

# Math 341: Homework 9

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## §1 A

Let  $W$  be a finite-dimensional subspace of an inner product space  $V$

- a. Prove that  $V = W \oplus W^\perp$

*Proof.*

- i.  $W \cap W^\perp = \{0\}$

Let  $v \in W \cap W^\perp$ . This means that  $\langle v, v \rangle = 0$ . This implies that  $v = 0$  by Theorem 6.1.

- ii.  $W + W^\perp = V$

Theorem 6.6. states that for any  $v \in V$ , there exist unique vectors  $u \in W$  and  $z \in W^\perp$  such that  $v = u + z$ .

Therefore,  $V = W \oplus W^\perp$ . □

- b. Prove that there exists a projection  $T$  on  $W$  along  $W^\perp$  that satisfies  $N(T) = W^\perp$ .

*Proof.* From (a), we know that  $V = W \oplus W^\perp$ , so this means that there is a projection  $T$  on  $W$  along  $W^\perp$  such that whenever  $x = x_1 + x_2$  where  $x_1 \in W$  and  $x_2 \in W^\perp$ ,  $T(x) = x_1$ .

Let  $n \in N(T)$ . We can express  $n$  as  $n = x_1 + x_2$  where  $x_1 \in W$  and  $x_2 \in W^\perp$  such that  $T(n) = 0$ . This implies that  $x_1 = 0$  so  $n = x_2 \in W^\perp$ . Thus,  $N(T) \subseteq W^\perp$ .

Let  $n \in W^\perp$ . We can express  $n$  as  $n = x_1 + x_2 = 0 + x_2$  where  $x_1 \in W$  and  $x_2 \in W^\perp$ . It is clear that  $T(n) = 0$ , so  $n \in N(T)$ . Thus,  $W^\perp \subseteq N(T)$ . □

- c. Prove that  $\|T(x)\| \leq \|x\|$  for all  $x \in V$ .

*Proof.* Let  $x \in V$ . We can express  $x$  as  $x = x_1 + x_2$  where  $x_1 \in W$  and  $x_2 \in W^\perp$ .

Notice that  $\|T(x)\| = \|x_1\|$ . Additionally,  $\|x\| = \|x_1 + x_2\| \leq \|x_1\| + \|x_2\|$  by Theorem 6.2. Combining these two equations together, we get  $\|T(x)\| \leq \|x\|$ , as desired. □

## §2 B

Let  $T : V \rightarrow W$  be a linear transformation, where  $V$  and  $W$  are finite dimensional inner product spaces with inner products  $\langle \cdot, \cdot \rangle_1$  and  $\langle \cdot, \cdot \rangle_2$  respectively. Prove the following results

- a. There is a unique adjoint  $T^*$  of  $T$ , and  $T^*$  is linear.

*Proof.* Suppose we have two unique adjoints,  $T^*$  and  $U^*$ . Then,  $\langle T(x), y \rangle_2 = \langle x, T^*(y) \rangle_1$  and  $\langle T(x), y \rangle_2 = \langle x, U^*(y) \rangle_1$ . Thus,  $\langle x, T^*(y) \rangle_1 = \langle x, U^*(y) \rangle_1$  and it follows that  $T^* = U^*$  by Theorem 6.1. Hence, the adjoint must be unique.

We prove that  $T^*$  is linear. Let  $y_1, y_2 \in W$ ,  $x \in V$ , and  $c \in F$ . Then,

$$\begin{aligned}\langle x, T^*(cy_1 + y_2) \rangle_1 &= \langle T(x), cy_1 + y_2 \rangle_2 \\ &= \langle T(x), cy_1 \rangle_2 + \langle T(x), y_2 \rangle_2 \\ &= c\langle T(x), y_1 \rangle_2 + \langle T(x), y_2 \rangle_2 \\ &= c\langle x, T^*(y_1) \rangle_1 + \langle x, T^*(y_2) \rangle_1\end{aligned}$$

as desired.  $\square$

- b. If  $\beta$  and  $\gamma$  are orthonormal bases for  $V$  and  $W$ , respectively, then  $[\mathbf{T}^*]_\gamma^\beta = ([\mathbf{T}]_\beta^\gamma)^*$

*Proof.* Let  $\beta = \{v_1, \dots, v_n\}$ ,  $\gamma = \{w_1, \dots, w_m\}$ ,  $A = [T^*]_\gamma^\beta$ ,  $B = [T]_\beta^\gamma$ .

