ eigenvector}$$

$$\underline{A} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \begin{pmatrix} 3-2 \\ 1-2 \end{pmatrix} = \underline{-1} \cdot \begin{pmatrix} 1 \\ -1 \end{pmatrix} \Rightarrow \underline{\lambda=-1} \text{ eigenvalue, } \begin{pmatrix} 1 \\ -1 \end{pmatrix} \text{ eigenvector}$$

This is very general definition and will work later for other cases in the same manner. Here, we are first interested in matrices $A \in \mathbb{R}^{n \times n}$ and eigenvalues $\lambda \in \mathbb{R}$. However, you may already see that this can also work for complex numbers. We may also include $\lambda \in \mathbb{C}$ later.

Proposition 6.2. Multiple of eigenvector = eigenvector

Every multiple (not 0) of an eigenvector \mathbf{x} for A is also an eigenvector for A , corresponding to the same eigenvalue λ .

Proof: $\underline{Ax} = \lambda \underline{x} \Leftrightarrow \alpha(Ax) = \alpha(\lambda x), \alpha \neq 0$
 $\Leftrightarrow \underline{A(\alpha \cdot x)} = \lambda(\alpha \cdot x)$

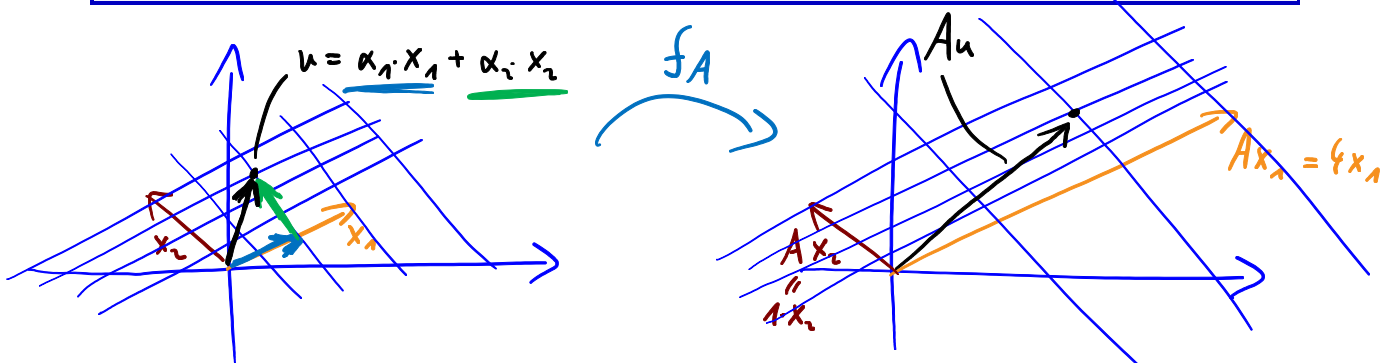
Looking again at the pictures above:

- We have $Ax = \lambda_1 x$ for all multiples x of $x_1 \in g_3$ (which means for all $x \in g_3$).
- Also we have $Ax = \lambda_2 x$ for all multiples x of $x_2 \in g_4$ (which means for all $x \in g_4$).
- Looking at the line g_3 , the map f_A acts like scaling with the factor λ_1 .
- Looking at the line g_4 , the map f_A acts like scaling with the factor λ_2 .

Optimal

Perfect coordinate system for the map f_A

Describing \mathbb{R}^2 with a coordinate system given by the two lines g_3 and g_4 (instead of g_1 and g_2), the acting of the map f_A is very simple: The coordinate axes are only stretched: The one with factor λ_1 , and the other one with factor λ_2 .



$$Au = \alpha_1 \underline{(Ax_1)} + \alpha_2 \underline{(Ax_2)}$$

6 Eigenvalues and similar things

$$= \alpha_1 (4x_1) + \alpha_2 (1 \cdot x_2)$$

To get this “perfect coordinate system” we need all the eigenvalues λ_1, λ_2 and the corresponding eigenvectors \mathbf{x}_1 and \mathbf{x}_2 .

Question:

- (a) How to find the eigenvalues and the eigenvectors of A ?
- (b) Do you always find n eigenvalues for an $n \times n$ matrix A ?
- (c) Do you find n different directions for eigenvectors?
- (d) How to change the coordinate system?
- (e) What are applications for this?

6.2 The characteristic polynomial

Our goal is to find $\lambda \in \mathbb{R}$ and $\mathbf{x} \neq \mathbf{o}$ such that $(A - \lambda \mathbf{1})\mathbf{x} = \mathbf{o}$, i.e., $(A - \lambda \mathbf{1})$ has a nontrivial kernel. This means that the corresponding map for $A - \lambda \mathbf{1}$ is not injective and, hence, it is a singular matrix.

