

## Spectral theorem

- For a real symmetric matrix  $A$ ,  $A = Q \Lambda Q^T$

Claim Schur's theorem: For a 方陣  $A$ ,  $A = Q T Q^{-1}$  ( $T$  is 上三角矩陣,  $\overline{Q^T} = Q^{-1}$ )

If  $A$  has real eigenvalues, then  $Q$  and  $T$  can be chosen real:  $Q^T = Q^{-1}$

Remark  $Q$  could be complex. If  $\overline{Q^T} = Q^{-1}$ , then  $Q$  is called a unitary matrix.

$Q \overline{Q^T} = I$ . For real matrices,  $Q Q^T = I$ ,  $Q$ : orthogonal matrix

pf: Schur's theorem

- We prove this by induction on  $n$
- $n=1$ :  $A = [a] \Rightarrow a = 1 \cdot a \cdot 1^{-1}$
- 假設  $k \times k$  matrices 成立 and let  $A$  be a  $(k+1) \times (k+1)$  matrix
- Let  $\lambda_1$  be a eigenvalue of  $A$  and  $q_1$  be a corresponding unit eigenvector.

Using Gram-Schmidt process, we construct  $q_2, \dots, q_{k+1}$  s.t.  $q_1, \dots, q_{k+1}$  constitute an orthonormal basis for  $\mathbb{C}^{k+1}$ .

Let  $Q_1 = [q_1 \dots q_{k+1}]$ , then  $Q_1$  is unitary. ( $\overline{Q_1^T} = Q_1^{-1}$ )

$$\text{And } \overline{Q_1^T} A Q_1 = \begin{bmatrix} \overline{q_1^T} \\ \vdots \\ \overline{q_{k+1}^T} \end{bmatrix} [A q_1 \dots A q_{k+1}] = \begin{bmatrix} \overline{q_1^T} \\ \vdots \\ \overline{q_{k+1}^T} \end{bmatrix} [\lambda_1 q_1 \quad A q_2 \dots A q_{k+1}] = \begin{bmatrix} \lambda_1 \times \dots \times \lambda_{k+1} & & \\ & \ddots & \\ & & A_2 \end{bmatrix}_{k+1}$$

$$(\lambda_1 \overline{q_1^T} \cdot q_1 = \lambda_1 \|q_1\|^2 = \lambda_1, \quad \overline{q_2^T} \lambda_1 q_1 = \lambda_1 \overline{q_2^T} q_1 = 0)$$

- By the induction hypothesis, since  $A_2$  is  $k \times k$ , we have  $A_2 = Q_2 T_2 \overline{Q_2^T}$   
( $Q_2$  is unitary,  $T_2$  is 上三角矩陣)  $T_2 = \overline{Q_2^T} A_2 Q_2$

Let  $Q = Q_1 \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & Q_2 \end{bmatrix}_{(k+1) \times (k+1)}$ , then  $Q$  is unitary, i.e.,  $Q \overline{Q^T} = I$

$$(Q \overline{Q^T} = Q_1 \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & Q_2 \end{bmatrix} \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & \overline{Q_2^T} \end{bmatrix} \overline{Q_1^T} = Q_1 \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & I_k \end{bmatrix} \overline{Q_1^T} = Q_1 I \overline{Q_1^T} = Q_1 \overline{Q_1^T} = I)$$

$$\text{Also, we have } \overline{Q^T} A Q = \underbrace{\begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & \overline{Q_2^T} \end{bmatrix}}_{\overline{Q^T}} \overline{Q_1^T} A Q_1 \underbrace{\begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & Q_2 \end{bmatrix}}_Q = \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & \overline{Q_2^T} \end{bmatrix} \begin{bmatrix} \lambda_1 \times \dots \times \lambda_{k+1} & & \\ & \ddots & \\ & & A_2 \end{bmatrix} \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & Q_2 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & \dots & 0 \\ \vdots & & & \\ 0 & & & \overline{Q_2^T} \end{bmatrix} \begin{bmatrix} \lambda_1 \times \dots \times \lambda_{k+1} & & \\ & \ddots & \\ & & A_2 Q_2 \end{bmatrix} = \begin{bmatrix} \lambda_1 \times \dots \times \lambda_{k+1} & & \\ & \ddots & \\ & & T_2 \end{bmatrix}_{(k+1) \times (k+1)} \triangleq T \quad \because T \text{ is 上三角矩陣}$$

- If  $\lambda_1$  is a real eigenvalue, then  $q_1$  and  $Q_1$  can stay real.

The induction step keeps everything real when  $A$  has real eigenvalues. Induction starts with the case of  $1 \times 1$ , and there is no problem.

pf: Spectral theorem

Every real symmetric matrix  $A$  has real eigenvalues. By Schur's Theorem,  $A = QTQ^T$

( $Q$  is orthogonal:  $Q^T = Q^{-1}$ ,  $T$  is 上三角矩阵) Then  $T = Q^T A Q$

This is a symmetric matrix since  $T^T = Q^T A Q = T$

If  $T$  is 三角矩阵 and also symmetric, it must be diagonal  $\Rightarrow T = \Lambda$

Therefore,  $A = Q \Lambda Q^T$

Claim If  $A$  is <sup>(real)</sup> symmetric, then # of nonzero eigenvalues = rank  $r$  = # of nonzero pivots

pf: Since  $A$  is diagonalizable, # of zero eigenvalues = AM of  $\lambda = 0$  = GM of  $\lambda = 0$   
 $= \dim(\lambda = 0 \text{ 的 eigenspace}) = \dim(N(A)) = n - r = n - (\text{\# of nonzero pivots})$

Positive Definitive Matrices 正定矩阵

ex:  $A = \begin{bmatrix} 1 & -2 \\ -2 & 6 \end{bmatrix}$   $\underline{x}^T A \underline{x} = [x_1 \ x_2] \begin{bmatrix} 1 & -2 \\ -2 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1^2 - 4x_1x_2 + 6x_2^2$  (quadratic form)

$B = \begin{bmatrix} 1 & -1 \\ -3 & 6 \end{bmatrix}$   $\underline{x}^T B \underline{x} = [x_1 \ x_2] \begin{bmatrix} 1 & -1 \\ -3 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1^2 - 4x_1x_2 + 6x_2^2 = \underline{x}^T A \underline{x}$

o In general, if  $B$  is not symmetric, then let  $A = \frac{1}{2}(B + B^T)$ , which is symmetric, and  $\underline{x}^T B \underline{x} = \underline{x}^T A \underline{x}$ .

Therefore, we only need to consider symmetric matrices in quadratic form.

Def A real symmetric matrix  $A$  is said to be positive definite if  $\underline{x}^T A \underline{x} > 0$  (negative) ( $<$ ) for every nonzero vector  $\underline{x}$

ex:  $A = \begin{bmatrix} 1 & -2 \\ -2 & 6 \end{bmatrix}$   $\underline{x}^T A \underline{x} = x_1^2 - 4x_1x_2 + 6x_2^2 = (x_1 - 2x_2)^2 + 2x_2^2 > 0$  for every nonzero  $\underline{x}$

$\therefore A$  is positive definite

$B = \begin{bmatrix} -1 & 2 \\ 2 & -6 \end{bmatrix} \Rightarrow B$  is negative definite

Claim If  $A$  is <sup>(real)</sup> symmetric, then # of positive eigenvalues = # of positive pivots