Since  $T^*(w_j) \in V = \text{span}(\beta) \Rightarrow T^*(w_j) = \sum_{i=1}^n \langle T^*(w_j), v_i \rangle_1 v_i$  by Corollary 1 of Theorem 6.3. Then using the corollary of Theorem 6.5, this implies that  $A_{ij} = ([T^*]_\gamma^\beta)_{ij} = \langle T^*(w_j), v_i \rangle_1$ . Similarly,  $T(v_j) \in W = \text{span}(\gamma) \Rightarrow T(v_j) = \sum_{i=1}^m \langle T(v_j), w_i \rangle_2 w_i$  and  $B_{ij} = \langle T(v_j), w_i \rangle_2$ . It follows that

$$\begin{aligned}(B_{ij})^* &= \overline{B_{ji}} \\ &= \overline{\langle T(v_i), w_j \rangle_2} \\ &= \langle w_j, T(v_i) \rangle_2 \\ &= \langle T^*(w_j), v_i \rangle_1 \\ &= A_{ij}\end{aligned}$$

Therefore,  $[\mathbf{T}^*]_\gamma^\beta = ([\mathbf{T}]_\beta^\gamma)^*$ .  $\square$

- c.  $\text{rank}(T^*) = \text{rank}(T)$

*Proof.* We know that

$$\dim(V) = \dim(N(T)) + \dim(R(T))$$

by the rank nullity theorem. By Theorem 6.7, we know that since  $N(T)$  is a subspace of  $V$ ,

$$\dim(V) = \dim(N(T)) + \dim(N(T)^\perp)$$

Hence,

$$\dim(R(T)) = \dim(N(T)^\perp)$$

We claim that  $N(T) = R(T^*)^\perp$ . Suppose  $x \in N(T)$ . Let  $y \in V$ . Then,

$$T(x) = 0 = \langle 0, y \rangle = \langle T(x), y \rangle = \langle x, T^*(y) \rangle$$

By definition,  $x \in N(T)^\perp$ , so  $N(T) \subseteq R(T^*)^\perp$ .

Suppose  $x \in R(T^*)^\perp$ . Then for all  $y \in V$ ,

$$0 = \langle x, T^*(y) \rangle = \langle T(x), y \rangle$$

This means that  $T(x) = 0$ , hence  $x \in N(T)$ . So  $R(T^*)^\perp \subseteq N(T)$

Thus,  $N(T) = R(T^*)^\perp$ . It directly follows that  $N(T)^\perp = (R(T^*)^\perp)^\perp = R(T^*)$ . Therefore,

$$\begin{aligned}\dim(R(T)) &= \dim(N(T)^\perp) \\ &= \dim(R(T^*)) \\ \text{rank}(T) &= \text{rank}(T^*)\end{aligned}$$

□

d.  $\langle T^*(x), y \rangle_1 = \langle x, T(y) \rangle_2$  for all  $x \in W$  and  $y \in V$

*Proof.*

$$\begin{aligned}\langle T^*(x), y \rangle_1 &= \overline{\langle y, T^*(x) \rangle_1} \\ &= \overline{\langle T(y), x \rangle_2} \\ &= \langle x, T(y) \rangle_2\end{aligned}$$

as desired. □

e. For all  $x \in V$ ,  $T^*T(x) = 0$  if and only if  $T(x) = 0$

*Proof.*

i.  $T^*T(x) = 0 \Rightarrow T(x) = 0$

This means that  $\forall x \in V$ ,

$$\begin{aligned}0 &= \langle x, T^*T(x) \rangle_1 \\ &= \langle T(x), T(x) \rangle_2\end{aligned}$$

Therefore,  $T(x) = 0$

ii.  $T(x) = 0 \Rightarrow T^*T(x) = 0$

$$T^*T(x) = T^*(0) = 0$$

as desired.

