

# Cauchy-Schwarz Inequality, Spectral Theorem, Cauchy Interlacing

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## 1 Cauchy–Schwarz inequality

**Theorem 1.1.** *For any two vectors  $u, v \in \mathbb{R}^n$*

$$|u^T v| \leq \|u\| \|v\|.$$

*Proof.* Let  $\{v_1, v_2, \dots, v_n\}$  be the set of orthogonal vector  $v_i \in \mathbb{R}^n$ , forming a basis of  $\mathbb{R}^n$ , where  $v_1 = v$  (we can always form such a set). Let us write  $u$  as a linear combination

$$u = \alpha_1 v + \alpha_2 v_2 + \dots + \alpha_n v_n,$$

$\alpha_i \in \mathbb{R}$ . Then as  $v_i^T v_i \geq 0$ ,

$$u^T u = \sum_{i=1}^n \alpha_i^2 v_i^T v_i \geq \alpha_1^2 v^T v.$$

Hence,

$$|\alpha_1| \leq \sqrt{\frac{u^T u}{v^T v}}.$$

Also,  $v^T u = \alpha_1 v^T v$ . Thus

$$|v^T u| = |\alpha_1| v^T v \leq \sqrt{\frac{u^T u}{v^T v}} v^T v = \sqrt{u^T u} \sqrt{v^T v} = \|u\| \|v\|.$$

It completes the proof. □

## 2 The Spectral Theorem (Orthogonal diagonalization of symmetric matrices)

**Theorem 2.1.** *Let  $M$  be a symmetric matrix of order  $n$ . Then*

$$M = V D V^T, \text{ equivalently, } V^T M V = D,$$

where  $V$  is a orthogonal matrix, (that is,  $V^T V = V V^T = I$ ) with columns  $v_1, v_2, \dots, v_n$  which are the eigenvectors of  $M$  corresponding to the eigenvalues  $\lambda_1, \dots, \lambda_n$ , and  $D = \text{diag}(\lambda_1, \dots, \lambda_n)$ .

*Proof.* We prove the theorem using induction on  $n$ . The case  $n = 1$  is true, since  $M = 1M1^T$ . Assume that the theorem is true when the order of matrix is  $n - 1$ .

Consider the case when the order of  $M$  is  $n$ . Using the eigenvector  $v_1$ , let us make an orthonormal basis of  $\mathbb{R}^n$ , suppose this be  $S = \{v_1, x_2, \dots, x_n\}$ . Note that since  $M$  is symmetric, for any  $i = 2, \dots, i = n$

$$(M x_i)^T v_1 = x_i^T M^T v_1 = x_i^T M v_1 = \lambda_1 x_i^T v_1 = 0. \tag{1}$$

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Now, let  $Q$  be an orthogonal matrix whose columns are  $v_1, x_2, \dots, x_n$ . Then using [1](#)

$$\begin{aligned} Q^T M Q &= \begin{bmatrix} - & v_1^T & - \\ - & x_2^T & - \\ \vdots & \vdots & \vdots \\ - & x_n^T & - \end{bmatrix} \begin{bmatrix} | & | & \dots & | \\ \lambda_1 v_1 & M x_2 & \dots & M x_n \\ | & | & \dots & | \end{bmatrix} \\ &= \begin{bmatrix} \lambda_1 & \mathbf{0} \\ \mathbf{0} & \hat{M} \end{bmatrix}. \end{aligned}$$

Note that  $\hat{M}$  is a symmetric matrix of order  $n-1$ . By induction hypothesis there exists an orthogonal matrix  $\hat{Q}$  such that  $\hat{Q}^T \hat{M} \hat{Q} = \hat{D}$  is a diagonal matrix. Considering the matrix

$$P = \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \hat{Q} \end{bmatrix},$$

we see that the product

$$P^T Q^T M Q P = \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \hat{Q} \end{bmatrix}^T \begin{bmatrix} \lambda_1 & \mathbf{0} \\ \mathbf{0} & \hat{M} \end{bmatrix} \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & \hat{Q} \end{bmatrix} \quad (2)$$

$$= \begin{bmatrix} \lambda_1 & \mathbf{0} \\ \mathbf{0} & \hat{D} \end{bmatrix} \quad (3)$$

is a diagonal matrix. As  $P^T Q^T Q P = I$ , matrix  $Q P$  is an orthogonal matrix. By [\(2\)](#)  $Q P$  contains the eigenvectors of  $M$  and the corresponding eigenvalues are in the diagonal matrix  $P^T Q^T M Q P$ . It completes the proof.  $\square$

