## Linear Algebra II

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## 1 Vector Spaces

## 2 Linear Transformations

## **3** Eigenvalues and Eigenvectors

**Definition 3.1 (Eigenvector).** Suppose  $T : U \to U$  is a linear transformation of U. A vector  $u \in U$  is an **eigenvector** of T with respect to  $\lambda \in \mathbb{F}$  if  $T(u) = \lambda u$ .

**Example 3.2.** Let  $U = \mathbb{R}^2$  and  $T \in L(\mathbb{R}^2, \mathbb{R}^2)$  defined by T(x, y) = (x + y, 2y). Determine some eigenvectors of T.

Note that u = 0 is an eigenvector of T with respect to any  $\lambda \in \mathbb{F}$ . Indeed,  $T(0) = 0 = \lambda 0$ , so the condition holds, and 0 is an eigenvector with respect to  $\lambda$ . Interestingly, in all other cases the scalar  $\lambda$  that works for an eigenvector u is unique.

**Theorem 3.3.** Suppose  $u \neq 0$  is an eigenvector for a linear transformation  $T: U \rightarrow U$ . Then there is only one  $\lambda \in \mathbb{F}$  such that  $T(u) = \lambda u$ .

**Definition 3.4** (Eigenvalue). A scalar  $\lambda \in \mathbb{F}$  is an eigenvalue of a linear transformation  $T: U \to U$  if  $T(u) = \lambda u$  for a nonzero eigenvector  $u \in U$ .

**Example 3.5.** Let  $U = \{f : f : \mathbb{R} \to \mathbb{R} \text{ is a differentiable function}\}$  and  $\mathbb{F} = \mathbb{R}$ . Let  $T \in L(U, U)$  defined by T(f) = f'. Determine some eigenvalues of T.

**Theorem 3.6.** Let U be an n-dimensional vector space over  $\mathbb{F}$  and  $T \in L(U,U)$ . Let M and M<sup>\*</sup> be matrices of T associate with bases  $G = \{u_1, \dots, u_n\}, G^* = \{u_1^*, \dots, u_n^*\}$  for U, respectively. Then for each  $\lambda \in \mathbb{F}$ ,  $|M - \lambda I_n| = |M^* - \lambda I_n|$ , where  $I_n$  is an n-by-n identity matrix.

**Definition 3.7** (Characteristic polynomial). Let U be an n-dimensional vector space over  $\mathbb{F}$  and  $T \in L(U, U)$ . Let M be a matrix of T associate with any basis for U. The determinant  $\Delta(t) := |M - tI_n| = (-1)^n t^n + b_{n-1} t^{n-1} + \cdots + b_1 t + b_0$  is called the **characteristic polynomial** of T.

**Example 3.8.** Let  $U = \mathbb{R}^2$  and  $T \in L(\mathbb{R}^2, \mathbb{R}^2)$  defined by T(x, y) = (y, -x). Find characteristic polynomial of T.

**Theorem 3.9.** Let U be an n-dimensional vector space over  $\mathbb{F}$ ,  $T \in L(U, U)$ and  $\Delta(t)$  be the characteristic polynomial of T. Then  $\lambda$  is an eigenvalue of T if and only if  $\lambda$  is a root of  $\Delta(t)$ , that is,  $\Delta(\lambda) = 0$ . **Example 3.10.** In Example 3.8,  $\Delta(t) = t^2 + 1$ .  $\Delta(t) \neq 0 \ \forall t \in \mathbb{R}$ . Therefore, T has no eigenvalues.

**Example 3.11.** Let  $U = \mathbb{R}^2$  and  $T \in L(\mathbb{R}^2, \mathbb{R}^2)$  defined by T(x, y) = (x + y, 2y). Find eigenvalues and eigenvectors of T.

**Example 3.12.** Let  $U = \mathbb{C}^2$ ,  $\mathbb{F} = \mathbb{C}$  and  $T \in L(\mathbb{C}^2, \mathbb{C}^2)$  defined by T(x, y) = (y, -x). Find eigenvalues and eigenvectors of T.

**Example 3.13.** Let  $U = \mathbb{C}^2$ ,  $\mathbb{F} = \mathbb{R}$  and  $T \in L(\mathbb{C}^2, \mathbb{C}^2)$  defined by T(x, y) = (y, -x). Find eigenvalues and eigenvectors of T.

**Remark 3.14.** Let U be an n-dimensional vector space over  $\mathbb{F}$  and  $T \in L(U, U)$ . Then T has at most n distinct eigenvalues.

**Definition 3.15** (Eigenspace). Suppose  $\lambda \in \mathbb{F}$ . The eigenspace of  $T : U \to U$  corresponding to  $\lambda$  is the following set:

$$U_{\lambda} = \{ u \in U : T(u) = \lambda u \}.$$

**Theorem 3.16.** The subset  $U_{\lambda}$  is a subspace of U for all  $\lambda \in \mathbb{F}$ . Furthermore,  $U_{\lambda} = \ker(T - \lambda I)$ , where I is the identity transformation.

We note that  $U_{\lambda} \neq 0$  if and only if  $\lambda$  is an eigenvalue of T. This leads to a practical method of computing eigenvalues and eigenvectors of T. Note that  $\lambda$  is an eigenvalue of T if and only if  $U_{\lambda} \neq 0$ , that is, if  $U_{\lambda}$  contains nonzero vectors. Hence we have the following result.