Therefore,  $T^*T(x) = 0$  if and only if  $T(x) = 0$ . □

### §3 C

a. Show that the equation  $(A^*A)x_0 = A^*y$  of Theorem 6.12 takes the form of the normal equations:

$$\left( \sum_{i=1}^m t_i^2 \right) c + \left( \sum_{i=1}^m t_i \right) d = \sum_{i=1}^m t_i y_i$$

and

$$\left( \sum_{i=1}^m t_i \right) c + md = \sum_{i=1}^m y_i$$

*Proof.* Suppose

$$A = \begin{bmatrix} t_1 & 1 \\ t_2 & 1 \\ \vdots & \vdots \\ t_n & 1 \end{bmatrix}$$

Then,

$$A^* = \begin{bmatrix} t_1 & t_2 & \cdots & t_n \\ 1 & 1 & \cdots & 1 \end{bmatrix}$$

and

$$A^*A = \begin{bmatrix} \sum_{i=1}^m (t_i)^2 & \sum_{i=1}^m t_i \\ \sum_{i=1}^m t_i & m \end{bmatrix}$$

Hence,

$$A^*Ax_0 = A^*y$$

$$\begin{bmatrix} \sum_{i=1}^m (t_i)^2 & \sum_{i=1}^m t_i \\ \sum_{i=1}^m t_i & m \end{bmatrix} \begin{bmatrix} c \\ d \end{bmatrix} = \begin{bmatrix} t_1 & t_2 & \cdots & t_m \\ 1 & 1 & \cdots & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix}$$

$$\left( \sum_{i=1}^m t_i^2 \right) c + \left( \sum_{i=1}^m t_i \right) d = \sum_{i=1}^m t_i y_i$$

and

$$\left( \sum_{i=1}^m t_i \right) c + md = \sum_{i=1}^m y_i$$

as desired.  $\square$

- b. Use the second normal equation of (a) to show that the least squares line must pass through the center of mass,  $(\bar{t}, \bar{y})$ , where

$$\bar{t} = \frac{1}{m} \sum_{i=1}^m t_i \quad \text{and} \quad \bar{y} = \frac{1}{m} \sum_{i=1}^m y_i$$

*Proof.* Consider what happens when we divide the following equation by  $m$ .

$$\begin{aligned} \left( \sum_{i=1}^m t_i \right) c + md &= \sum_{i=1}^m y_i \\ \frac{1}{m} \left( \sum_{i=1}^m t_i \right) c + d &= \frac{1}{m} \sum_{i=1}^m y_i \\ \bar{t}c + d &= \bar{y} \end{aligned}$$

as desired.  $\square$

## §4 D

An  $n \times n$  real matrix  $A$  is said to be a Gramian matrix if there exists a real (square) matrix  $B$  such that  $A = B^t B$ . Prove that  $A$  is a Gramian matrix if and only if  $A$  is symmetric and all of its eigenvalues are nonnegative. Hint: Apply Theorem 6.17 to  $T = L_A$  to obtain an orthonormal basis  $\{v_1, v_2, \dots, v_n\}$  of eigenvectors with the associated eigenvalues  $\{\lambda_1, \lambda_2, \dots, \lambda_n\}$ . Define the linear operator  $U$  by  $U(v_i) = \sqrt{\lambda_i} v_i$ .

*Proof.*

- i.  $A$  is a Gramian matrix  $\Rightarrow A$  is symmetric and all of its eigenvalues are nonnegative.

Since  $A$  is a Gramian matrix, there exists a real (square) matrix  $B$  such that  $A = B^t B$ . This implies that  $A$  is symmetric

$$\begin{aligned} A^t &= (B^t B)^t \\ &= B^t B \\ &= A \end{aligned}$$

as desired.

