# MATH 311 Topics in Applied Mathematics Lecture 15: Eigenvalues and eigenvectors (continued).

#### Eigenvalues and eigenvectors of a matrix

Definition. Let A be an  $n \times n$  matrix. A number  $\lambda \in \mathbb{R}$  is called an **eigenvalue** of the matrix A if  $\overline{A\mathbf{v} = \lambda \mathbf{v}}$  for a nonzero column vector  $\mathbf{v} \in \mathbb{R}^n$ . The vector  $\mathbf{v}$  is called an **eigenvector** of A belonging to (or associated with) the eigenvalue  $\lambda$ .

If  $\lambda$  is an eigenvalue of A then the nullspace  $N(A - \lambda I)$ , which is nontrivial, is called the **eigenspace** of A corresponding to  $\lambda$ . The eigenspace consists of all eigenvectors belonging to the eigenvalue  $\lambda$  plus the zero vector.

## How to find eigenvalues and eigenvectors?

**Theorem** Given a square matrix A and a scalar  $\lambda$ , the following statements are equivalent:

• 
$$\lambda$$
 is an eigenvalue of  $A$ ,

• 
$$N(A - \lambda I) \neq \{\mathbf{0}\},\$$

• the matrix  $A - \lambda I$  is singular,

• 
$$det(A - \lambda I) = 0.$$

Definition.  $det(A - \lambda I) = 0$  is called the **characteristic equation** of the matrix A.

Eigenvalues  $\lambda$  of A are roots of the characteristic equation. Associated eigenvectors of A are nonzero solutions of the equation  $(A - \lambda I)\mathbf{x} = \mathbf{0}$ .

Example. 
$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$
.  
 $det(A - \lambda I) = \begin{vmatrix} a - \lambda & b \\ c & d - \lambda \end{vmatrix}$  $= (a - \lambda)(d - \lambda) - bc$  $= \lambda^2 - (a + d)\lambda + (ad - bc).$ 

Example. 
$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}$$
.

$$\det(A - \lambda I) = egin{bmatrix} a_{11} - \lambda & a_{12} & a_{13} \ a_{21} & a_{22} - \lambda & a_{23} \ a_{31} & a_{32} & a_{33} - \lambda \ = -\lambda^3 + c_1\lambda^2 - c_2\lambda + c_3, \end{cases}$$

where  $c_1 = a_{11} + a_{22} + a_{33}$  (the *trace* of A),  $c_2 = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} + \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix} + \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix}$ ,  $c_3 = \det A$ . **Theorem.** Let  $A = (a_{ij})$  be an  $n \times n$  matrix. Then  $det(A - \lambda I)$  is a polynomial of  $\lambda$  of degree n:  $det(A - \lambda I) = (-1)^n \lambda^n + c_1 \lambda^{n-1} + \dots + c_{n-1} \lambda + c_n$ . Furthermore,  $(-1)^{n-1}c_1 = a_{11} + a_{22} + \dots + a_{nn}$ and  $c_n = det A$ .

*Definition.* The polynomial  $p(\lambda) = det(A - \lambda I)$  is called the **characteristic polynomial** of the matrix A.

**Corollary** Any  $n \times n$  matrix has at most n eigenvalues.

Example. 
$$A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$$
.  
Characteristic equation:  $\begin{vmatrix} 2 - \lambda & 1 \\ 1 & 2 - \lambda \end{vmatrix} = 0.$   
 $(2 - \lambda)^2 - 1 = 0 \implies \lambda_1 = 1, \ \lambda_2 = 3.$   
 $(A - I)\mathbf{x} = \mathbf{0} \iff \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$   
 $\iff \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \iff x + y = 0$ 

The general solution is (-t, t) = t(-1, 1),  $t \in \mathbb{R}$ . Thus  $\mathbf{v}_1 = (-1, 1)$  is an eigenvector associated with the eigenvalue 1. The corresponding eigenspace is the line spanned by  $\mathbf{v}_1$ .

$$(A-3I)\mathbf{x} = \mathbf{0} \iff \begin{pmatrix} -1 & 1\\ 1 & -1 \end{pmatrix} \begin{pmatrix} x\\ y \end{pmatrix} = \begin{pmatrix} 0\\ 0 \end{pmatrix}$$
$$\iff \begin{pmatrix} 1 & -1\\ 0 & 0 \end{pmatrix} \begin{pmatrix} x\\ y \end{pmatrix} = \begin{pmatrix} 0\\ 0 \end{pmatrix} \iff x - y = \mathbf{0}.$$

The general solution is (t, t) = t(1, 1),  $t \in \mathbb{R}$ . Thus  $\mathbf{v}_2 = (1, 1)$  is an eigenvector associated with the eigenvalue 3. The corresponding eigenspace is the line spanned by  $\mathbf{v}_2$ .

