

# Hint for Problem 1.12

Note: normal  $A$ , i.e.,  $A^t A = A A^t$

there is unitary  $V$  s.t.  $V^t A V = \begin{bmatrix} a_1 & & 0 \\ & \ddots & \\ 0 & & a_n \end{bmatrix}$

$$\Leftrightarrow A = V \begin{bmatrix} a_1 & & 0 \\ & \ddots & \\ 0 & & a_n \end{bmatrix} V^t$$

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## Problem 1.12

$H = H^t$  so  $H$  is normal

$$\therefore H = V \begin{bmatrix} \Re_1 & & 0 \\ & \ddots & \\ 0 & & \Re_n \end{bmatrix} V^t$$

$$\begin{aligned} I \pm iH &= V I V^t \pm i V \begin{bmatrix} \Re_1 & & 0 \\ & \ddots & \\ 0 & & \Re_n \end{bmatrix} V^t \\ &= V \begin{pmatrix} 1 \pm i\Re_1 & & 0 \\ & \ddots & \\ 0 & & 1 \pm i\Re_n \end{pmatrix} V^t \end{aligned}$$

$$\begin{aligned} (I - iH)^{-1} &= \left[ V \begin{pmatrix} 1 - i\Re_1 & & 0 \\ & \ddots & \\ 0 & & 1 - i\Re_n \end{pmatrix} V^t \right]^{-1} \\ &= (V^t)^{-1} \begin{pmatrix} \frac{1}{1 - i\Re_1} & & 0 \\ & \ddots & \\ 0 & & \frac{1}{1 - i\Re_n} \end{pmatrix} V^t \\ &= V \begin{pmatrix} \frac{1}{1 - i\Re_1} & & 0 \\ & \ddots & \\ 0 & & \frac{1}{1 - i\Re_n} \end{pmatrix} V^t \end{aligned}$$

$$(I + iH)(I - iH)^{-1}$$

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$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$

§1.6 & 1.8 Eigenvalues

Let  $A$  be an  $n \times n$  complex matrix. Then  $\lambda \in \mathbb{C}$  is an eigenvalue of  $A$  if there is a nonzero eigenvector  $|\lambda\rangle$  such that  $A|\lambda\rangle = \lambda|\lambda\rangle$ .

- The matrix  $A$  always has complex eigenvalues because  $\det(tI - A) = 0$  always has a solution  $\lambda$  so that  $(\lambda I - A)|\lambda\rangle = |0\rangle$  has non-trivial solution.
- There is a unitary matrix  $U$  such that  $U^\dagger A U = T$  is in upper triangular form. Moreover,  $\det(tI - A) = \det(tI - T) = (t - T_{11}) \cdots (t - T_{nn})$ .
- The matrix  $A$  is normal if and only if  $T$  is diagonal; the matrix  $A$  is unitary if and only if  $T$  is a diagonal matrix so that the diagonal entries have moduli 1; the matrix  $A$  is Hermitian if and only if  $T$  is a real diagonal matrices.

$\det(\lambda I - A) = \det(\lambda U U^\dagger - U^\dagger U)$   
 $= \det(U (\lambda I - T) U^\dagger)$   
 $= \det(U) \det(\lambda I - T) \det(U^\dagger)$   
 $= \det(\lambda I - T)$

Proof of the Schur Triangularization Lemma. Suppose  $A \in M_n$ . We show that there a unitary  $U \in M_n$  such that  $U^\dagger A U = T$  is in upper triangular form. We prove the result by induction on  $n$ . If  $n = 1$ , the result is trivial. Suppose the result holds for matrices in  $M_{n-1}$ . Let  $A \in M_n$ . Solve  $\det(tI - A) = 0$  to get a solution  $\lambda$ . There is a nonzero vector  $|\lambda\rangle$  such that  $A|\lambda\rangle = \lambda|\lambda\rangle$ . We may replace  $|\lambda\rangle$  by  $|\lambda\rangle/\|\lambda\|$  and assume that  $|\lambda\rangle$  has unit length, and we can let  $U_1 \in M_n$  be unitary with  $|\lambda\rangle$  as the first column. Then  $A U_1 = [A|x_1\rangle \cdots A|x_n\rangle] = [\lambda|x_1\rangle |x_2\rangle \cdots |x_n\rangle]$  if  $U_1$  has columns  $|x_1\rangle = |\lambda\rangle, |x_2\rangle, \dots, |x_n\rangle$ . Because  $\langle x_1 | A | x_1 \rangle = \lambda$ , and  $\langle x_j | A | x_1 \rangle = 0$  for  $j = 2, \dots, n$ , we see that  $U_1^\dagger A U_1 = \begin{pmatrix} \lambda & * \\ 0 & A_1 \end{pmatrix}$ . By induction assumption there is a unitary  $U_2 \in M_{n-1}$  such that