Suppose  $\lambda$  is an eigenvalue of  $A$  with a normal/unit eigenvector of  $x$ , i.e.  $Ax = \lambda x$ . It follows that

$$\begin{aligned}\lambda &= \lambda \langle x, x \rangle \\ &= \langle Ax, x \rangle \\ &= \langle B^t Bx, x \rangle \\ &= \langle Bx, Bx \rangle \\ &\geq 0\end{aligned}$$

because  $Bx \neq 0$  and Theorem 6.1.

Therefore,  $A$  is symmetric and all of its eigenvalues are nonnegative.

- ii.  $A$  is symmetric and all of its eigenvalues are nonnegative  $\Rightarrow A$  is a Gramian matrix

Since  $A$  is symmetric it follows that  $A$  and  $T = L_A$  must be self-adjoint because  $A$  is a real matrix, by definition. By Theorem 6.17, there exists an orthonormal basis  $\{v_1, v_2, \dots, v_n\}$  of eigenvectors with the associated eigenvalues  $\{\lambda_1, \lambda_2, \dots, \lambda_n\}$ . By Theorem 6.16.,  $T$  is normal and there exists a diagonal matrix  $[T]_\beta$  such that the  $i$ th diagonal entries are  $\lambda_i$ . Suppose we construct a diagonal matrix  $C$  such that its  $i$ th diagonal entries are  $\sqrt{\lambda_i}$ , i.e.  $C^2 = [T]_\beta$ . This will still be a real matrix because the eigenvalues are nonnegative. Let  $\alpha$  be the standard ordered basis. Hence,

$$A = [I]_\beta^\alpha [T]_\beta [I]_\alpha^\beta = [I]_\beta^\alpha C^2 [I]_\alpha^\beta$$

Since  $\beta$  is orthonormal,  $([I]_\beta^\alpha)^t [I]_\beta^\alpha = [I]$ . Thus,  $([I]_\beta^\alpha)^t = ([I]_\beta^\alpha)^{-1} = ([I]_\alpha^\beta)$ . We can fix  $B$  to be  $C[I]_\alpha^\beta$  such that

$$\begin{aligned}A &= [I]_\beta^\alpha C^2 [I]_\alpha^\beta \\ &= [I]_\beta^\alpha C C [I]_\alpha^\beta \\ &= B^t B\end{aligned}$$

as desired. Therefore,  $A$  is a Gramian matrix.

□

## §5 E

Let  $T$  and  $U$  be self-adjoint linear operators on an  $n$ -dimensional inner product space  $V$ , and let  $A = [T]_\beta$ , where  $\beta$  is an orthonormal basis for  $V$ . Prove the following results.

- a.  $T$  is positive definite [semidefinite] if and only if all of its eigenvalues are positive [nonnegative].

*Proof.*

- i.  $T$  is positive definite  $\Rightarrow$  all of its eigenvalues are positive

By definition of positive definite,  $\langle T(x), x \rangle > 0$  for all  $x \neq 0$ . Let  $\lambda$  be a eigenvalue of

$T$  with  $v$  being its corresponding orthonormal eigenvector. Consider,

$$\begin{aligned}\lambda &= \lambda \langle v, v \rangle \\ &= \langle \lambda v, v \rangle \\ &= \langle T(v), v \rangle \\ &> 0\end{aligned}$$

as desired.

- ii. All of  $T$ 's eigenvalues are positive  $\Rightarrow T$  is positive definite

By Theorem 6.17, we know that there exists an orthonormal basis  $\beta = \{v_1, \dots, v_n\}$  for  $V$  consisting of eigenvectors of  $T$ . Let  $x \in V$ . We can express  $x$  in terms of  $\beta$ ,

$$x = \sum_{i=1}^n c_i v_i$$

Now consider the following inner product,

$$\begin{aligned}\langle T(x), x \rangle &= \langle T\left(\sum_{i=1}^n c_i v_i\right), \sum_{i=1}^n c_i v_i \rangle \\ &= \left\langle \sum_{i=1}^n c_i \lambda_i v_i, \sum_{i=1}^n c_i v_i \right\rangle \\ &= \sum_{i=1}^n |c_i|^2 \lambda_i \langle v_i, v_i \rangle \\ &= \sum_{i=1}^n |c_i|^2 \lambda_i \\ &> 0\end{aligned}$$

as desired.