Summary. 
$$A = \begin{pmatrix} 2 & 1 \ 1 & 2 \end{pmatrix}$$
.

- The matrix A has two eigenvalues: 1 and 3.
- The eigenspace of A associated with the eigenvalue 1 is the line t(-1, 1).

• The eigenspace of A associated with the eigenvalue 3 is the line t(1, 1).

• Eigenvectors  $\mathbf{v}_1 = (-1, 1)$  and  $\mathbf{v}_2 = (1, 1)$  of the matrix A form a basis for  $\mathbb{R}^2$ .

• Geometrically, the mapping  $\mathbf{x} \mapsto A\mathbf{x}$  is a stretch by a factor of 3 away from the line x + y = 0 in the orthogonal direction.

Example. 
$$A = \begin{pmatrix} 1 & 1 & -1 \\ 1 & 1 & 1 \\ 0 & 0 & 2 \end{pmatrix}$$
.

Characteristic equation:

$$egin{array}{ccc|c} 1-\lambda & 1 & -1 \ 1 & 1-\lambda & 1 \ 0 & 0 & 2-\lambda \end{array} = 0.$$

Expand the determinant by the 3rd row:

$$(2-\lambda)\begin{vmatrix} 1-\lambda & 1\\ 1 & 1-\lambda \end{vmatrix} = 0.$$

$$((1-\lambda)^2-1)(2-\lambda)=0 \iff -\lambda(2-\lambda)^2=0$$
  
 $\implies \lambda_1=0, \ \lambda_2=2.$ 

$$A\mathbf{x} = \mathbf{0} \iff \begin{pmatrix} 1 & 1 & -1 \\ 1 & 1 & 1 \\ 0 & 0 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Convert the matrix to reduced row echelon form:

$$\begin{pmatrix} 1 & 1 & -1 \\ 1 & 1 & 1 \\ 0 & 0 & 2 \end{pmatrix} \to \begin{pmatrix} 1 & 1 & -1 \\ 0 & 0 & 2 \\ 0 & 0 & 2 \end{pmatrix} \to \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$
$$A\mathbf{x} = \mathbf{0} \iff \begin{cases} x + y = 0, \\ z = 0. \end{cases}$$

The general solution is (-t, t, 0) = t(-1, 1, 0),  $t \in \mathbb{R}$ . Thus  $\mathbf{v}_1 = (-1, 1, 0)$  is an eigenvector associated with the eigenvalue 0. The corresponding eigenspace is the line spanned by  $\mathbf{v}_1$ .

$$(A-2I)\mathbf{x} = \mathbf{0} \iff \begin{pmatrix} -1 & 1 & -1 \\ 1 & -1 & 1 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$
$$\iff \begin{pmatrix} 1 & -1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} \iff x - y + z = 0.$$

The general solution is x = t - s, y = t, z = s, where  $t, s \in \mathbb{R}$ . Equivalently,

$$\mathbf{x} = (t - s, t, s) = t(1, 1, 0) + s(-1, 0, 1).$$

Thus  $\mathbf{v}_2 = (1, 1, 0)$  and  $\mathbf{v}_3 = (-1, 0, 1)$  are eigenvectors associated with the eigenvalue 2. The corresponding eigenspace is the plane spanned by  $\mathbf{v}_2$  and  $\mathbf{v}_3$ .

Summary. 
$$A = \begin{pmatrix} 1 & 1 & -1 \\ 1 & 1 & 1 \\ 0 & 0 & 2 \end{pmatrix}$$
.

• The matrix A has two eigenvalues: 0 and 2.

• The eigenvalue 0 is *simple:* the corresponding eigenspace is a line.

• The eigenvalue 2 is of *multiplicity* 2: the corresponding eigenspace is a plane.

Eigenvectors v<sub>1</sub> = (-1, 1, 0), v<sub>2</sub> = (1, 1, 0), and v<sub>3</sub> = (-1, 0, 1) of the matrix A form a basis for ℝ<sup>3</sup>.
Geometrically, the map x → Ax is the projection on the plane Span(v<sub>2</sub>, v<sub>3</sub>) along the lines parallel to v<sub>1</sub> with the subsequent scaling by a factor of 2.

## Eigenvalues and eigenvectors of an operator

Definition. Let V be a vector space and  $L: V \rightarrow V$ be a linear operator. A number  $\lambda$  is called an **eigenvalue** of the operator L if  $L(\mathbf{v}) = \lambda \mathbf{v}$  for a nonzero vector  $\mathbf{v} \in V$ . The vector  $\mathbf{v}$  is called an **eigenvector** of L associated with the eigenvalue  $\lambda$ . (If V is a functional space then eigenvectors are also called **eigenfunctions**.)