$U_2^\dagger A_1 U_2 = T_1$  in upper triangular form. Let  $U = U_1 \begin{pmatrix} 1 & 0 \\ 0 & U_2 \end{pmatrix} \in M_n$ . Then  $U^\dagger U = I_n$  so that  $U$  is unitary, and  $U^\dagger A U = \begin{pmatrix} 1 & 0 \\ 0 & U_2^\dagger \end{pmatrix} \begin{pmatrix} \lambda & * \\ 0 & A_1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & U_2 \end{pmatrix} = \begin{pmatrix} \lambda & * \\ 0 & T_1 \end{pmatrix}$  is upper triangular form.

$\langle x_j | \lambda | x_1 \rangle = 0$   
 $\langle x_j | A | x_1 \rangle = 0$   
 $\begin{bmatrix} \lambda & * \\ 0 & A_1 \end{bmatrix}$

Proof of the consequences for normal, unitary, Hermitian matrices. Let  $U^\dagger A U = T$ . Suppose  $A$  is normal, i.e.,  $A^\dagger A = A A^\dagger$ . Then

$T^\dagger T = (U^\dagger A U)^\dagger (U^\dagger A U) = (U^\dagger A^\dagger U) (U^\dagger A U) = U^\dagger A^\dagger A U = U^\dagger A A^\dagger U = T T^\dagger$

$\det(\lambda I - T) = \det \begin{pmatrix} \lambda - T_{11} & & \\ & \ddots & \\ 0 & & \lambda - T_{nn} \end{pmatrix} = (\lambda - T_{11}) \cdots (\lambda - T_{nn})$

Now, the (1,1) entry of  $T^\dagger T$  is  $|T_{11}|^2$  and the (1,1) entry of  $T T^\dagger$  is  $|T_{11}|^2 + \dots + |T_{1n}|^2$ . So,  $T_{12} = \dots = T_{1n} = 0$ . Now, consider the (2,2), ..., (n,n) entries, we see that only the diagonal entries of  $T$  can be nonzero. So,  $T$  is a diagonal matrix. Conversely, if  $U^\dagger A U = T$  is in diagonal form, then  $A = U T U^\dagger$  so that  $A^\dagger A = U^\dagger T^\dagger U U^\dagger T U = U^\dagger T^\dagger T U$  and  $A A^\dagger = U^\dagger T T^\dagger U$ . Note that for a diagonal matrix  $T$ , we have  $T T^\dagger = T^\dagger T$  is a diagonal matrix with diagonal entries  $|T_{11}|^2, \dots, |T_{nn}|^2$ . So,  $A^\dagger A = A A^\dagger$ .

Now, if  $A$  is unitary, then  $A$  is normal and  $T = U^\dagger A U$  is diagonal, and is unitary. So,  $T^\dagger T = I_n$  implies  $|T_{jj}|^2 = 1$  for all  $j$ . Conversely, if  $A = U^\dagger T U$  such that  $T$  is diagonal with all diagonal entries satisfying  $|T_{jj}| = 1$ , then  $A^\dagger A = U T^\dagger U^\dagger U T U = U T^\dagger T U = U U^\dagger = I_n$ .

Finally, suppose  $A$  is Hermitian, then  $A = A^\dagger$  so that the diagonal matrix  $T$  satisfies  $T^\dagger = U A^\dagger U^\dagger = U A U^\dagger = T$ . So,  $T$  is Hermitian and all the diagonal entries of  $T$  are real. Conversely, if  $T$  is a real diagonal matrix, then  $T = T^\dagger$  so that  $A^\dagger = U^\dagger T^\dagger U = U^\dagger T U = A$ .  $\square$

$\begin{bmatrix} T_{11} & 0 & 0 \\ 0 & T_{22} & \\ & 0 & \ddots \\ * & & & T_{33} \end{bmatrix} = \begin{bmatrix} T_{11} & & \\ & T_{22} & \\ & & T_{33} \end{bmatrix}$



### §1.9 Spectral decomposition

$$\rightarrow U \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} U^\dagger$$