Without loss of generality, the above statements hold for the semidefinite case. Therefore,  $T$  is positive definite [semidefinite] if and only if all of its eigenvalues are positive [nonnegative].  $\square$

- b.  $T$  is positive definite if and only if  $\sum_{i,j} A_{ij} a_j \bar{a}_i > 0$  for all nonzero  $n$ -tuples  $(a_1, \dots, a_n)$

*Proof.*

- i.  $T$  is positive definite  $\Rightarrow \sum_{i,j} A_{ij} a_j \bar{a}_i > 0$  for all nonzero  $n$ -tuples  $(a_1, \dots, a_n)$

Let  $\beta = \{v_1, \dots, v_n\}$  be an orthonormal basis for  $V$ . Let  $a = (a_1, \dots, a_n)$  be a nonzero  $n$ -tuple. We can define  $x = \sum_{i=1}^n a_i v_i \in V$ . It follows that by corollary of Theorem 6.5

and the definition of positive definite,

$$\begin{aligned}
 \langle T(x), x \rangle &= \langle T(\sum_{j=0}^n a_j v_j), \sum_{i=0}^n a_i v_i \rangle \\
 &= \sum_{j=0}^n a_j \langle T(v_j), \sum_{i=0}^n a_i v_i \rangle \\
 &= \sum_{i,j=0}^n a_j \bar{a}_i \langle T(v_j), v_i \rangle \\
 &= \sum_{i,j=0}^n A_{ij} \bar{a}_i a_j \\
 &> 0
 \end{aligned}$$

as desired.

- ii.  $\sum_{i,j} A_{ij} a_j \bar{a}_i > 0$  for all nonzero  $n$ -tuples  $(a_1, \dots, a_n) \Rightarrow T$  is positive definite

Let  $x = \sum_{i=0}^n a_i v_i \in V$ . It follows from corollary of Theorem 6.5,

$$\begin{aligned}
 \sum_{i,j} A_{ij} a_j \bar{a}_i &= \sum_{i,j} a_j \bar{a}_i \langle T(v_j), v_i \rangle \\
 &= \langle T(\sum_j a_j v_j), \sum_i a_i v_i \rangle \\
 &= \langle T(x), x \rangle \\
 &> 0
 \end{aligned}$$

as desired.

Therefore,  $T$  is positive definite if and only if  $\sum_{i,j} A_{ij} a_j \bar{a}_i > 0$  for all nonzero  $n$ -tuples  $(a_1, \dots, a_n)$   $\square$

- c.  $T$  is positive semidefinite if and only if  $A = B^* B$  for some square matrix  $B$ .

*Proof.*

- i.  $T$  is positive semidefinite  $\Rightarrow A = B^* B$  for some square matrix  $B$ .

Since  $T$  is a self adjoint operator, there exists an orthonormal basis  $\gamma = \{v_1, \dots, v_n\}$  for  $V$  consisting of eigenvectors of  $T$  by Theorem 6.17. We also know from (a) that all of the corresponding eigenvalues are positive. Additionally, since  $T$  is a self adjoint operator we know that  $A = P^* D P$  where  $D$  is diagonal. It follows that on the diagonals on  $D$  are positive. Let's define a matrix  $E_{ii} = \sqrt{D_{ii}}$ , i.e.  $E^2 = D$ . Then we have  $A = (P^* E)(E P)$  where we can fix  $B = E P$  and we know that  $B^* = (E P)^* = P^* E^* = P^* E$  because  $E$  is a self-adjoint matrix. Therefore,  $A = B^* B$ .

- ii.  $A = B^* B$  for some square matrix  $B \Rightarrow T$  is positive semidefinite.

Consider  $v \in V$  and the following inner product,

$$\begin{aligned}\langle L_A(v), v \rangle &= \langle Av, v \rangle \\ &= \langle B^*Bv, v \rangle \\ &= B^*\langle Bv, v \rangle \\ &= \langle Bv, Bv \rangle \\ &\geq 0\end{aligned}$$

Since  $A = [T]_\beta$ ,  $T$  is positive semidefinite.