**Theorem 3.17.** A scalar  $\lambda \in \mathbb{F}$  is an eigenvalue of T if and only if ker $(T - \lambda I) \neq 0$ .

**Definition 3.18 (Geometric multiplicity).** Let U be vector space over  $\mathbb{F}$ ,  $T \in L(U, U)$  and  $\lambda$  be an eigenvalue of T. The dimension of  $U_{\lambda}$  is called the *geometric multiplicity* of  $\lambda$  and denoted by  $GM(\lambda)$ .

**Definition 3.19** (Algebraic multiplicity). Let U be an n-dimensional vector space over  $\mathbb{F}$ ,  $T \in L(U, U)$  and  $\lambda$  is an eigenvalue of T. The algebraic multiplicity of  $\lambda$  is defined to be its multiplicity as a root of the characteristic polynomial of T, and denoted by  $AM(\lambda)$ .

**Remark 3.20.**  $AM(\lambda) = n$  if and only if  $\Delta(t) = (t - \lambda)^n g(t)$ , where g(t) is a polynomial of t and  $g(\lambda) \neq 0$ .

**Example 3.21.** Let  $U = \mathbb{R}^2$  and  $T \in L(\mathbb{R}^2, \mathbb{R}^2)$  defined by T(x, y) = (x, x + y). Find algebraic multiplicity and geometric multiplicity of each eigenvalue of T.

**Theorem 3.22.** Let U be an n-dimensional vector space over  $\mathbb{F}$ ,  $T \in L(U, U)$ and  $\lambda$  is an eigenvalue of T. Then  $GM(\lambda) \leq AM(\lambda)$ .

**Theorem 3.23.** Let U be an n-dimensional vector space over  $\mathbb{F}$ ,  $T \in L(U, U)$ and  $\lambda$  is an eigenvalue of T. If  $AM(\lambda) = 1$ , then  $GM(\lambda) = 1$ .

**Theorem 3.24.** Let U be a vector space over  $\mathbb{F}$  and  $T \in L(U,U)$ . Let  $u_1, \dots, u_m$  be eigenvectors of T associate with distinct eigenvalues  $\lambda_1, \dots, \lambda_m$ . Then  $\{u_1, \dots, u_m\}$  is linearly independent.

**Theorem 3.25.** Let U be a vector space over  $\mathbb{F}$  and  $T \in L(U, U)$ . Let  $\lambda$  and  $\sigma$  be distinct eigenvalues of T. Then  $U_{\lambda} \cap U_{\sigma} = \{0\}$ .

**Definition 3.26.** Let U be a finite dimensional vector space over  $\mathbb{F}$  and  $T \in L(U, U)$ . A basis S of U **diagonalizes** T if the matrix of T with respect to S is a diagonal matrix.

**Definition 3.27** (Diagonalizable). Let U be a vector space over  $\mathbb{F}$  and  $T \in L(U, U)$ . If there exist a basis S of U which diagonalizes T, then T is said to be **diagonalizable**.

**Example 3.28.** Let  $U = \mathbb{R}^2$  and  $T \in L(\mathbb{R}^2, \mathbb{R}^2)$  defined by T(x, y) = (y, x). Let  $S_1 = \{(1,0), (0,1)\}, S_2 = \{(0,1), (1,0)\}, S_3 = \{(1,1), (2,3)\}$  and  $S_4 = \{(1,1), (1,-1)\}$ . Which of these bases diagonalizes T. **Remark 3.29.** Let U be an n-dimensional vector space over  $\mathbb{F}$  and  $T \in L(U, U)$ . T is diagonalizable if and only if

- 1. The characteristic polynomial of T,  $\Delta(t)$ , must be of the form  $\Delta(t) = \alpha(t \lambda_1)^{r_1} \cdots (t \lambda_k)^{r_k}$ , where  $r_1 + \cdots + r_k = n$ ,  $\lambda_i \in \mathbb{F}$ ,  $\lambda_i \neq \lambda_j$ whenever  $i \neq j$ , and  $\alpha \in \mathbb{F} - \{0\}$ .
- 2. For each eigenvalue  $\lambda_i$  of T,  $AM(\lambda_i) = GM(\lambda_i)$  for  $i = 1, \dots, k$ .

**Example 3.30.** In Example 3.11, T is a diagnalizable since it satisfies the conditions of Remark 3.29.

**Example 3.31.** In Example 3.21, T is not diagnalizable since  $AM(1) \neq GM(1)$ .

**Example 3.32.** In Example 3.8, T is not diagnalizable since  $\Delta(t) = t^2 + 1 \neq 0 \forall t \in \mathbb{R}$ .

**Remark 3.33.** Let U be an n-dimensional vector space over  $\mathbb{F}$  and  $T \in L(U,U)$ . Then T is diagonalizable if and only if there exist a basis for U consisting of eigenvectors of T.

**Theorem 3.34.** Let U be an n-dimensional vector space over  $\mathbb{F}$  and  $T \in L(U, U)$ . If T has n distinct eigenvalues, then T is diagonalizable.

**Example 3.35.** Let  $U = \mathbb{R}^2$  and  $T \in L(\mathbb{R}^2, \mathbb{R}^2)$  defined by T(x, y) = (y, x). Find a basis for  $\mathbb{R}^2$  which diagonalizes T.