

# ST 661 Note

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## Outline

Orthogonal matrices (Cont.)

Trace

Eigenvalues and eigenvector

- general
- functions of matrices
- symmetric
- pd

## Definition

$$C \text{ is orthogonal} \iff C^T C = I \iff C C^T = I \iff C^{-1} = C^T$$

## Thm

If  $C$  is a  $p \times p$  orthogonal and  $A$  is  $p \times p$ ,

i)  $\det(C) = \pm 1$

This is because  $C^T C = I$ ,  $\det(C^T C) = \det(C^T) \det(C) = \det(C) \det(C) = 1$ .

ii)  $\det(C^T A C) = \det(A)$ ,

where  $C^T A C$  is called conjugate transformation of  $A$ .

proof.

$$\det(C^T A C) = \det(A C C^T) = \det(A C C^T) = \det(A I) = \det(A) \quad \square$$

iii)  $-1 \leq C_{ij} \leq 1$  for all  $i, j$  since they were normalized.

## Trace (For Square Matrices Only)

For  $n \times n$  matrix  $A = (a_{ij})$ ,  $1 \leq i, j \leq n$ , trace is a linear scalar function of  $A$  s.t.  $\text{tr}(A) = \sum_{i=1}^n a_{ii}$ .

## Thm

i) If A, B are both  $n \times n$ , then

$$\text{tr}(A + B) = \text{tr}(A) + \text{tr}(B).$$

ii) For A:  $n \times p$  and B:  $p \times n$ ,

$$\text{tr}(AB) = \text{tr}(BA).$$

Proof. Exercise. Use definition of trace.

Example. If C is orthogonal, is  $\text{tr}(C^T AC) = \text{tr}(A)$ ?

Proof. Yes, and the proof is similar to  $\det(A^T AC) = \det(A)$ . Here, we have

$$\text{tr}(C^T AC) = \text{tr}(ACC^T) = \text{tr}(AI) = \text{tr}(A).$$

This is called the conjugate transformation. There are other transformations. Suppose P is nonsingular, then the similarity transformation is given as  $P^{-1}AP$  and

$$\text{tr}(P^{-1}AP) = \text{tr}(A).$$

## Taking A Step Back

Saying A is  $n \times p$  matrix is same as saying that  $A \in \mathbb{R}^n$ . A can also be thought as a linear transformation. If

$$A_{n \times p} X_{p \times 1} = nx1,$$

then  $A : \mathbb{R}^p \rightarrow \mathbb{R}^n$  for  $x \rightarrow AX$ .

If A is  $n \times n$ , then  $A : \mathbb{R}^n \rightarrow \mathbb{R}^n$  for  $x \rightarrow AX$ .

Next, we want to ask

1. For what A, we have  $\|x\|$  (length of x) =  $\|Ax\|$  for all  $x \in \mathbb{R}^n$  (i.e. rotation)?

To make

$$\|x\| = \sqrt{x^T x} = \sqrt{(Ax)^T (Ax)} = \sqrt{x^T A^T A x},$$

we need

$$x^T x = x^T A^T A x.$$

Or more like we need

$$x^T I x = x^T A^T A x \Leftrightarrow x^T (A^T A - I) x = 0 \quad \forall x$$

$$\Rightarrow A^T A = I$$

$\Leftrightarrow$  A has to be orthogonal (and A is a rotation matrix). I.e., orthogonal matrices preserve length.

2. For what A, do we have  $x$  and  $Ax$  in the same direction  $\forall x$ ?

$x$  and  $ax$  are in the direction  $\forall a$ . For scalar  $a$ , we need

$$Ax = ax$$

Then

$$A = aI,$$

if  $a$  is the same since

$$(A - aI)x = 0 \quad \forall x.$$

Or more flexible  $A$  can be

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & \ddots \end{bmatrix}.$$

Relaxing it a bit, we can ask

3. For a given  $A$ , what  $x$  guarantee that (nontrivial)  $x$  and  $Ax$  are in the same direction?

## Eigenvalues

$A$  is  $n \times n$ . A scalar  $\lambda$  is called an eigenvalue of  $A$  if  $\exists x$  s.t.

$$Ax = \lambda x.$$

Such a vector  $x$  is called an eigenvector of  $A$ .

To find  $\lambda$  and  $x$ , set

$$(A - \lambda I)x = 0.$$

Since  $(A - \lambda I)$  has to be singular (???)  $\Leftrightarrow \det(A - \lambda I) = 0$ . This is a degree  $n$  polynomial about  $\lambda$ . In addition, root of degree  $n$  polynomial has  $n$  items and exists, so  $\lambda$  always exists.