□

- d. If  $T$  and  $U$  are positive semidefinite operators such that  $T^2 = U^2$  then  $T = U$ .

*Proof.* Since  $T$  is a self adjoint operator, there exists an orthonormal basis  $\gamma = \{v_1, \dots, v_n\}$  for  $V$  consisting of eigenvectors of  $T$  by Theorem 6.17. We also know from (a) that all of the corresponding eigenvalues are nonnegative. Then,  $T^2(v_i) = \lambda^2 v_i = U^2(v_i)$ . It follows that

$$\begin{aligned}0 &= U^2 v_i - \lambda^2 v_i \\ &= (U^2 - \lambda^2 I)v_i \\ &= (U + \lambda I)(U - \lambda I)v_i\end{aligned}$$

This means that  $(U - \lambda I)v_i = 0$  because eigenvalues are nonnegative for  $T$ . Thus

$$\begin{aligned}(U - \lambda I)v_i &= 0 \\ U(v_i) &= \lambda v_i \\ &= T(v_i)\end{aligned}$$

Therefore,  $T = U$ .

□

- e. If  $T$  and  $U$  are positive definite operators such that  $TU = UT$  then  $TU$  is positive definite.

*Proof.* Since  $T$  and  $U$  are positive definite operators,  $TU$  is self adjoint because

$$(TU)^* = U^*T^* = UT = TU$$

Let  $Q^2 = T$  such that  $Q$  is positive definite. We know that  $Q^2U = QQU = QUQ$  because of part (d).

$$\begin{aligned}\langle TU(x), x \rangle &= \langle Q^2U(x), x \rangle \\ &= \langle QUQ(x), x \rangle \\ &= \langle UQ(x), Q(x) \rangle \\ &> 0\end{aligned}$$

Therefore,  $TU$  is positive definite.

□

- f.  $T$  is positive definite [semidefinite] if and only if  $A$  is positive definite [semidefinite]. Because of (f), results analogous to items (a) through (d) hold for matrices as well as operators.

*Proof.* Let  $\beta = \{v_1, \dots, v_n\}$  be an orthonormal basis for  $V$ . Let  $a = (a_1, \dots, a_n)$  be a nonzero  $n$ -tuple. We can define  $x = \sum_{i=1}^n a_i v_i \in V$ . It follows that by corollary of Theorem



6.5 and the definition of positive definite,

$$\begin{aligned}
 \langle T(x), x \rangle &= \left\langle T\left(\sum_{j=0}^n a_j v_j\right), \sum_{i=0}^n a_i v_i \right\rangle \\
 &= \sum_{j=0}^n a_j \left\langle T(v_j), \sum_{i=0}^n a_i v_i \right\rangle \\
 &= \sum_{i,j=0}^n a_j \overline{a_i} \langle T(v_j), v_i \rangle \\
 &= \sum_{i,j=0}^n A_{ij} \overline{a_i} a_j \\
 &= \langle Ax, x \rangle
 \end{aligned}$$

as desired. □

## §6 F

Let  $T$  and  $U$  be positive definite operators on an inner product space  $V$ . Prove the following results.

- a.  $T + U$  is positive definite

*Proof.* By definition,  $\langle T(x), x \rangle > 0$ ,  $\langle U(x), x \rangle > 0$  for all  $x \neq 0$

$$\begin{aligned}
 \langle (T + U)(x), x \rangle &= \langle T(x) + U(x), x \rangle \\
 &= \langle T(x), x \rangle + \langle U(x), x \rangle \\
 &> 0
 \end{aligned}$$

Therefore,  $T + U$  is positive definite. □

- b. If  $c > 0$ , then  $cT$  is positive definite.

*Proof.* By definition,  $\langle T(x), x \rangle > 0$ . It follows that

$$\begin{aligned}
 \langle cT(x), x \rangle &= c \langle T(x), x \rangle \\
 &> 0
 \end{aligned}$$

Therefore, if  $c > 0$ , then  $cT$  is positive definite. □

- c.  $T^{-1}$  is positive definite.