### 3 Cauchy's Interlace Theorem

**Theorem 3.1.** *Let  $M$  be a symmetric matrix of order  $n$ , and  $B$  be its principal submatrix of order  $m$ , where,  $m < n$ . Suppose the eigenvalues of  $A$  are  $\lambda_1 \geq \dots \geq \lambda_n$  and the eigenvalues of  $B$  are  $\beta_1 \geq \dots \geq \beta_m$ . Then*

$$\lambda_k \geq \beta_k \geq \lambda_{k+n-m}, \quad \text{for } k = 1, \dots, m.$$

*Proof.* For some suitable matrices  $X, Z$  we can write

$$M = \begin{bmatrix} B & X^T \\ X & Z \end{bmatrix}.$$

Let  $x_1, \dots, x_n$  be eigenvectors of  $A$  corresponding to the eigenvalues  $\lambda_1 \geq \dots \geq \lambda_n$ . And let  $y_1, \dots, y_m$  be eigenvectors of  $B$  corresponding to the eigenvalues  $\beta_1 \geq \dots \geq \beta_m$ . Let us define the following two vector spaces.

1.  $W = \text{span}(y_1, \dots, y_m)$ . Note that

$$\beta_k = \min_{x \in W} \frac{x^T B x}{x^T x} \quad (4)$$

2.  $V = \text{span}(x_k, \dots, x_n)$ . Note that

$$\lambda_k = \max_{x \in V} \frac{x^T M x}{x^T x} \quad (5)$$

Construct a new vector space

$$\tilde{W} = \left\{ \begin{bmatrix} w \\ \mathbf{0} \end{bmatrix} \in \mathbb{R}^n \right\}, \quad w \in W.$$

Since  $\text{rank } V = n - k + 1$ , and  $\text{rank } \tilde{W} = k$ , there exists  $\tilde{w} = \begin{bmatrix} w \\ 0 \end{bmatrix} \in V \cap \tilde{W}$  for some  $w \in W$ . Then

$$\tilde{w}^T M \tilde{w} = \begin{bmatrix} w^T & 0 \end{bmatrix} \begin{bmatrix} B & X^T \\ X & Z \end{bmatrix} \begin{bmatrix} w \\ 0 \end{bmatrix} = w^T B w.$$

Using (4), (5)

$$\lambda_k \geq \frac{\tilde{w}^T M \tilde{w}}{\tilde{w}^T \tilde{w}} = \frac{w^T B w}{w^T w} \geq \beta_k.$$

The proof of the other inequality is similar. Now let us define the following two vector spaces.

1.  $W = \text{span}(y_k, \dots, y_m)$ . Note that

$$\beta_k = \max_{x \in W} \frac{x^T B x}{x^T x} \quad (6)$$

2.  $V = \text{span}(x_1, \dots, x_{k+n-m})$ . Note that

$$\lambda_{k+n-m} = \min_{x \in V} \frac{x^T M x}{x^T x} \quad (7)$$

Construct a new vector space

$$\tilde{W} = \left\{ \begin{bmatrix} w \\ 0 \end{bmatrix} \in \mathbb{R}^n \right\}, \quad w \in W.$$

Since  $\text{rank } V = k + n - m$ , and  $\text{rank } \tilde{W} = m - k + 1$ , there exists  $\tilde{w} = \begin{bmatrix} w \\ 0 \end{bmatrix} \in V \cap \tilde{W}$  for some  $w \in W$ . Then as before we have  $\tilde{w}^T M \tilde{w} = w^T B w$ . Using (6), (7)

$$\lambda_{k+n-m} \leq \frac{\tilde{w}^T M \tilde{w}}{\tilde{w}^T \tilde{w}} = \frac{w^T B w}{w^T w} \leq \beta_k.$$

It completes the proof. □