## SECOND MIDTERM SOLUTION MATH H54

(1) (15 points) Consider the symmetric matrix

$$A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}.$$

Find an orthogonal matrix P and a diagonal matrix D such that  $A = PDP^T$ . (You have to write down every steps of your calculations, not just the final answer.)

Solution.

$$\begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} -1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 0 & 0 & 1 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 \end{bmatrix} \begin{bmatrix} -1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 0 & 0 & 1 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 \end{bmatrix}^{T}.$$

(2) Let  $M_2(\mathbb{R})$  be the set of all real  $2 \times 2$  matrices. It is a vector space with the standard matrix addition and scalar multiplication. Consider the function  $\langle -, - \rangle : M_2(\mathbb{R}) \times M_2(\mathbb{R}) \to \mathbb{R}$  given by

$$\langle A, B \rangle := \operatorname{tr}(AB^T),$$

where  $A, B \in M_2(\mathbb{R})$  and tr denotes the trace function. It is not hard to check that  $\langle -, - \rangle$  gives an inner product on  $M_2(\mathbb{R})$ .

- (a) (15 points) Construct an orthonormal basis (with respect to  $\langle -, \rangle$ ) for the subspace of  $M_2(\mathbb{R})$  spanned by  $\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$  and  $\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$ .
- (b) (5 points) Consider another function  $\langle -, \rangle_2 : M_2(\mathbb{R}) \times M_2(\mathbb{R}) \to \mathbb{R}$  given by  $\langle A, B \rangle_2 := \operatorname{tr}(AB)$ .

Does  $\langle -, - \rangle_2$  give an inner product on  $M_2(\mathbb{R})$  as well? Prove your answer.

Solution. (a)

$$\left\{ \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 1/\sqrt{3} & 1/\sqrt{3} \\ 0 & 1/\sqrt{3} \end{bmatrix} \right\}.$$

(b) No.

$$\left\langle \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \right\rangle_2 = -2 < 0.$$

- (3) True or False. For each of the following statements, either prove the statement, or give an explicit counterexample.
  - (a) (5 points) Let A be a square matrix. "If  $A^2$  is diagonalizable, then so is A."

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- (b) (5 points) Let A be a square matrix. "If A is diagonalizable, then so is  $A^2$ ."
- (c) (15 points) "There does not exist an orthogonal matrix such that 2 is one of its eigenvalues."

**Solution.** (a) False. Counterexample:  $A = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$ .

- (b) True. If  $A = PDP^{-1}$ , then  $A^2 = PD^2P^{-1}$ .
- (c) True. Let A be an orthogonal matrix. Then  $||A\vec{v}|| = ||\vec{v}||$  for any  $\vec{v}$ . Hence any eigenvalue  $\lambda \in \mathbb{C}$  of A satisfies  $|\lambda| = 1$ .
- (4) (20 points) Let  $(V, \langle -, \rangle)$  be an inner product space, and let  $T: V \to V$  be a linear transformation. Suppose that  $||T(\vec{x})|| = ||\vec{x}||$  for any  $\vec{x} \in V$ . Prove that

$$\langle T(\vec{x}), T(\vec{y}) \rangle = \langle \vec{x}, \vec{y} \rangle$$
 for any  $\vec{x}, \vec{y} \in V$ .

(Hint: Consider  $||T(\vec{x} + \vec{y})||^2 = ||\vec{x} + \vec{y}||^2$ .)

**Solution.**  $||\vec{x} + \vec{y}||^2 = \langle \vec{x} + \vec{y}, \vec{x} + \vec{y} \rangle = ||\vec{x}||^2 + ||\vec{y}||^2 + 2\langle x, y \rangle$ . Similarly, we have  $||T(\vec{x} + \vec{y})|^2 = ||\vec{x}||^2 + ||\vec{y}||^2 + 2\langle x, y \rangle$ .  $|\vec{y}||^2 = ||T(\vec{x})||^2 + ||T(\vec{y})||^2 + 2\langle T(x), T(y) \rangle$ . The assumption that  $||T(\vec{x})|| = ||\vec{x}||$  for any  $\vec{x} \in V$  therefore implies that  $\langle T(\vec{x}), T(\vec{y}) \rangle = \langle \vec{x}, \vec{y} \rangle$  for any  $\vec{x}, \vec{y} \in V$ .

(5) (20 points) Let  $A_1, \ldots, A_k$  be  $n \times n$  real symmetric matrices. Suppose that  $A_1^2 + a_1 + a_2 + a_3 + a_4 + a_4$  $\cdots + A_k^2 = 0$  (the zero matrix). Prove that  $A_1 = \cdots = A_k = 0$  (the zero matrix). (Hint:

Consider  $\vec{x}^T(A_1^2 + \dots + A_k^2)\vec{x}$ .) **Solution.** We have  $0 = \vec{x}^T(A_1^2 + \dots + A_k^2)\vec{x} = \vec{x}^TA_1^TA_1\vec{x} + \dots \vec{x}^TA_k^TA_k\vec{x} = ||A_1\vec{x}||^2 + \dots + |A_k^2||^2$  $||A_k \vec{x}||^2$  for any  $\vec{x}$ . Hence  $A_i \vec{x} = 0$  for any  $1 \le i \le k$  and  $\vec{x}$ . Thus  $A_i = 0$  for any i.

(6) (20 points) Let A be an  $n \times n$  diagonalizable matrix with n-1 distinct eigenvalues. Prove that for any  $\vec{v} \in \mathbb{R}^n$ , the set  $\{\vec{v}, A\vec{v}, \dots, A^{n-1}\vec{v}\}$  is linearly dependent.

**Solution.** Write  $A = PDP^{-1}$ . Note that the statements "the set  $\{\vec{v}, A\vec{v}, \dots, A^{n-1}\vec{v}\}$  is linearly dependent for any  $\vec{v}$ " and "the set  $\{P\vec{v}, PD\vec{v}, \dots, PD^{n-1}\vec{v}\}$  is linearly dependent for any  $\vec{v}$ " are equivalent (why?). Also, since P is invertible, the above statement is equivalent to "the set  $\{\vec{v}, D\vec{v}, \dots, D^{n-1}\vec{v}\}$  is linearly dependent for any  $\vec{v}$ " (why?). Let  $\vec{v} = \begin{bmatrix} v_1 & \cdots & v_n \end{bmatrix}^T$  and let  $\lambda_1, \dots, \lambda_n$  be the diagonal entries of D. Consider the

 $n \times n$  matrix with columns  $\{\vec{v}, D\vec{v}, \dots, D^{n-1}\vec{v}\}$ :

$$\begin{bmatrix} v_1 & \lambda_1 v_1 & \cdots & \lambda_1^{n-1} v_1 \\ v_2 & \lambda_2 v_2 & \cdots & \lambda_2^{n-1} v_2 \\ \vdots & \vdots & & \vdots \\ v_n & \lambda_n v_n & \cdots & \lambda_n^{n-1} v_n \end{bmatrix}.$$

Since A has repeat eigenvalues, so this matrix has linearly dependent rows, hence not invertible. Therefore the set  $\{\vec{v}, D\vec{v}, \dots, D^{n-1}\vec{v}\}$  is linearly dependent for any  $\vec{v}$ .