Idea:

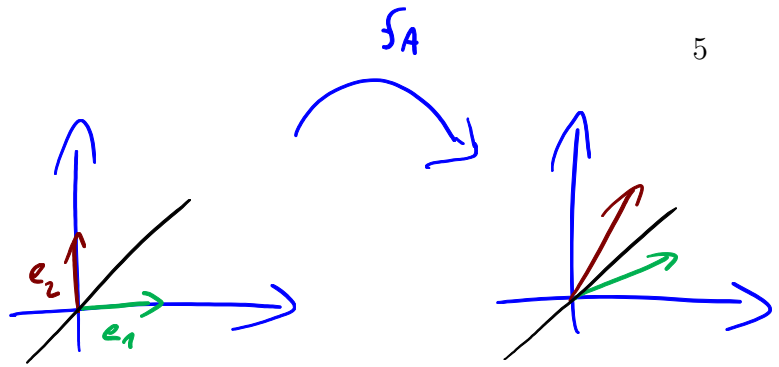
Compute $\det(A - \lambda \mathbf{1})$, which yields a polynomial of degree n in λ and determine its zeros, because

$$\begin{aligned} \det(A - \lambda \mathbf{1}) = 0 &\Leftrightarrow A - \lambda \mathbf{1} \text{ is singular} \\ &\Leftrightarrow \text{Ker}(A - \lambda \mathbf{1}) \text{ is non-trivial} \\ &\Leftrightarrow \lambda \text{ is an eigenvalue} \end{aligned}$$

Then, compute a basis for $\text{Ker}(A - \lambda \mathbf{1})$ for each eigenvalue.

solving an LES

$\mathbf{x} \in \text{Ker}(A - \lambda \mathbf{1})$ eigenvector



Example 6.3.

$$A = \begin{pmatrix} 3 & 2 \\ 1 & 4 \end{pmatrix}$$

$$\det(A - \lambda \mathbf{1}) = \det \begin{pmatrix} 3-\lambda & 2 \\ 1 & 4-\lambda \end{pmatrix} = (3-\lambda)(4-\lambda) - 2 \cdot 1 = 10 - 7\lambda + \lambda^2 = 0$$

quadratic equation

$$\lambda_{1,2} = \frac{7 \pm \sqrt{49 - 40}}{2} = \frac{7 \pm 3}{2} \Rightarrow \lambda_1 = 2, \lambda_2 = 5$$

Thus we have the eigenvalues $\lambda_1 = 2$ and $\lambda_2 = 5$. Let us compute the eigenvectors:

$$\mathbf{0} = (A - 2\mathbf{1})\mathbf{v} = \begin{pmatrix} 1 & 2 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} v_1 + 2v_2 \\ v_1 + 2v_2 \end{pmatrix} \Rightarrow \mathbf{v} = \alpha \begin{pmatrix} 2 \\ -1 \end{pmatrix}, \alpha \in \mathbb{R}$$

$$\mathbf{0} = (A - 5\mathbf{1})\mathbf{v} = \begin{pmatrix} -2 & 2 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} v_1 \\ v_2 \end{pmatrix} = \begin{pmatrix} -2v_1 + 2v_2 \\ v_1 - v_2 \end{pmatrix} \Rightarrow \mathbf{v} = \alpha \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \alpha \in \mathbb{R}$$

$\rightarrow \begin{pmatrix} 2 \\ -1 \end{pmatrix}$ is an eigenvector of A , to eigenvalue $\lambda_1 = 2$ $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ is eigenvector of A to eigenvalue of $\lambda_2 = 5$

Proposition 6.4. Five properties of an eigenvalue

For a square matrix A and a number λ the following is equivalent:

- (i) λ is an eigenvalue A .
- (ii) There is a vector $\mathbf{x} \neq \mathbf{0}$ with $A\mathbf{x} = \lambda\mathbf{x}$.
- (iii) The space $\text{Ker}(A - \lambda\mathbf{1})$ contains a vector $\mathbf{x} \neq \mathbf{0}$.
- (iv) The matrix $A - \lambda\mathbf{1}$ is not invertible.
- (v) $\det(A - \lambda\mathbf{1}) = 0$

$$\det(A - \lambda \mathbf{1}) = \det \begin{pmatrix} a_{11} - \lambda & a_{12} & \dots & \vdots \\ a_{21} & a_{22} - \lambda & & \vdots \\ \vdots & & \ddots & \vdots \\ \vdots & & & a_{nn} - \lambda \end{pmatrix}$$

Let $A \in \mathbb{R}^{n \times n}$. Then we observe that $\det(A - \lambda \mathbf{1}) = p_A(\lambda)$ is a polynomial of order n in the variable λ . For example, there could be coefficients c_i such that

$$p_A(\lambda) = (-1)^n \lambda^n + c_{n-1} \lambda^{n-1} + \dots + c_1 \lambda + c_0.$$

Definition 6.5. Characteristic polynomial

For an $n \times n$ -Matrix A , the polynomial $\lambda \mapsto \det(A - \lambda \mathbb{1})$ is called the **characteristic polynomial** of the matrix A and is denoted by p_A .

