

Lecture 18: Orthogonal Matrix & Diagonalization of real symmetric matrices

We have seen that if a square matrix has distinct eigen values then it is diagonalizable. Here, we will see that a real symmetric matrix is diagonalizable. First we prove the following lemma

Lemma: Let A be an $n \times n$ matrix.

Then the following conditions are equivalent.

1. The column vectors are orthonormal.
2. $A^T A = A A^T = I_n$ (so $A^{-1} = A^T$)
3. $\|A x\| = \|x\| \quad \forall x \in \mathbb{R}^n$
4. $\langle A x, A y \rangle = \langle x, y \rangle \quad \forall x, y \in \mathbb{R}^n$
5. The row vectors are orthonormal.

Proof:- Let $A = (a_{ij}) \quad 1 \leq i \leq n, 1 \leq j \leq n.$

Column vectors $v_j = (a_{1j}, a_{2j}, \dots, a_{nj})$

& row vectors $w_i = (a_{i1}, a_{i2}, \dots, a_{in})$

$$1 \Rightarrow 2: \because \langle v_p, v_q \rangle = 0 \quad p \neq q$$

$$\& \|v_p\| = 1$$

$$\text{Hence } \sum_{i=1}^n a_{ip} a_{iq} = 0 \quad \text{for } p \neq q$$

$$\& \sum_{k=1}^n a_{kp}^2 = 1$$

p th row vector of A^T is $(a_{1p}, a_{2p}, \dots, a_{np})$

$\&$ q th column vector of A is $(a_{1q}, a_{2q}, \dots, a_{nq})$

pq th entry of $A^T A$ is dot product of $(a_{1p}, a_{2p}, \dots, a_{np})$ with $(a_{1q}, a_{2q}, \dots, a_{nq})$

$$\text{So, } A^T A = I_n$$

$$\Rightarrow A^T \text{ is invertible } \& (A^T)^{-1} = A$$

$$\& A A^T = I_n, \text{ So } A A^T = A^T A = I_n.$$

$$(2) \Rightarrow (3) \therefore \text{Let } e_i = (0, \dots, 1, \dots, 0)$$

¹
ith position

$\{e_1, e_2, \dots, e_n\}$ is a basis of \mathbb{R}^n

Let $x = (x_1, \dots, x_n) \in \mathbb{R}^n$ then

$$x = (x_1, \dots, x_n) = \sum_{i=1}^n x_i e_i$$

$$A x = \sum_{i=1}^n x_i A e_i$$

$A e_i$ is the i th column vector

$(a_{1i}, a_{2i}, \dots, a_{ni})$ of A

$$\|A e_i\| = \|(a_{1i}, a_{2i}, \dots, a_{ni})\| = \sqrt{a_{1i}^2 + \dots + a_{ni}^2}$$

$$= \sqrt{1} \text{ (as } i\text{th element} \\ = 1 \text{ if } A^T A \text{ is } I)$$

$$\begin{aligned}
\|Ax\|^2 &= \langle Ax, Ax \rangle \\
&= \left\langle \sum_{i=1}^n x_i A e_i, \sum_{i=1}^n x_i A e_i \right\rangle \\
&= \sum_{i,j=1}^n x_i x_j \langle A e_i, A e_j \rangle \\
&= \sum_{i=1}^n x_i^2 \quad \left(\text{since for } i \neq j \right. \\
&\quad \left. \langle A e_i, A e_j \rangle = \text{entry of } A^T A \text{ which is 0 as } A^T A = I \right) \\
&= \|x\|^2
\end{aligned}$$

$$\Rightarrow \|Ax\| = \|x\| \quad \forall x \in \mathbb{R}^n.$$

$$(3) \Rightarrow (4) \quad \|A(x+y)\| = \|x+y\|$$

$$\Rightarrow \langle A(x+y), A(x+y) \rangle = \langle x+y, x+y \rangle$$

$$\begin{aligned}
\Rightarrow \langle Ax + Ay, Ax + Ay \rangle &= \langle x, x \rangle \\
&\quad + \langle x, y \rangle + \langle y, x \rangle \\
&\quad + \langle y, y \rangle
\end{aligned}$$

$$\begin{aligned}
\Rightarrow \|Ax\|^2 + 2\langle Ax, Ay \rangle + \|Ay\|^2 \\
= \|x\|^2 + 2\langle x, y \rangle + \|y\|^2
\end{aligned}$$

$$\begin{aligned}
\Rightarrow \langle Ax, Ay \rangle &= \langle x, y \rangle \quad \text{as} \\
&\quad \|Ax\| = \|x\| \\
&\quad \& \|Ay\| = \|y\|.
\end{aligned}$$

(4) \Rightarrow (5) Suppose $\langle Ax, Ay \rangle = \langle x, y \rangle \forall x, y \in \mathbb{R}^n$
 Consider the orthonormal basis $\{e_1, \dots, e_n\}$
 of \mathbb{R}^n then for $i \neq j$ $\langle Ae_i, Ae_j \rangle = \langle e_i, e_j \rangle$
 $= 0$

$$\& \langle Ae_i, Ae_i \rangle = \langle e_i, e_i \rangle = 1$$

Let $A = (a_{ij})$ then Ae_i is the i th
 column vector $\begin{pmatrix} a_{1i} \\ a_{2i} \\ \vdots \\ a_{ni} \end{pmatrix}$ of A .

or i th row vector of A^T .

