

## Theory sheet 5

### Definition of eigenvalues and eigenvectors

**Definition 1.** Let  $A \in M_{n,n}$  be a square matrix and  $\lambda \in \mathbb{R}$  a number. We say that a non-zero vector  $\mathbf{x} \in \mathbb{R}^n$  is an eigenvector of  $A$  corresponding to the eigenvalue  $\lambda$  if

$$A\mathbf{x} = \lambda\mathbf{x}.$$

In other words, if  $A$  sends  $\mathbf{x}$  to itself, but stretched by  $\lambda$ .

A few remarks:

- If  $\mathbf{x}$  is an eigenvector of  $A$  and  $\lambda$  is its eigenvalue, then

$$A^2\mathbf{x} = \lambda^2\mathbf{x}, \quad A^3\mathbf{x} = \lambda^3\mathbf{x}, \quad \dots, \quad A^n\mathbf{x} = \lambda^n\mathbf{x}$$

for all  $n$ .

- Compare notion of eigenvector and eigenvalue with scaling (homothetic) linear transformation from previous lecture. The transformation

$$S_\lambda(\mathbf{x}) = \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

stretches every vector by  $\lambda$ , whereas  $A$  with eigenvalue  $\lambda$  stretches by  $\lambda$  only some vectors – its eigenvectors.

- Note that  $\lambda$  is clearly an eigenvalue of  $\begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix}$ , and every vector is its eigenvector.
- If a matrix is proportional to the identity, that is, of the form  $A = \lambda I$ , we say that  $A$  is a scalar matrix. Scalar matrices are the only matrices which commute with every other:  $AB = BA$  for all  $B$  implies that  $A = \lambda I$  for some number  $\lambda$ . We will show later on that  $\lambda$  is the only eigenvalue of  $A = \lambda I$ .
- Note that if  $\lambda$  is an eigenvalue of  $A$  and we want to find  $\mathbf{x}$  such that  $A\mathbf{x} = \lambda\mathbf{x}$ , then there are an infinite number of solutions. Why? Because if  $\mathbf{x}$  is one such solution, then so is  $2\mathbf{x}$ , as well as  $3\mathbf{x}$ ,  $4\mathbf{x}$ ,  $\pi\mathbf{x}$ , et cetera, because  $A(2\mathbf{x}) = 2A\mathbf{x} = 2\lambda\mathbf{x} = \lambda(2\mathbf{x})$ .
- Eigenvalues and eigenvectors only make sense for square matrices. Indeed, if  $A \in M_{m,n}$ , then  $A\mathbf{x} \in \mathbb{R}^m$ , but  $A\mathbf{x} = \lambda\mathbf{x} \in \mathbb{R}^n$ . Hence,  $m = n$ .

# Finding eigenvalues and eigenvectors

**Remark 1.** Recall that if  $A\mathbf{x} = \mathbf{0}$  and  $A$  is invertible, then  $A^{-1}A\mathbf{x} = \mathbf{x}$  on one hand and  $A^{-1}A\mathbf{x} = A^{-1}\mathbf{0} = \mathbf{0}$  on the other, so  $\mathbf{x}$  must be zero.

**Theorem 1.**  $\lambda$  is an eigenvalue of  $A$  if and only if  $\det(A - \lambda I) = 0$ .

*Proof.* ( $\implies$ ) If  $\lambda$  is an eigenvalue, then  $A\mathbf{x} = \lambda\mathbf{x}$  for some non-zero vector  $\mathbf{x}$ . Hence,

$$A\mathbf{x} - \lambda\mathbf{x} = \mathbf{0}, \quad \text{hence } (A - \lambda I)\mathbf{x} = \mathbf{0}.$$

Since  $A - \lambda I$  sends a non-zero vector to zero, it cannot be invertible. Hence,  $\det(A - \lambda I) = 0$ , as claimed.