If  $V = \mathbb{R}^n$  then the linear operator L is given by  $L(\mathbf{x}) = A\mathbf{x}$ , where A is an  $n \times n$  matrix. In this case, eigenvalues and eigenvectors of the operator L are precisely eigenvalues and eigenvectors of the matrix A.

#### **Eigenspaces**

Let  $L: V \rightarrow V$  be a linear operator.

For any  $\lambda \in \mathbb{R}$ , let  $V_{\lambda}$  denotes the set of all solutions of the equation  $L(\mathbf{x}) = \lambda \mathbf{x}$ .

Then  $V_{\lambda}$  is a *subspace* of V since  $V_{\lambda}$  is the *kernel* of a linear operator given by  $\mathbf{x} \mapsto L(\mathbf{x}) - \lambda \mathbf{x}$ .

 $V_{\lambda}$  minus the zero vector is the set of all eigenvectors of L associated with the eigenvalue  $\lambda$ . In particular,  $\lambda \in \mathbb{R}$  is an eigenvalue of L if and only if  $V_{\lambda} \neq \{\mathbf{0}\}$ .

If  $V_{\lambda} \neq \{\mathbf{0}\}$  then it is called the **eigenspace** of *L* corresponding to the eigenvalue  $\lambda$ .

Example. 
$$V=C^\infty(\mathbb{R}),\ D:V o V,\ Df=f'.$$

A function  $f \in C^{\infty}(\mathbb{R})$  is an eigenfunction of the operator D belonging to an eigenvalue  $\lambda$  if  $f'(x) = \lambda f(x)$  for all  $x \in \mathbb{R}$ .

It follows that  $f(x) = ce^{\lambda x}$ , where c is a nonzero constant.

Thus each  $\lambda \in \mathbb{R}$  is an eigenvalue of D. The corresponding eigenspace is spanned by  $e^{\lambda x}$ .

Example.  $V = C^{\infty}(\mathbb{R}), L: V \to V, Lf = f''.$  $Lf = \lambda f \iff f''(x) - \lambda f(x) = 0$  for all  $x \in \mathbb{R}$ . It follows that each  $\lambda \in \mathbb{R}$  is an eigenvalue of L and the corresponding eigenspace  $V_{\lambda}$  is two-dimensional. If  $\lambda > 0$  then  $V_{\lambda} = \text{Span}(\exp(\sqrt{\lambda}x), \exp(-\sqrt{\lambda}x))$ . If  $\lambda < 0$  then  $V_{\lambda} = \operatorname{Span}(\sin(\sqrt{-\lambda}x), \cos(\sqrt{-\lambda}x))$ . If  $\lambda = 0$  then  $V_{\lambda} = \text{Span}(1, x)$ .

Let V be a vector space and  $L: V \rightarrow V$  be a linear operator.

**Proposition 1** If  $\mathbf{v} \in V$  is an eigenvector of the operator *L* then the associated eigenvalue is unique.

*Proof:* Suppose that  $L(\mathbf{v}) = \lambda_1 \mathbf{v}$  and  $L(\mathbf{v}) = \lambda_2 \mathbf{v}$ . Then  $\lambda_1 \mathbf{v} = \lambda_2 \mathbf{v} \implies (\lambda_1 - \lambda_2) \mathbf{v} = \mathbf{0} \implies \lambda_1 - \lambda_2 = \mathbf{0} \implies \lambda_1 = \lambda_2$ .

**Proposition 2** Suppose  $\mathbf{v}_1$  and  $\mathbf{v}_2$  are eigenvectors of *L* associated with different eigenvalues  $\lambda_1$  and  $\lambda_2$ . Then  $\mathbf{v}_1$  and  $\mathbf{v}_2$  are linearly independent.

*Proof:* For any scalar  $t \neq 0$  the vector  $t\mathbf{v}_1$  is also an eigenvector of L associated with the eigenvalue  $\lambda_1$ . Since  $\lambda_2 \neq \lambda_1$ , it follows that  $\mathbf{v}_2 \neq t\mathbf{v}_1$ . That is,  $\mathbf{v}_2$  is not a scalar multiple of  $\mathbf{v}_1$ . Similarly,  $\mathbf{v}_1$  is not a scalar multiple of  $\mathbf{v}_2$ .

Let  $L: V \rightarrow V$  be a linear operator.