Example 6.6. Look at $A = \begin{pmatrix} 3 & 2 \\ 1 & 2 \end{pmatrix}$.

$$p_A(\lambda) = \det(A - \lambda \mathbb{1}) = \det\left(\begin{pmatrix} 3 & 2 \\ 1 & 2 \end{pmatrix} - \lambda \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}\right) = \det\begin{pmatrix} 3-\lambda & 2 \\ 1 & 2-\lambda \end{pmatrix}$$

$$= (3-\lambda) \cdot (2-\lambda) - 2 \cdot 1 = 6 - 3\lambda - 2\lambda + \lambda^2 - 2 = \lambda^2 - 5\lambda + 4$$

other solutions are possible

char. polynomial for A .

Solving the quadratic equation:

$$\lambda_{1,2} = -\frac{-5}{2} \pm \sqrt{\frac{25}{4} - 4} = \frac{5}{2} \pm \sqrt{\frac{9}{4}} = \frac{5 \pm 3}{2} \in \{1, 4\}, \quad \text{hence } \underline{\lambda_1 = 4}, \underline{\lambda_2 = 1}.$$

$$\text{Ker}(A - \overset{\lambda_2}{1} \cdot \mathbb{1}) = \text{Ker}\begin{pmatrix} 2 & 2 \\ 1 & 1 \end{pmatrix} = ?$$

$$\text{Ker}(A - 4 \cdot \mathbb{1}) = \text{Ker}\begin{pmatrix} -1 & 2 \\ 1 & -2 \end{pmatrix} = ?$$

Theorem 6.7. Fundamental theorem of algebra (Gauß 1799)

Let $a_0, a_1, \dots, a_n \in \mathbb{C}$ with $a_n \neq 0$. Then the polynomial equation

$$\underbrace{a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0}_{=: p(x)} = 0$$

has n (not necessarily different) solutions x_1, \dots, x_n in \mathbb{C} . Moreover, we find for $x \in \mathbb{C}$:

$$p(x) = a_n(x - x_1)(x - x_2) \cdots (x - x_n).$$

need \mathbb{C}

Looking at p_A :

$\rightarrow p_A(\lambda) = 0$ has at least one solution in \mathbb{C}

$\Rightarrow A$ has at least one eigenvalue in \mathbb{C}

$$\left(p_A(\lambda) = \lambda^2 + 1 \Rightarrow \lambda_{1,2} = \pm i \right)$$

$$A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}$$

$\rightarrow p_A(\lambda) = 0$ multiple roots \Rightarrow multiplicity of an eigenvalue

$$p_A(x) = (\pm 1)(x - \lambda_1)(x - \lambda_2) \cdots (x - \lambda_n)$$

6.2 The characteristic polynomial

for example: $p_A(x) = (x-1)^2(x-3)^4$

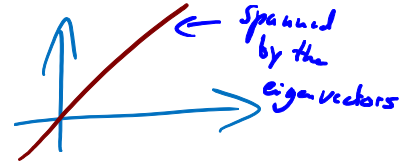
Definition 6.8.

If the same eigenvalue λ appears $\alpha(\lambda)$ times in this factorisation, we say:

λ has algebraic multiplicity $\alpha(\lambda)$.

- If we have k different eigenvalues $\lambda_1, \dots, \lambda_k \in \mathbb{C}$, then $\alpha(\lambda_1) + \dots + \alpha(\lambda_k) = n$, because polynomials of degree n can be factorised into n linear factors.
- If λ is an eigenvalue, then $A - \lambda \mathbb{1}$ is singular, so $\gamma(\lambda) := \dim(\text{Ker}(A - \lambda \mathbb{1})) \geq 1$.

↳ geometric multiplicity



Proposition 6.9. Spectrum for triangular matrices

Let $A \in \mathbb{R}^{n \times n}$ a square matrix.

(a) For a matrix in triangular form

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ 0 & a_{22} & & a_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & a_{nn} \end{pmatrix},$$

we get $\text{spec}(A) = \{a_{11}, a_{22}, \dots, a_{nn}\}$.

(b) For a square block matrix in triangular form

$$A = \begin{pmatrix} B & C \\ 0 & D \end{pmatrix}$$

with square matrices B and D , we get $\text{spec}(A) = \text{spec}(B) \cup \text{spec}(D)$.

(c) Also $\text{spec}(A) = \text{spec}(A^T)$. Hence (a) and (b) also hold for lower triangular matrices.

Proof.

$$(c) \lambda \text{ eigenvalue of } A \Leftrightarrow \det(A - \lambda \mathbb{1}) = 0$$

$$\Leftrightarrow \det((A - \lambda \mathbb{1})^T) = 0$$

$$\Leftrightarrow \det(A^T - \lambda \mathbb{1}) = 0$$

$$\Leftrightarrow \lambda \text{ eigenvalue of } A^T$$

(b), (a)

□