( $\impliedby$ ) If  $\det(A - \lambda I) = 0$ , then  $A - \lambda I$  is not invertible. Hence, there exists a non-zero vector  $\mathbf{x}$  which it sends to zero:

$$(A - \lambda I)\mathbf{x} = \mathbf{0}.$$

Hence,  $A\mathbf{x} = \lambda\mathbf{x}$ , as claimed.  $\square$

The following theorem is given without proof:

**Theorem 2.** If  $A \in M_{n,n}$ , the function  $p(\lambda) = \det(A - \lambda I)$  is a polynomial of degree  $n$ . It is called the characteristic polynomial of matrix  $A$ .

An important corollary of this theorem is that eigenvalues always exist:

**Corollary 1.** Any square matrix  $A \in M_{n,n}$  has at least one eigenvalue.

*Proof.* Eigenvalues of  $A$  are the roots of  $p(\lambda)$ . Since  $p$  is a polynomial, it has a root (fundamental theorem of algebra).  $\square$

Recipe for finding eigenvalues. To find eigenvalues of  $A$ ,

- denote an unknown eigenvalue by  $\lambda$
- subtract  $\lambda I$  from  $A$
- calculate the characteristic polynomial  $p(\lambda) = \det(A - \lambda I)$
- set it equal to zero:  $\det(A - \lambda I) = 0$
- find solutions of this equation.

## Example

$$\begin{aligned} A = \begin{pmatrix} -1 & 4 \\ -3 & -8 \end{pmatrix} \implies \det(A - \lambda I) &= \det \begin{pmatrix} -1 - \lambda & 4 \\ -3 & -8 - \lambda \end{pmatrix} \\ &= (-1 - \lambda)(-8 - \lambda) - 4 \cdot (-3) \\ &= 8 + 8\lambda + \lambda + \lambda^2 + 12 \\ &= \lambda^2 + 9\lambda + 20. \end{aligned}$$

Therefore, we need to solve

$$\lambda^2 + 9\lambda + 20 = 0.$$

It is clear that there are two solutions:

$$\lambda_1 = -4 \quad \text{and} \quad \lambda_2 = -5.$$

## Finding eigenvectors

To find eigenvectors, we need to first find the eigenvalues. If eigenvalues are known, we can find corresponding eigenvectors by solving the linear system

$$(A - \lambda I) \mathbf{x} = \mathbf{0}.$$

To solve this system, use approach developed in the previous lecture.

## Example

We have found above that  $\lambda_1 = -4$  and  $\lambda_2 = -5$  are eigenvalues of  $\begin{pmatrix} -1 & 4 \\ -3 & -8 \end{pmatrix}$ . Let us find the eigenvectors corresponding to  $\lambda_1$ :

$$A - \lambda_1 I = A + 4I = \begin{pmatrix} 3 & 4 \\ -3 & -4 \end{pmatrix}.$$

Hence, we need to find  $\mathbf{x}$  such that

$$\begin{pmatrix} 3 & 4 \\ -3 & -4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

Adding first line to the second, we obtain

$$\begin{pmatrix} 3 & 4 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

We can drop the second line because there is zero against it, hence any  $\mathbf{x}$  satisfying

$$3x_1 + 4x_2 = 0$$

is an eigenvector. We can parametrize these eigenvectors by  $x_2$ :

$$\left\{ \begin{pmatrix} -4x_2/3 \\ x_2 \end{pmatrix}, \quad x_2 \in \mathbb{R} \right\}.$$

We also need to find the eigenvectors corresponding to  $\lambda = -5$ :

$$(A - \lambda I) \mathbf{x} = \mathbf{0} \iff \begin{pmatrix} 4 & 4 \\ -3 & -3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix},$$

and the set of solutions is given by

$$\left\{ \begin{pmatrix} -x_2 \\ x_2 \end{pmatrix}, \quad x_2 \in \mathbb{R} \right\}.$$

## Properties of eigenvalues and eigenvectors

**Definition 2.** *The set of eigenvectors  $\mathbf{x}$  corresponding to an eigenvalue  $\lambda$  of  $A$  is called the eigenspace.*

Proof of this theorem is left as an exercise:

