### 2. Eigenvalues and Eigenvectors Chap. 5 of Lay Section 5.1 to 5.6

- (a) Definition of eigenvector and eigenvalue (5.1)
- i) for an  $n \times n$  matrix
- ii) for a linear transformation (an operator) Examples:

 $A = I_n$ 

L = projection in  $\mathbb{R}^2$  onto line  $x_1 = x_2$ A' = orthogonal matrix, e.g.,  $A' = \begin{bmatrix} \frac{1}{2}\sqrt{3} & -\frac{1}{2} \\ \frac{1}{2} & \frac{1}{2}\sqrt{3} \end{bmatrix} = \begin{bmatrix} \cos 30^\circ & -\sin 30^\circ \\ \sin 30^\circ & \cos 30^\circ \end{bmatrix}$ 

 $L' = \text{rotation by } \varphi \text{ in } \mathbb{R}^6 \text{ (Example 3, p. 84)}.$ 

For an orthogonal matrix, the only eigenvalues are  $\pm 1$ .

See examples of Linear transformations in Lay, Table 1, 2, pages 85, 86.

(b) Calculating Eigenvalues (5.1, 5.2)

*Prop:* A scalar  $\lambda$  is an eigenvalue of A iff  $det(A = \lambda I) = 0$  (p. 313)  $c_A(\lambda) \stackrel{def}{=} det(A - \lambda I) = \text{characteristic polynomial.}$ 

In each of the following matrices find the characteristic polynomial and its roots:

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \qquad C = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

$$B = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 2 & 3 \\ 0 & 0 & 3 \end{bmatrix}$$

Def.:  $n \times n$  matrix B is similar to matrix A if there exists an invertible matrix P such that  $B = P^{-1}AP$ .

Prove that similar matrices have the same characteristic polynomial.

Q. Suppose  $\lambda$  is an eigenvalue of A; is it an eigenvalue of  $A^3$ ? What is true?

Theorem 1 (p. 307): The eigenvalues of a triangular matrix are the entries on its main diagonal.

Proposition (p. 306): 0 is an eigenvalue of A iff A has no inverse.

#### (c) Eigenvectors and Eigenspaces

Find the eigenvalues and eigenvectors of the following matrices.

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \qquad B = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

Def: The algebraic multiplicity  $mult_a(\lambda_i)$  of eigenvalue  $\lambda_i$  if A is the multiplicity of  $\lambda_i$  as a root of the characteristic polynimial  $c_A(\lambda)$ .

The geometric multiplicity  $\operatorname{mult}_g(\lambda_i)$  of eigenvalue  $\lambda_i$  of A is the dimension of the eigenspace  $E_{\lambda_i}$  corresponding to  $\lambda_i$ .

The eigenspace corresponding to  $\lambda_i$  is the subspace  $E_{\lambda_i}$  of  $\mathbb{R}^n$  consisting of the zero vector and all eigenvectors corresponding to  $\lambda_i$ .

Multiplicity Theorem (Theorem 7a, p. 324 Lay): For any eigenvalue  $\lambda_i$  of  $n \times n$  matrix A mult $_a(\lambda_i) \le \text{mult}_a(\lambda_i)$ . Prove this if you have time.

## (d) Diagonalization (5.3)

Def: An  $n \times n$  matrix is diagonalizable if there is an invertible matrix P such that  $P^{-1}AP = D$ , a diagonal matrix

Example: (p. 321)
$$A = \begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & -3 \\ 3 & 3 & 1 \end{bmatrix} \qquad c_A(\lambda) = -(\lambda - 1)(\lambda + 2)^2$$

$$E_1 = \operatorname{span} \left\{ \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \right\} \qquad E_2 = \operatorname{span} \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

Thm. 5 (p. 321): An  $n \times n$  matrix is diagonalizable iff A has n linearly independent eigenvectors.

*Problem:* Is 
$$A = \begin{bmatrix} 4 & -2 \\ 1 & 1 \end{bmatrix}$$
 similar to  $B = \begin{bmatrix} -3 & 10 \\ -3 & 8 \end{bmatrix}$ ?

**Answer**: Yes. Each can be diagonalized to  $\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$ . Other possible examples:

$$B = \left[ \begin{array}{ccc} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{array} \right] \qquad C = \left[ \begin{array}{ccc} 0 & -1 \\ 1 & 0 \end{array} \right].$$

#### (e) Diagonalization

Eigenvectors belonging to distinct eigenvalues are linearly independent:

Theorem 2 (p. 307): If  $v_1, \ldots, v_r$  are eigenvectors corresponding to distinct eigenvalues  $\lambda_1, \ldots, \lambda_r$  in  $n \times n$  matrix A, then  $\{v_1, \ldots, v_4\}$  is linearly independent.

#### **Proof:**

Theorem 6 (p. 323). An  $n \times n$  matrix with n distinct eigenvalues is diagonalizable.

More generally

Multiplicity Theorem: (Theorem 7a, p. 324). For each eigenvalue  $\lambda_v$  of  $n \times n$  matrix A,

$$mult_g(\lambda_v) \leq mult_a(\lambda_v).$$

Diagonalization Theorem. A square matrix A is diagonalizable iff for each eigenvalue  $\lambda_v$  of  $n \times n$  matrix A,

$$mult_g(\lambda_v) = mult_a(\lambda_v).$$

# (f) Complex eigenvalues

Examples:

$$A = \left[ \begin{array}{cc} 0 & -1 \\ 1 & 0 \end{array} \right].$$

 $L_A = \text{rotation by } 90^{\circ} \text{ clockwise.}$ 

Theorem: An  $n \times n$  matrix over  $\mathbb{R}$  or  $\mathbb{C}$  has n eigenvalues if multiplicity is counted.

Examples:

$$A = \begin{bmatrix} 3 & -5 \\ 1 & -1 \end{bmatrix}$$

$$c_A(\lambda) = \lambda^2 - 2\lambda + 2$$

$$\lambda = 1 \pm i$$

Study the dynamical system:

$$x(t+1) = \begin{bmatrix} 3 & -5 \\ 1 & -1 \end{bmatrix} x(t), \quad x_0, = \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$