## Math 22 – Linear Algebra and its applications

- Lecture 25 -

Instructor: Bjoern Muetzel

## **GENERAL INFORMATION**

• **Office hours:** Tu 1-3 pm, **Th**, Sun **2-4 pm** in **KH 229** 

Tutorial: Tu, Th, Sun 7-9 pm in KH 105

• <u>Homework 8</u>: due Wednesday at 4 pm outside KH 008. There is only Section B,C and D.

# **5** Eigenvalues and Eigenvectors

5.1

#### EIGENVECTORS AND EIGENVALUES



FIFTH EDITION

DAVID C. LAY • STEVEN R. LAY • JUDI J. McDonald



#### **Summary:**

Given a linear transformation  $T: \mathbb{R}^n \to \mathbb{R}^n$ , then there is always a **good basis** on which the **transformation** has a **very simple form**. To find this basis we have to find the **eigenvalues of** *T*.

### **GEOMETRIC INTERPRETATION**

**Example:** Let  $A = \begin{bmatrix} 5 & -3 \\ -6 & 2 \end{bmatrix}$  and let  $u = x_0 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  and  $v = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ .

1.) Find Av and Au.

Draw a picture of v and Av and u and Au.

2.) Find A(3u + 2v) and  $A^2 (3u + 2v)$ . Hint: Use part 1.)

• **Definition:** An **eigenvector** of an  $n \times n$  matrix A is a **nonzero** vector **x** such that

$$Ax = \lambda x$$
 for so

for some scalar  $\lambda$  in  $\mathbb{R}$ .

In this case  $\lambda$  is called an **eigenvalue** and the solution  $\mathbf{x} \neq \mathbf{0}$  is called an **eigenvector corresponding to**  $\lambda$ .

**Definition:** Let *A* be an  $n \times n$  matrix. The set of solutions  $\mathbf{Eig}(A, \lambda) = \{x \text{ in } \mathbb{R}^n, \text{ such that } (A - \lambda I_n) x = 0\}$ is called the **eigenspace Eig**( $A, \lambda$ ) of *A* corresponding to  $\lambda$ .

It is the null space of the matrix  $A - \lambda I_n$ :

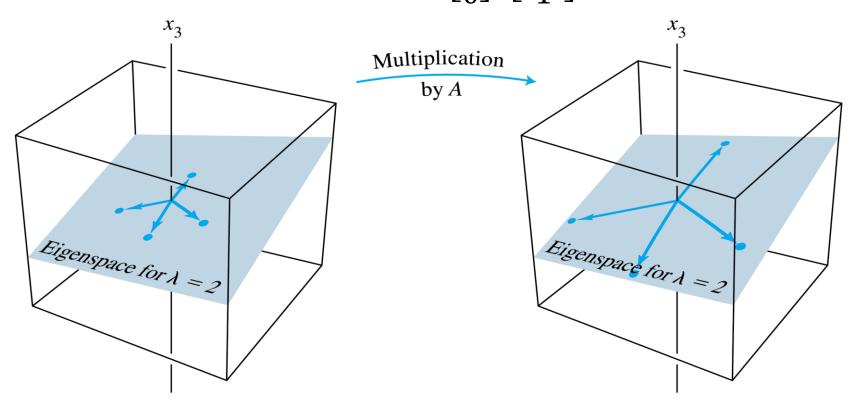
$$\operatorname{Eig}(A, \lambda) = \operatorname{Nul}(A - \lambda I_n)$$

**Example:** Show that  $\lambda = 7$  is an eigenvalue of matrix  $A = \begin{vmatrix} 1 & 6 \\ 5 & 2 \end{vmatrix}$ 

and find the corresponding eigenspace Eig(A,7).

• Example: Let  $A = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix}$ . An eigenvalue of A is  $\lambda = 2$ . Find a basis for the corresponding eigenspace Eig(A,2).

• The eigenspace  $\operatorname{Eig}(A,2) = \operatorname{Span}\left\{ \begin{bmatrix} 1\\2\\0 \end{bmatrix}, \begin{bmatrix} -3\\0\\1 \end{bmatrix} \right\}$  is a subspace of  $\mathbb{R}^3$ .



A acts as a dilation on the eigenspace.

## THEOREMS ABOUT EIGENVALUES

- **Theorem 1:** The eigenvalues of a triangular matrix are the entries on its main diagonal.
- Warning: We can not find the eigenvalues of a matrix A by row reducing to echelon form U. As A and U have usually different eigenvalues.
- Proof of Theorem 1:

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- Theorem 2: If v<sub>1</sub>, ..., v<sub>r</sub> are eigenvectors that correspond to distinct eigenvalues λ<sub>1</sub>, ..., λ<sub>r</sub> of an n × n matrix A, then the set {v<sub>1</sub>, ..., v<sub>r</sub>} is linearly independent.
- **Proof:** Suppose  $S = \{v_1, ..., v_r\}$  is linearly dependent.
- Then there is a subset of S={v<sub>1</sub>, ..., v<sub>r</sub>}, say {v<sub>1</sub>, ..., v<sub>p</sub>} that is a basis for Span(S) and a vector, say v<sub>p+1</sub> that is a linear combination of these vectors.
- Then there exist scalars  $c_1, \ldots, c_p$  such that

$$c_1 v_1 + \dots + c_p v_p = v_{p+1}$$
 (1)

• Multiplying both sides of (1) by *A* and using the fact that  $Av_k = \lambda_k v_k$  for each *k*, we obtain by the linearity of *A* 

or (2)

Multiplying both sides of (1) by λ<sub>p+1</sub> and substituting the result in the right hand side of (2), we obtain

= 0. (3)

- Since {v<sub>1</sub>, ..., v<sub>p</sub>} is linearly independent, the weights in (3) are all zero. But none of the factors λ<sub>i</sub> − λ<sub>p+1</sub> are zero, because the eigenvalues are distinct. Hence c<sub>i</sub> = 0 for i = 1, ..., p.
- But then (1) states that  $v_{p+1} = 0$ , which is impossible.

#### EIGENVECTORS AND DIFFERENCE EQUATIONS

#### Application to a recursive sequence in $\mathbb{R}^n$

Let A be an  $n \times n$  matrix and consider the **recursive sequence**  $\{x_k\}$  in  $\mathbb{R}^n$  given by  $x_0 = u$  in  $\mathbb{R}^n$  and  $x_{k+1} = Ax_k$  for k = 0, 1, 2, 3, ...,

**Definition:** We call a **solution** of this equation an <u>explicit description</u> of  $\{x_k\}$  whose formula for each  $x_k$  does **not depend directly on** A or on the preceding terms in the sequence <u>other than</u> the initial term  $x_0 = u$ .

Note: It follows that

$$x_{k+1} = A^k x_0 = A^k u,$$

However, this is **not explicit enough** to be a solution.

#### **Proof of the Note:**

#### EIGENVECTORS AND DIFFERENCE EQUATIONS

• **Example:** Let A be an  $n \times n$  matrix such that

$$Ab_1 = 2b_1$$
 and  $Ab_2 = \frac{1}{3}b_2$  where  $b_1, b_2 \neq 0$ .

- 1.) Calculate  $A^2 b_1$  and  $A^2 b_2$ .
- 2.) Calculate  $A^k b_1$  and  $A^k b_2$  and describe geometrically what happens to  $A^k b_1$  and  $A^k b_2$ .
- 3.) Find a formula for  $A^k(4b_1+5b_2)$ .