**Theorem 3.** *The eigenspace of  $A$  corresponding to  $\lambda$  is a vector space. In other words, if  $\mathbf{x}, \mathbf{y}$  satisfy  $A\mathbf{x} = \lambda\mathbf{x}$  and  $A\mathbf{y} = \lambda\mathbf{y}$ , then (a)  $\mathbf{x} + \mathbf{y}$  also satisfies this:  $A(\mathbf{x} + \mathbf{y}) = \lambda(\mathbf{x} + \mathbf{y})$  and (b) for any number  $\mu$ ,  $A(\mu\mathbf{x}) = \lambda(\mu\mathbf{x})$ .*

The following theorem is a simple yet deep result:

**Theorem 4.** *Determinant of  $A$  is equal to the product of all eigenvalues of  $A$ :*

$$\det A = \lambda_1 \cdot \lambda_2 \cdots \lambda_n = \prod_{i=1}^n \lambda_i.$$

*Proof.* Note that  $\det A = \det(A - \lambda I)|_{\lambda=0} = p(0)$ , where  $p$  is the characteristic polynomial of  $A$ . By Vieta's theorem,  $p(0)$  is equal to the product of all roots of  $p$  for all polynomials (not just characteristic polynomials of matrices).  $\square$

**Corollary 2.**  *$A$  is invertible if and only if it does not have zero eigenvalues.*

*Proof.* If  $\lambda_i \neq 0$  for all  $i$ , then  $\det A = \prod_{i=1}^n \lambda_i \neq 0$ , so the matrix is invertible. On the other hand, if  $A$  is invertible, we have that  $\det A \neq 0$  and by this product representation there cannot be  $\lambda_i = 0$  among eigenvalues of  $A$ .  $\square$

**Remark 2.** *Hence, if  $A \in M_{2,2}$  and we somehow know one eigenvalue, we can find the other using previous theorem. For example, if we are given that  $\lambda_1 = 1$  is an eigenvalue of  $\begin{pmatrix} 1-a & a \\ b & 1-b \end{pmatrix}$ , we can find the second eigenvalue  $\lambda_2$  by*

$$\lambda_2 = \lambda_1 \cdot \lambda_2 = \det \begin{pmatrix} 1-a & a \\ b & 1-b \end{pmatrix} = (1-a)(1-b) - ab = 1 - a - b.$$

**Theorem 5.** If  $A$  is symmetric ( $A^\top = A$ ), then eigenvectors corresponding to different eigenvalues are orthogonal. In other words,  $\mathbf{x}^\top \mathbf{y} = 0$  for every  $\mathbf{x}$  such that  $A\mathbf{x} = \lambda\mathbf{x}$ ,  $\mathbf{y}$  such that  $A\mathbf{y} = \mu\mathbf{y}$  and  $\lambda \neq \mu$ .

*Proof.* Since  $\lambda \neq \mu$ , we have  $\lambda - \mu \neq 0$ , therefore

$$\begin{aligned} (\lambda - \mu) \mathbf{x}^\top \mathbf{y} &= \lambda \mathbf{x}^\top \mathbf{y} - \mu \mathbf{x}^\top \mathbf{y} \\ &= \lambda \mathbf{y}^\top \mathbf{x} - \mu \mathbf{x}^\top \mathbf{y} \\ &= \mathbf{y}^\top (\lambda \mathbf{x}) - \mathbf{x}^\top (\mu \mathbf{y}) \\ &= \mathbf{y}^\top A\mathbf{x} - \mathbf{x}^\top A\mathbf{y} \\ &= \mathbf{x}^\top A^\top \mathbf{y} - \mathbf{x}^\top A\mathbf{y} \\ &= 0, \end{aligned}$$

where the last line follows from  $A^\top = A$ . Hence,  $\mathbf{x}^\top \mathbf{y} = 0$ .  $\square$

**Remark 3.** Eigenvectors may be orthogonal without  $A = A^\top$ .

**Remark 4.** Two different matrices may have the same characteristic polynomial. For example,

$$A = \begin{pmatrix} 2 & 2 \\ 1 & 3 \end{pmatrix} \quad \text{and} \quad B = \begin{pmatrix} 1 & 5 \\ 0 & 4 \end{pmatrix}.$$

Hence, they have the same eigenvalues, but their eigenvectors are different.