## Characteristic Equation

$(A - \lambda I)x = 0$  is called the characteristic (or polynomial) equation of  $A$ .  $\lambda$  is an eigenvalue of  $A$  and  $x$  is a corresponding eigenvector.

$$Ax = \lambda x$$

Claim.  $2x$  is also eigenvector of  $A$  corresponding to  $\lambda$ .

Proof.

$$A(2x) = 2Ax = 2\lambda x = \lambda(2x)$$

In general,  $cx$  is also an eigenvector of  $A$ , where  $c$  is a scalar and  $\neq 0$ . Among this class of eigenvectors, we often require  $\|x^T x\| = 1$  for eigenvector.

Ques. Suppose  $\lambda$  is an eigenvalue of  $A$  with corresponding eigenvector  $x$ , when does  $g(A)$  has the same  $\lambda$ ?

- i)  $g(A) = cA$  where  $c$  is a scalar

Proof.

$$Ax = \lambda x \Rightarrow (cA)x = c\lambda x,$$

so  $c\lambda$  is an eigenvalue of  $cA$ .

ii)  $g(A) = cA + bI$

$$g(A)x = (cA + bI)x = cAx + bIx = cAx + bx = c\lambda x + bx = (c\lambda + b)x,$$

so  $(c\lambda + b)$  is an eigenvalue of  $g(A) = cA + bI$ .

iii) Suppose  $g(A) = A^2$ , is  $\lambda^2$  an eigenvalue of  $A^2$ ? Yes.

Proof.

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

In general,  $A^k$  has  $\lambda^k$  as an eigenvalue.

iv) Combining ii) and iii), we get that

$$(A^3 + 4A^2 - 3A + 3I)x = A^3x + 4A^2x - 3Ax + 3x = \lambda^3x + 4\lambda^2x - 3\lambda x + 3x = (\lambda^3 + 4\lambda^2 - 3\lambda + 3)x.$$

v) Suppose  $A$  is nonsingular (has inverse),  $g(A) = A^{-1}$  has  $\frac{1}{\lambda}$  as an eigenvalue.

Claim. If  $A$  is nonsingular, then  $\lambda \neq 0$ .

Proof. If  $\lambda = 0$ , then  $\exists x \neq 0$  s.t.  $Ax = 0$   $x = 0$ .

## Thm

$A$  is  $n \times n$ ,

- i) If  $P$  is nonsingular, then  $P^{-1}AP$  and  $A$  share the same eigenvalue.
- ii) If  $C$  is orthogonal, then  $C^TAC$  and  $A$  share the same eigenvalue.

Proof. HW. Remark. Eigenvectors may be different.

## (!Important) Thm

$A$  is  $n \times n$  symmetric (i.e.  $A = A^T$  or  $a_{ij} = a_{ji}$ ), then

- i) The eigenvalue  $\lambda_1, \lambda_2, \dots, \lambda_n$  are real (some of them have the same eigenvector).
- ii) The eigenvector  $x_1, x_2, \dots, x_k$  of  $A$  corresponding to the distinct eigenvalue  $\lambda_1, \lambda_2, \dots, \lambda_k$  are mutually orthogonal and the eigenvector  $x_{k+1}, x_{k+2}, \dots, x_{k+n}$  corresponding to the nondistinct eigenvalue can be chosen to be mutually orthogonal, so  $x_i^T x_j = 0 \forall i \neq j$ .

Ques. Why though?  $C$  is  $n \times n$

$$C = \begin{bmatrix} c_1 & c_2 & \cdots & c_n \end{bmatrix} = \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix} = \begin{bmatrix} | & | & \cdots & | \end{bmatrix}$$

$C$  is an orthogonal matrix because using this theorem, there is a way to make sure we have orthogonal vectors.

$$AC = A(x_1, x_2, \dots, x_n) = (Ax_1, Ax_2, \dots, Ax_n) = (\lambda_1 x_1, \lambda_2 x_2, \dots, \lambda_n x_n) = C \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix} = (or \ CD),$$

where  $x_i$  are eigenvectors.

$AC = CD$  where  $C$  is an orthogonal matrix.

$\Rightarrow A = CDC^T$  for any symmetric  $A$  (This is called spectral decomposition.)

$\Leftrightarrow C^T AC = D$  This is called the diagonalization of matrix.