**Proposition 3** If  $\mathbf{v}_1$ ,  $\mathbf{v}_2$ , and  $\mathbf{v}_3$  are eigenvectors of L associated with distinct eigenvalues  $\lambda_1$ ,  $\lambda_2$ , and  $\lambda_3$ , then they are linearly independent.

*Proof:* Suppose that  $t_1\mathbf{v}_1 + t_2\mathbf{v}_2 + t_3\mathbf{v}_3 = \mathbf{0}$  for some  $t_1, t_2, t_3 \in \mathbb{R}$ . Then

$$L(t_1\mathbf{v}_1 + t_2\mathbf{v}_2 + t_3\mathbf{v}_3) = \mathbf{0},$$
  

$$t_1L(\mathbf{v}_1) + t_2L(\mathbf{v}_2) + t_3L(\mathbf{v}_3) = \mathbf{0},$$
  

$$t_1\lambda_1\mathbf{v}_1 + t_2\lambda_2\mathbf{v}_2 + t_3\lambda_3\mathbf{v}_3 = \mathbf{0}.$$

It follows that

$$t_1\lambda_1\mathbf{v}_1 + t_2\lambda_2\mathbf{v}_2 + t_3\lambda_3\mathbf{v}_3 - \lambda_3(t_1\mathbf{v}_1 + t_2\mathbf{v}_2 + t_3\mathbf{v}_3) = \mathbf{0}$$
  
$$\implies t_1(\lambda_1 - \lambda_3)\mathbf{v}_1 + t_2(\lambda_2 - \lambda_3)\mathbf{v}_2 = \mathbf{0}.$$

By the above,  $\mathbf{v}_1$  and  $\mathbf{v}_2$  are linearly independent. Hence  $t_1(\lambda_1 - \lambda_3) = t_2(\lambda_2 - \lambda_3) = 0$  $\implies t_1 = t_2 = 0 \implies t_3 = 0.$  **Theorem** If  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$  are eigenvectors of a linear operator L associated with distinct eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_k$ , then  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$  are linearly independent.

**Corollary** Let A be an  $n \times n$  matrix such that the characteristic equation  $det(A - \lambda I) = 0$  has n distinct real roots. Then  $\mathbb{R}^n$  has a basis consisting of eigenvectors of A.

*Proof:* Let  $\lambda_1, \lambda_2, \ldots, \lambda_n$  be distinct real roots of the characteristic equation. Any  $\lambda_i$  is an eigenvalue of A, hence there is an associated eigenvector  $\mathbf{v}_i$ . By the theorem, vectors  $\mathbf{v}_1, \mathbf{v}_2, \ldots, \mathbf{v}_n$  are linearly independent. Therefore they form a basis for  $\mathbb{R}^n$ .

**Theorem** If  $\lambda_1, \lambda_2, \ldots, \lambda_k$  are distinct real numbers, then the functions  $e^{\lambda_1 x}, e^{\lambda_2 x}, \ldots, e^{\lambda_k x}$  are linearly independent.

*Proof:* Consider a linear operator  $D: C^{\infty}(\mathbb{R}) \to C^{\infty}(\mathbb{R})$  given by Df = f'. Then  $e^{\lambda_1 x}, \ldots, e^{\lambda_k x}$  are eigenfunctions of Dassociated with distinct eigenvalues  $\lambda_1, \ldots, \lambda_k$ .

## Characteristic polynomial of an operator

Let *L* be a linear operator on a finite-dimensional vector space *V*. Let  $\mathbf{u}_1, \mathbf{u}_2, \ldots, \mathbf{u}_n$  be a basis for *V*. Let *A* be the matrix of *L* with respect to this basis.

*Definition.* The characteristic polynomial of the matrix A is called the **characteristic polynomial** of the operator L.

Then eigenvalues of L are roots of its characteristic polynomial.

**Theorem.** The characteristic polynomial of the operator L is well defined. That is, it does not depend on the choice of a basis.

**Theorem.** The characteristic polynomial of the operator L is well defined. That is, it does not depend on the choice of a basis.

*Proof:* Let *B* be the matrix of *L* with respect to a different basis  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ . Then  $A = UBU^{-1}$ , where *U* is the transition matrix from the basis  $\mathbf{v}_1, \dots, \mathbf{v}_n$  to  $\mathbf{u}_1, \dots, \mathbf{u}_n$ . We obtain  $\det(A - \lambda I) = \det(UBU^{-1} - \lambda I)$ 

 $= \det \left( UBU^{-1} - U(\lambda I)U^{-1} \right) = \det \left( U(B - \lambda I)U^{-1} \right)$  $= \det(U) \det(B - \lambda I) \det(U^{-1}) = \det(B - \lambda I).$