

## 8.3 DIAGONALIZATION

Suppose  $A_{n \times n}$  has  $n$  linearly independent eigenvectors. If these eigenvectors are the columns of a matrix  $S$ , then  $A$  has a factorization such that  $S^{-1}AS = \Lambda$  is a diagonal matrix where the eigenvalues of  $A$  are on the diagonal. We can see this in the following calculation,

$$\begin{aligned} AS &= A \begin{bmatrix} | & | & & | \\ x_1 & x_2 & \cdots & x_n \\ | & | & & | \end{bmatrix} = \begin{bmatrix} | & | & & | \\ Ax_1 & Ax_2 & \cdots & Ax_n \\ | & | & & | \end{bmatrix} = \begin{bmatrix} | & | & & | \\ \lambda_1 x_1 & \lambda_2 x_2 & \cdots & \lambda_n x_n \\ | & | & & | \end{bmatrix} \\ &= \begin{bmatrix} | & | & & | \\ x_1 & x_2 & \cdots & x_n \\ | & | & & | \end{bmatrix} \begin{bmatrix} \lambda_1 & & & 0 \\ & \lambda_2 & & \\ & & \ddots & \\ 0 & & & \lambda_n \end{bmatrix} = S\Lambda \Rightarrow A = S\Lambda S^{-1} \end{aligned}$$

**A matrix is not diagonalizable if the eigenvectors are not linearly independent.**

Caveat: there is no direct connection between this and invertibility. Diagonalizability depends on the linear independence of eigenvectors and invertibility depends on the linear independence of the columns of a matrix.

Now lets look at some examples.

Ex: Find  $S$  and  $\Lambda$  in the  $A = SAS^{-1}$  factorization of

$$A = \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}$$

**Solution:** First we find the eigenvalues

$$\begin{vmatrix} 1/2 - \lambda & 1/2 \\ 1/2 & 1/2 - \lambda \end{vmatrix} = \frac{1}{4} - \lambda + \lambda^2 - \frac{1}{4} = \lambda(\lambda - 1) = 0 \Rightarrow \lambda = 0, 1$$

Next we find the eigenvectors

$$\begin{pmatrix} 1/2 - \lambda & 1/2 \\ 1/2 & 1/2 - \lambda \end{pmatrix} x = 0 \Rightarrow x = \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Then

$$\Lambda = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \quad S = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$

Ex: Find  $S$  and  $\Lambda$  in the  $K = SAS^{-1}$  factorization of

$$K = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

First we find the eigenvalues,

$$\begin{vmatrix} -\lambda & -1 \\ 1 & -\lambda \end{vmatrix} = \lambda^2 + 1 = 0 \Rightarrow \lambda = \pm i.$$

Next we find the eigenvectors

$$\begin{pmatrix} -\lambda & -1 \\ 1 & -\lambda \end{pmatrix} x = 0 \Rightarrow x = \begin{pmatrix} 1 \\ -i \end{pmatrix}, \begin{pmatrix} 1 \\ i \end{pmatrix}$$

Then

$$\Lambda = \begin{bmatrix} i & 0 \\ 0 & -1 \end{bmatrix}, \quad S = \begin{bmatrix} 1 & 1 \\ -i & i \end{bmatrix}$$

Lets look at a couple of interesting results that may be useful to us.

**Theorem 1.** The eigenvalues of  $A^2$  is  $\lambda_1^2, \lambda_2^2, \dots, \lambda_n^2$  if the eigenvalues of  $A$  are  $\lambda_1, \lambda_2, \dots, \lambda_n$ .

*Proof.*

$$A^2x = A(Ax) = A(\lambda x) = \lambda Ax = \lambda^2 x. \quad \square$$

*Alternate proof.*

$$A = SAS^{-1} \Rightarrow A^2 = (S\Lambda S^{-1})(S\Lambda S^{-1}) = S\Lambda S^{-1}S\Lambda S^{-1} = S\Lambda\Lambda S^{-1} = S\Lambda^2 S^{-1}. \quad \square$$

We can extend this to higher powers and inverses; i.e.,  $A^k = S\Lambda^k S^{-1}$  and  $A^{-1} = S\Lambda^{-1} S^{-1}$  if  $\lambda_1, \dots, \lambda_n \neq 0$ .