*Proof.* Let  $y = T^{-1}(x)$ . We know that  $T^* = T$  because  $T$  is self adjoint.

$$\begin{aligned}
 \langle T^{-1}(x), x \rangle &= \langle y, T(y) \rangle \\
 &= \langle T^*(y), y \rangle \\
 &= \langle T(y), y \rangle \\
 &> 0
 \end{aligned}$$

□

## §7 G

Prove that if  $T$  is a unitary operator on a finite-dimensional inner product space  $V$ , then  $T$  has a unitary square root; that is, there exists a unitary operator  $U$  such that  $T = U^2$ .

*Proof.* By Theorem 6.18, there exists an orthonormal basis  $\beta = \{v_1, \dots, v_n\}$  consisting of eigenvectors of  $T$  with corresponding eigenvalues  $\{\lambda_1, \dots, \lambda_n\}$  of absolute value 1. It follows by Theorem 5.1 that

$$[T]_{\beta} = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$$

We can fix  $[U]_{\beta}$  as  $\sqrt{[T]_{\beta}}$

$$[U]_{\beta} = \begin{bmatrix} \sqrt{\lambda_1} & & 0 \\ & \ddots & \\ 0 & & \sqrt{\lambda_n} \end{bmatrix}$$

$U$  is a unitary operator by Theorem 6.18 because.

$$|\sqrt{\lambda_i}| = \sqrt{|\lambda_i|} = \sqrt{1} = 1$$

Therefore, there exists a unitary operator  $U$  such that  $T = U^2$ . □

## §8 H

Let  $A$  and  $B$  be  $n \times n$  matrices that are unitarily equivalent.

- a. Prove that  $\text{tr}(A^*A) = \text{tr}(B^*B)$

*Proof.* By definition, there exists a unitary matrix  $P$  such that  $A = P^*BP$ . It follows that

$$\begin{aligned} \text{tr}(A^*A) &= \text{tr}((P^*BP)^*(P^*BP)) \\ &= \text{tr}(P^*B^*PP^*BP) \\ &= \text{tr}((P^*B^*)(BP)) \\ &= \text{tr}((BP)(P^*B^*)) \\ &= \text{tr}(BB^*) \\ &= \text{tr}(B^*B) \end{aligned}$$

as desired. □

- b. Use (a) to prove that

$$\sum_{i,j=1}^n |A_{ij}|^2 = \sum_{i,j=1}^n |B_{ij}|^2$$

*Proof.* Notice that

$$\begin{aligned}
 \sum_{i,j=1}^n |A_{ij}|^2 &= \sum_{i,j=1}^n |A_{ij}|^2 \\
 &= \sum_{j=1}^n \sum_{i=1}^n \overline{A_{ij}} A_{ij} \\
 &= \sum_{j=1}^n \sum_{i=1}^n (A^*)_{ji} A_{ij} \\
 &= \sum_{i=1}^n (A^* A)_{ii} \\
 &= \text{tr}(A^* A)
 \end{aligned}$$

Without loss of generality,

$$\sum_{i,j=1}^n |B_{ij}|^2 = \text{tr}(B^* B)$$

Therefore, by (a)

$$\sum_{i,j=1}^n |A_{ij}|^2 = \sum_{i,j=1}^n |B_{ij}|^2$$

□

c. Use (b) to show that the matrices

$$A = \begin{bmatrix} 1 & 2 \\ 2 & i \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} i & 4 \\ 1 & 1 \end{bmatrix}$$

are not unitarily equivalent.

*Proof.* Consider,

$$\begin{aligned}
 \sum_{i,j=1}^n |A_{ij}|^2 &= |1|^2 + |2|^2 + |2|^2 + |i|^2 \\
 &= 10
 \end{aligned}$$

and

$$\begin{aligned}
 \sum_{i,j=1}^n |B_{ij}|^2 &= |i|^2 + |4|^2 + |1|^2 + |1|^2 \\
 &= 19
 \end{aligned}$$

Therefore,  $A$  and  $B$  are not unitarily equivalent.

□