- Suppose  $A$  is normal and  $A = UDU^\dagger$ , where  $D$  has diagonal entries  $\lambda_1, \dots, \lambda_n$  and  $U$  has orthonormal columns  $|\lambda_1\rangle, \dots, |\lambda_n\rangle$ . Then  $A = \lambda_1 |\lambda_1\rangle\langle\lambda_1| + \dots + \lambda_n |\lambda_n\rangle\langle\lambda_n|$  so that

$$A^k = \sum_{j=1}^n \lambda_j^k |\lambda_j\rangle\langle\lambda_j| \quad \text{and} \quad e^A = \sum_{k=0}^{\infty} \frac{1}{k!} A^k = \sum_{j=1}^n e^{\lambda_j} |\lambda_j\rangle\langle\lambda_j|.$$

- Also,  $A = \sum_{\alpha} \alpha P_{\alpha}$ , where  $P_{\alpha} = \sum_{\lambda_j = \alpha} |\lambda_j\rangle\langle\lambda_j|$ . Then  $A^k = \sum_{\alpha} \alpha^k P_{\alpha}$  and  $e^A = \sum_{\alpha} e^{\alpha} P_{\alpha}$ .

*Proof.* Because  $A|\lambda_j\rangle = \lambda_j|\lambda_j\rangle$  for  $j = 1, \dots, n$ , we have  $AU = UD$  so that

$$A = UDU^\dagger = \sum_{j=1}^n \lambda_j |\lambda_j\rangle\langle\lambda_j|. \quad A = \begin{bmatrix} |\lambda_1\rangle & |\lambda_2\rangle & \dots & |\lambda_n\rangle \end{bmatrix} \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \begin{bmatrix} \langle\lambda_1| \\ \vdots \\ \langle\lambda_n| \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 |\lambda_1\rangle & \dots & \lambda_n |\lambda_n\rangle \end{bmatrix} \begin{bmatrix} \langle\lambda_1| \\ \vdots \\ \langle\lambda_n| \end{bmatrix} = \lambda_1 |\lambda_1\rangle\langle\lambda_1| + \dots + \lambda_n |\lambda_n\rangle\langle\lambda_n|$$

Now,

$$A^k = \underbrace{UDU^\dagger \dots UDU^\dagger}_k = UD^k U^\dagger = \sum_{j=1}^n \lambda_j^k |\lambda_j\rangle\langle\lambda_j|,$$

and

$$e^A = \sum_{k=0}^{\infty} \frac{1}{k!} A^k = \sum_{k=0}^{\infty} \frac{1}{k!} \sum_{j=1}^n \lambda_j^k |\lambda_j\rangle\langle\lambda_j| = \sum_{j=1}^n e^{\lambda_j} |\lambda_j\rangle\langle\lambda_j|.$$

$$e^x = \sum_{k=0}^{\infty} \frac{1}{k!} x^k$$

$$\sum_{k=0}^{\infty} \frac{1}{k!} \lambda_1^k |\lambda_1\rangle\langle\lambda_1| + \sum_{k=0}^{\infty} \frac{1}{k!} \lambda_2^k |\lambda_2\rangle\langle\lambda_2| + \dots$$

$$A = U \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} U^\dagger$$

$$A^k = U \begin{bmatrix} \lambda_1^k & & 0 \\ & \ddots & \\ 0 & & \lambda_n^k \end{bmatrix} U^\dagger$$

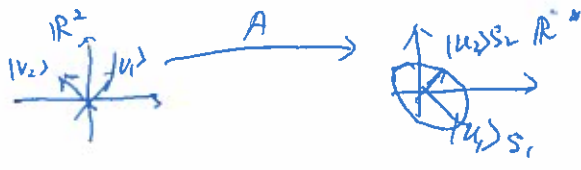
$$= \begin{bmatrix} \lambda_1^k |\lambda_1\rangle & \lambda_2^k |\lambda_2\rangle & \dots & \lambda_n^k |\lambda_n\rangle \end{bmatrix} \begin{bmatrix} \langle\lambda_1| \\ \vdots \\ \langle\lambda_n| \end{bmatrix}$$

Remark:

$$\begin{aligned} & \text{If } \lambda_1 = \lambda_2 \\ & \lambda_1 |\lambda_1\rangle\langle\lambda_1| + \lambda_2 |\lambda_2\rangle\langle\lambda_2| \\ & = \lambda_1 (|\lambda_1\rangle\langle\lambda_1| + |\lambda_2\rangle\langle\lambda_2|) \\ & = \lambda_1 P_1 \end{aligned}$$

