

1. Let  $A \in \mathbf{C}^{n \times n}$ .

- (a) A scalar  $\lambda$  is an *eigenvalue* of an  $n \times n$   $A$  and a nonzero vector  $x \in \mathbf{C}^n$  is a corresponding *(right) eigenvector* if

$$Ax = \lambda x.$$

- (b) A nonzero vector  $y$  is called a *left eigenvector* if

$$y^H A = \lambda y^H.$$

- (c) The set of all eigenvalues of  $A$ , denoted as  $\lambda(A)$ , is called the *spectrum* of  $A$ .

- (d) The *characteristic polynomial* of  $A$  is a polynomial of degree  $n$ , and defined as

$$p(\lambda) = \det(\lambda I - A).$$

2. The following is a list of properties straightforwardly from above definitions:

- (a)  $\lambda$  is  $A$ 's eigenvalue  $\Leftrightarrow \lambda I - A$  is singular  $\Leftrightarrow \det(\lambda I - A) = 0 \Leftrightarrow p(\lambda) = 0$ .
- (b) There is at least one eigenvector  $x$  associated with  $A$ 's eigenvalue  $\lambda$ .
- (c) Suppose  $A$  is real.  $\lambda$  is  $A$ 's eigenvalue  $\Leftrightarrow$  conjugate  $\bar{\lambda}$  is also  $A$ 's eigenvalue.
- (d)  $A$  is singular  $\Leftrightarrow 0$  is  $A$ 's eigenvalue.
- (e) If  $A$  is upper (or lower) triangular, then its eigenvalues consist of its diagonal entries.  
(*Question: what if  $A$  is a block upper (or lower) triangular matrix ?*).

3. Schur decomposition.

Let  $A$  be of order  $n$ . Then there is an  $n \times n$  unitary matrix  $U$  (i.e.,  $U^H U = I$ ) such that

$$A = U T U^H,$$

where  $T$  is upper triangular and the diagonal elements of  $T$  are the eigenvalues of  $A$ .

4. When  $A$  is Hermitian, i.e.,  $A^H = A$ , then by Schur decomposition, we know that there exist an unitary matrix  $U$  such that

$$A = U \Lambda U^H,$$

where  $\Lambda = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$  and all eigenvalues  $\lambda_i$  are real.

5.  $A \in \mathbf{C}^{n \times n}$  is *simple* if it has  $n$  linearly independent eigenvectors; otherwise it is *defective*.  
Examples.

- (a)  $I$  and any diagonal matrices is simple.  $e_1, e_2, \dots, e_n$  are  $n$  linearly independent eigenvectors.
- (b)  $A = \begin{pmatrix} 1 & 2 \\ 4 & 3 \end{pmatrix}$  is simple. It has two different eigenvalues  $-1$  and  $5$ , it has 2 linearly independent eigenvectors:  $\frac{1}{\sqrt{2}} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$  and  $\frac{1}{\sqrt{5}} \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ .

(c) If  $A \in \mathbf{C}^{n \times n}$  has  $n$  different eigenvalues, then  $A$  is simple.

(d)  $A = \begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix}$  is defective. It has two repeated eigenvalues 2, but only one eigenvector  $e_1 = (1, 0)^T$ .

## 6. Eigenvalue decomposition

$A \in \mathbf{C}^{n \times n}$  is simple if and only if there exists a nonsingular matrix  $X \in \mathbf{C}^{n \times n}$  such that

$$A = X\Lambda X^{-1},$$

where  $\Lambda = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$ . In this case,  $\lambda_i$  are eigenvalues, and the columns of  $X$  are eigenvectors, and  $A$  is called *diagonalizable*.

7. An *invariant subspace* of  $A$  is a subspace  $\mathcal{V}$  of  $\mathbf{R}^n$ , with the property that

$$v \in \mathcal{V} \text{ implies that } Av \in \mathcal{V}.$$

We also write this as  $A\mathcal{V} \subseteq \mathcal{V}$ .

Examples.

(a) The simplest, one-dimensional invariant subspace is the set  $\text{span}(x)$  of all scalar multiples of an eigenvector  $x$ .

(b) Let  $x_1, x_2, \dots, x_m$  be any set of independent eigenvectors with eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_m$ . Then  $\mathcal{X} = \text{span}(\{x_1, x_2, \dots, x_m\})$  is an invariant subspace.

PROPOSITION. Let  $A$  be  $n$ -by- $n$ , let  $V = [v_1, v_2, \dots, v_m]$  be any  $n$ -by- $m$  matrix with linearly independent columns, and let  $\mathcal{V} = \text{span}(V)$ , the  $m$ -dimensional space spanned by the columns of  $V$ . Then  $\mathcal{V}$  is an invariant subspace if and only if there is an  $m$ -by- $m$  matrix  $B$  such that

$$AV = VB.$$

In this case, the  $m$  eigenvalues of  $B$  are also eigenvalues of  $A$ .

8. Two  $n \times n$  matrices  $A$  and  $B$  are *similar* if there is an  $n \times n$  non-singular matrix  $P$  such that  $B = P^{-1}AP$ . We also say  $A$  is *similar* to  $B$ , and likewise  $B$  is similar to  $A$ ;  $P$  is a *similarity transformation*.  $A$  is *unitarily similar* to  $B$  if  $P$  is unitary.

PROPOSITION. Suppose that  $A$  and  $B$  are similar:  $B = P^{-1}AP$ .

(a)  $A$  and  $B$  have the same eigenvalues. In fact  $p_A(\lambda) \equiv p_B(\lambda)$ .

(b)  $Ax = \lambda x \Rightarrow B(P^{-1}x) = \lambda(P^{-1}x)$ .

(c)  $Bw = \lambda w \Rightarrow A(Pw) = \lambda(Pw)$ .