$\langle Ae_i, Ae_j \rangle = 0$ for $i \neq j$, $\langle Ae_i, Ae_i \rangle = 1$
 implies $A^T A = I \Rightarrow A A^T = I$

\Rightarrow row vectors of
 A are orthonormal

(5) \Rightarrow (i) Suppose row vectors of A
 are orthonormal. This implies $A A^T = I$

$\Rightarrow A$ is invertible & $A^T A = I$

$A^T A = I \Rightarrow$ column vectors are
 orthogonal

FACT: Every non-constant polynomial over \mathbb{C} has a root in \mathbb{C} i.e. if $p(x) = a_0 + a_1x + \dots + a_nx^n$ is a polynomial of degree n ($n \geq 1$) then $\exists \alpha \in \mathbb{C}$ s.t. $p(\alpha) = 0$. In fact, for $p(x) \exists \alpha_1, \dots, \alpha_n$ in \mathbb{C} such that $p(\alpha_1) = \dots = p(\alpha_n) = 0$. We will use this fact to prove that eigen values of real symmetric matrix is real.

Theorem:- Let A be a real symmetric matrix of order $n \times n$. Then all eigen values of A are real.

Proof:- Eigen values are roots of the characteristic polynomial $\det(A - xI)$. Let $p(x) = \det(A - xI)$. Then by the fact above $\exists \lambda \in \mathbb{C}$ such that $p(\lambda) = 0$

$$\text{i.e. } \det(A - \lambda I) = 0$$

Let $u \in \mathbb{R}^n$ be a non-zero eigen vector of A corresponding to eigen value λ .

$$\text{Then } Au = \lambda u.$$

$$\Rightarrow (Au)^T = \lambda u^T$$

$$\Rightarrow u^T A = \lambda u^T \quad (\text{since } A^T = A)$$

Take complex conjugate both sides, then

$$\bar{u}^T \bar{A} = \bar{\lambda} \bar{u}^T \quad \text{if } u = (u_1, \dots, u_n)$$

where $u_i \in \mathbb{C}$

$$\text{then } \bar{u} = (\bar{u}_1, \dots, \bar{u}_n)$$

$$\& \text{ if } A = (a_{ij}),$$

$$\bar{A} = (\bar{a}_{ij}).$$

$$\text{Now, } \bar{u}^T A u = \bar{\lambda} \bar{u}^T u$$

$$\Rightarrow \bar{u}^T (\lambda u) = \bar{\lambda} u^T u$$

$$\Rightarrow \lambda \bar{u}^T u = \bar{\lambda} u^T u$$

$$\Rightarrow (\lambda - \bar{\lambda}) \bar{u}^T u = 0$$

$\bar{u}^T u$ is a non-zero number

Hence $\lambda = \bar{\lambda} \Rightarrow \lambda$ is real. \square

Lemma: Let A be a real matrix with order $n \times n$ and suppose there exists a real eigen value λ of A . Then \exists a real non-zero eigen vector of λ .

Proof: Consider the system of linear equation $(A - \lambda I)x = 0$.

$\det(A - \lambda I) = 0$ implies it has a non-zero solution $u \in \mathbb{R}^n$.

$$(A - \lambda I)u = 0 \Rightarrow Au = \lambda u$$

u is an eigen vector corresponding to λ . \square

Orthogonal Matrix: A square matrix A is said to be orthogonal if $AA^T = A^T A = I$.

Theorem: Let A be a real symmetric matrix of order $n \times n$. Then there exists a real orthogonal matrix P such that $P^{-1}AP$ is a diagonal matrix.

Proof:- Let λ be an eigen value of A .

Then by previous theorem, λ is real.

By lemma, there exists a non-zero real vector u of A corresponding to λ .

By Gram-Schmidt orthogonalization process, \exists an orthonormal basis $\{u, f_2, \dots, f_n\}$ of \mathbb{R}^n containing u .

Let P_1 be the matrix whose column vectors are u, f_2, \dots, f_n . As column vectors are orthonormal therefore the matrix $P_1 \equiv (u \ f_2 \ \dots \ f_n)$ is orthonormal.

Consider the matrix $P_1^T A P_1$ and let $e_1 = (1, 0, \dots, 0)$.

The first column of $P_1^T A P_1$ is given by

$$\begin{aligned} (P_1^T A P_1) e_1 &= (P_1^T A) (P_1 e_1) = P_1^T A u \\ &= P_1^T \lambda u \\ &= \lambda P_1^T u \\ &= \lambda e_1 \quad \text{since } P_1 e_1 = u \end{aligned}$$

Note that $P_1^T = P_1^{-1}$ & $P_1^T A P_1$ is also symmetric as $(P_1^T A P_1)^T = (P_1^T A P_1)^T$

$$\begin{aligned} &= P_1^T A^T (P_1^T)^T \\ &= P_1^T A P_1 \end{aligned}$$

Thus, first row of $P_1^T A P_1$ is also λe_1

So, $P_1^T A P_1$ is of the form

$$\begin{pmatrix} \lambda & 0 \\ 0 & A_1 \end{pmatrix} \quad \text{where } A_1 \text{ is real symmetric matrix of order } (n-1) \times (n-1)$$

The proof of theorem is by induction on n .

If $n=1$, it is obvious.

Let the statement be true for all real symmetric matrices of order $(n-1) \times (n-1)$.

Then \exists a real orthogonal matrix

P_2 of order $(n-1) \times (n-1)$ such that

$P_2^{-1} A_1 P_2$ is a diagonal matrix D .

$$\text{Let } P = P_1 \begin{pmatrix} 1 & 0 \\ 0 & P_2 \end{pmatrix}$$

Check that P is orthogonal and

$$P^{-1} A P = \begin{pmatrix} \lambda & 0 \\ 0 & D \end{pmatrix} \quad \square$$