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} d_1 & 0 \\ 0 & d_2 \end{bmatrix} = \begin{bmatrix} a_{11}d_1 & a_{12}d_2 \\ a_{21}d_1 & a_{22}d_2 \end{bmatrix}$$

$$= \begin{bmatrix} | & | \\ | & | \end{bmatrix} \begin{bmatrix} S_1 \\ S_2 \end{bmatrix} = \begin{bmatrix} | & | \\ | & | \end{bmatrix} \begin{bmatrix} S_1 \\ S_2 \end{bmatrix}$$



$$AV = UD$$

$$A[|v_1\rangle \dots |v_n\rangle] = [|u_1\rangle \dots |u_n\rangle]$$

§1.9 Singular value decomposition (SVD)

**Theorem** Every square matrix  $A$  can be written as  $A = UDV^T$ , where  $U, V$  are unitary and  $D = \text{diag}(s_1, \dots, s_n)$  with  $s_1 \geq \dots \geq s_n \geq 0$ .

*Proof.* Proof by induction on the size of  $A$ . Clearly, the result holds for  $n = 1$ . Assume that the result is true for matrices of size less than  $n$  with  $n \geq 2$ . Choose unitary  $U, V$  such that  $S = U_1^T A V_1$  has largest (1,1) entry with largest real part  $S_{11}$ . First, we have  $S_{11} \geq 0$ , else replace  $V$  by  $e^{i\theta} V$  for some suitable  $e^{i\theta}$ . Next, we claim that  $S_{1j} = 0$  for  $j = 2, \dots, n$ . If not, we can replace  $V_1$  by  $V_1 R$  where  $R$  is obtained from  $I_n$  by changing the  $2 \times 2$  submatrix in rows and columns 1,  $j$  by  $\frac{1}{\sqrt{S_{11}^2 + |S_{1j}|^2}} \begin{pmatrix} S_{11} & -S_{1j} \\ S_{1j} & S_{11} \end{pmatrix}$ . Then  $U_1^T A V_2 R$  has (1,1) entry equal to  $\sqrt{S_{11}^2 + |S_{1j}|^2} > S_{11}$ , which is a contradiction. Similarly, we can show that  $S_{j1} = 0$  for all  $j = 2, \dots, n$ . Thus,  $S = \begin{pmatrix} 1 & 0 \\ 0 & \hat{S} \end{pmatrix}$ . By induction assumption, there are unitary  $U_2, V_2$  such that  $U_2^T \hat{S} V_2 = \text{diag}(s_2, \dots, s_n)$ .

Let  $V = V_1 \begin{pmatrix} 1 & \\ & V_2 \end{pmatrix}$  and  $U = U_1 \begin{pmatrix} 1 & \\ & U_2 \end{pmatrix}$ . Then  $U^T A V = \text{diag}(s_1, \dots, s_n)$ . □

**Note**  $s_1, \dots, s_n$  are the singular values of  $A$  and equal the nonnegative square roots of the eigenvalues of  $A^T A$  or / and  $AA^T$ .

The matrix  $A$  has rank  $k$  if and only if it has  $k$  nonzero singular values.

In practice, write  $A^T A = V D^2 V^T$  so that the columns of the unitary matrix  $V$ ,  $|v_1\rangle, \dots, |v_n\rangle$ , are the right singular vectors. Then  $AV = UD$  for some unitary  $U$  such that  $A = UDV^T$ . In particular, if  $s_j \neq 0$  then  $|u_j\rangle = A|v_j\rangle/s_j$ ; if  $A$  has rank  $k$  so that  $s_k > 0 = s_{k+1} = \dots = s_n$  then  $|u_{k+1}\rangle, \dots, |u_n\rangle$  are chosen so that  $\{|u_1\rangle, \dots, |u_n\rangle\}$  is an orthonormal basis.

In Matlab, we use the command  $[U, D, V] = \text{svd}(A)$ .

$$A = UDV^T$$

$$A^T A = V D^T U^T U D V^T = V D^2 V^T$$

$$AA^T = U D^2 U^T$$

$$\begin{bmatrix} s_1^2 & & 0 \\ & \ddots & \\ 0 & & s_n^2 \end{bmatrix}$$

$$S_{11} \cdot (-\bar{s}_{1j}) + (\bar{s}_{1j} \cdot S_{11})$$

$$\begin{pmatrix} \bar{s}_{1j} & -S_{1j} \\ S_{11} & S_{11} \end{pmatrix} \begin{pmatrix} -S_{1j} \\ S_{11} \end{pmatrix}$$

$$\begin{pmatrix} S_{11} & S_{1j} \\ S_{11} & S_{11} \end{pmatrix} \begin{pmatrix} -S_{1j} \\ S_{11} \end{pmatrix}$$

$$U_1^T A V_1 = \begin{bmatrix} S_{11} & S_{12} & \dots \\ \times & \times & \dots \\ \times & \times & \dots \\ \times & \times & \dots \end{bmatrix}$$

$$\begin{bmatrix} v_{11} & v_{21} & \dots \\ v_{12} & v_{22} & \dots \\ \vdots & \vdots & \ddots \end{bmatrix} U^T$$

$$\begin{bmatrix} v_{11} \\ v_{21} \end{bmatrix} = \begin{bmatrix} S_{11} \\ S_{12} \end{bmatrix} / \sqrt{S_{11}^2 + |S_{12}|^2}$$

§1.10 Tensor product (Kronecker product)

$A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$   $m \times n$   
 $B$   $p \times q$   
 $A \otimes B = \begin{pmatrix} a_{11}B & a_{12}B \\ a_{21}B & a_{22}B \end{pmatrix}$   $2m$   
 $A \otimes B : m \times p \times n \times q$

- $A \otimes B = (A_{ij}B)$  satisfies
  - $(A \otimes B)(C \otimes D) = (AC) \otimes (BD)$ ,
  - $A \otimes (B + C) = A \otimes B + A \otimes C$ ,  $(A \otimes B)^\dagger = A^\dagger \otimes B^\dagger$ ,  $(A \otimes B)^{-1} = A^{-1} \otimes B^{-1}$ .
  - $\text{tr}(A \otimes B) = (\text{tr } A)(\text{tr } B)$  and  $\det(A \otimes B) = \det(A)^n \det(B)^m$  if  $A \in M_m, B \in M_n$ .

*Proof.* (a) Let  $A$  be  $m \times n$ ,  $B$  be  $r \times s$ ,  $C$  be  $n \times p$ ,  $D$  be  $s \times q$ . Then  $A \otimes B = (A_{ij}B)$  is  $mr \times ns$ , and  $C \otimes D = (C_{ij}D)$  is  $ns \times pq$ . Now

$$(A \otimes B)(C \otimes D) = (A_{ij}B)(C_{ij}D) = (f_{rs}BD),$$

$$\begin{bmatrix} a_{11}B & a_{12}B \\ a_{21}B & a_{22}B \end{bmatrix} \begin{bmatrix} C_{11}D & C_{12}D & C_{13}D \\ C_{21}D & C_{22}D & C_{23}D \end{bmatrix}$$

where  $f_{rs} = \sum_{\ell=1}^n A_{r\ell}C_{\ell s} = (AC)_{rs}$ . So,  $(f_{rs}BD) = AC \otimes BD$ .

- $A \otimes (B + C) = (A_{ij}(B + C)) = (A_{ij}B) + (A_{ij}C) = A \otimes B + A \otimes C$ .
- $(A \otimes B)^\dagger = \overline{(A \otimes B)}^t = \overline{(A^\dagger \otimes B^\dagger)} = A^\dagger \otimes B^\dagger$ .

$$(A^{-1} \otimes B^{-1})(A \otimes B) = (A^{-1}A) \otimes (B^{-1}B) = I_m \otimes I_n = I_{mn} \implies A^{-1} \otimes B^{-1} = (A \otimes B)^{-1}.$$

(c) Note that  $\text{tr}(AB) = \sum_{i=1}^n \sum_{j=1}^n A_{ij}B_{ji} = \text{tr}(BA)$ . Assume that  $A$  is  $m \times m$  and  $B$  is  $n \times n$ . Let  $U, V$  be unitary and  $S, T$  be triangular such that  $A = USU^\dagger$  and  $B = VTV^\dagger$ . Then  $\text{tr } A = \text{tr } S, \text{tr } B = \text{tr } T$ ,

$$\begin{aligned} \text{tr}(A \otimes B) &= \text{tr}((U \otimes V)(S \otimes T)(U \otimes V)^\dagger) = \text{tr}(S \otimes T) \\ &= \sum_{i=1}^m S_{ii} \sum_{j=1}^n T_{jj} \left( \sum_{i=1}^m S_{ii} \right) \left( \sum_{j=1}^n T_{jj} \right) = (\text{tr } S)(\text{tr } T) = (\text{tr } A)(\text{tr } B). \end{aligned}$$

Also,

$$\begin{aligned} \det(A \otimes B) &= \det((U \otimes V)(S \otimes T)(U \otimes V)^\dagger) = \det(S \otimes T) \\ &= \prod_{i=1}^m S_{ii}^n \det(T) = (\det S)^n (\det T)^m = (\det A)^n (\det B)^m. \end{aligned}$$