# Solution to Homework 2

# MA502 Linear Algebra

#### Problem 1.

Exercise 1 on P83 of Roman

Let  $A \in \mathcal{M}_{m,n}$  have rank k. Prove that there are matrices  $X \in \mathcal{M}_{m,k}$  and  $y \in \mathcal{M}_{k,n}$ , both of rank k, for which A = XY. Prove that A has rank 1 if and only if it has form  $A = x^t y$  where x and y are row matrices.

*Proof.* Suppose  $A \in \mathcal{M}_{m,n}$  have rank k, and A can be written as

$$A = \begin{bmatrix} v_1 & v_2 & \dots & v_k & v_{k+1} & \dots & v_n \end{bmatrix}$$

where  $v_i$  is a  $m \times 1$  vector. Meanwhile, we know that rank of its column space is also k. Hence, WOLOG, we can suppose  $v_1 \ldots v_k$  are linear independent, and  $v_i(i > k)$  can be represented as a linear combination of  $v_1 \ldots v_k$ . (Otherwise, we can multiply column exchange matrices to A, and they will not affect rank of a matrix.) Suppose for  $j \ge k+1$ ,

$$v_j = \sum_{i=1}^k a_i^j v_i$$

Then define

$$V = \begin{bmatrix} v_1 & v_2 & \dots & v_k \end{bmatrix}$$
$$W = \begin{bmatrix} 1 & \dots & 0 & a_1^{k+1} & \dots & a_i^n \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 1 & a_k^{k+1} & \dots & a_k^n \end{bmatrix}$$

Obviously, rank(V) = rank(W) = k.  $V \in \mathcal{M}_{m,k}$ ,  $W \in \mathcal{M}_{k,n}$ , and A = VW. For the second part, we need to prove A has rank 1 iff  $A = x^T y$ , where, x, y are row vectors. Suppose x, y are row vectors, and  $A = x^T y$ . Suppose  $y = \begin{bmatrix} y_1 & y_2 & \dots & y_n \end{bmatrix}$ . Then A =

 $\begin{bmatrix} y_1 x^T & y_2 x^T & \dots & y_n x^T \end{bmatrix}$ , which obviously has rank 1. Then, suppose A has rank 1, and then A can be represented as

$$A = \begin{bmatrix} k_1 x_1 & k_2 x_1 & \dots & k_n x_1 \end{bmatrix}$$

where  $x_1$  is  $m \times 1$  vector, and  $k_i \in \mathbb{F}$ . Then,

$$A = x_1 \begin{bmatrix} k_1 & \dots & k_n \end{bmatrix}$$
  
=  $x_1 y$   
=  $(x_1^T)^T y$  (1)

where  $x_1^T$ , y are row matrix, and  $y = \begin{bmatrix} k_1 & \dots & k_n \end{bmatrix}$ . Hence, A has rank 1 iff  $A = x^T y$ , where, x, y are row vectors.

## Problem 2.

Exercise 9 on P83 of Roman Let  $\tau \in \mathcal{L}(V)$  where  $dim(V) < \infty$ . If  $rank(\tau^2) = rank(\tau)$  show that  $Im(\tau) \cap ker(\tau) = \{0\}$ 

Proof. Since  $Im(\tau^2) \subseteq Im(\tau)$ , and  $rank(\tau^2) = rank(\tau)$ , we can get  $Im(\tau^2) = Im(\tau)$ . By Rank-Nullity Theorem, we can get  $dim(ker(\tau^2)) = dim(ker(\tau))$ . Meanwhile,  $ker(\tau) \subseteq ker(\tau^2)$ , and then  $ker(\tau^2) = ker(\tau)$ .

Suppose  $x \in Im(\tau) \cap ker(\tau)$ . Then,  $\exists y \in V$  s.t.  $x = \tau y$ , and  $\tau x = 0$ . We can have  $\tau^2 y = 0$ , and  $y \in ker(\tau^2)$ . Since we have proven above that  $ker(\tau^2) = ker(\tau)$ ,  $y \in ker(\tau)$ , and  $\tau y = 0$ , which means x = 0.

This proves that  $\{0\} = Im(\tau) \cap ker(\tau)$ .

#### Problem 3.

Exercise 13 on P84 of Roman Let  $\tau, \sigma \in \mathcal{L}$ . Show that  $rank(\tau + \sigma) \leq rank(\tau) + rank(\sigma)$ .

Proof. Suppose A is matrix of  $\tau$ , and B is matrix of  $\sigma$ . Then, it is suffice to prove  $rank(A+B) \leq rank(A) + rank(B)$ . Denote  $C_A$  as collection of column vectors of A. Clearly,  $span(C_{A+B}) \subseteq span(C_A \cup C_B)$ , since  $\forall w \in C_{A+B}, \exists v_1 \in C_A, v_2 \in C_B$  s.t  $w = v_1 + v_2$ .

$$rank(A + B) \leq dim(C_A) + dim(C_B) - dim(C_A \cup C_B)$$
  
$$\leq dim(C_A) + dim(C_B)$$
  
$$= rank(A) + rank(B)$$
(2)

### Problem 4.

Consider the operator  $\tau$  on  $\mathcal{C}^{\infty}(\mathbb{R})$  defined by  $\tau: f(x) \to f'(x)$  (first derivative). Describe all eigenvectors of operator  $\tau$ . Explain.

*Proof.* We claim that all eigenvectors (eigenfunctions) are in the following form:  $f(x) = f(0)e^{\lambda x}$ .

We know that  $\tau f = f'$ . We want to find eigenpairs by calculating  $\tau f = \lambda f'$ , where  $\lambda$  is the eigenvalue. Hence,  $f' = \lambda f \Rightarrow f(x) = f(0)e^{\lambda x}$ , where  $\lambda \in \mathbb{R}$ .

Problem 5.

a

Construct an example  $A \in \mathcal{M}_{3\times 3}(\mathbb{C})$  with exactly one eigenvector up to equivalence under scalar multiplication. Choose A so that the corresponding eigenvalue is *i*.

*Proof.* 
$$A = \begin{bmatrix} i & 1 & 0 \\ 0 & i & 1 \\ 0 & 0 & i \end{bmatrix}$$
. Eigenvalue is *i*, and eigenvector is  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ .

b

Construct, or prove non-existence of  $A \in \mathcal{M}_{3\times 3}(\mathbb{C})$  with no eigenvectors.

*Proof.* We claim that every square matrix with size n in  $\mathbb{C}$  has at least one eigenvector. Suppose that A has size n, and choose  $x \in \mathbb{C}^n$  that is not zero. Define

$$S = \{x, Ax, A^2x, \dots, A^nx\}$$

Since S is a set of n + 1 vectors in  $\mathbb{C}^n$ , it is linearly dependent. Suppose

$$a_0x + a_1Ax + a_2A^2x + \ldots + a_nA^nx = 0$$

where  $a_i \in \mathbb{C}$ . By linearly dependence of vectors in S, there exists some  $a_i \neq 0$ . We need to prove that there are more than one  $a_i \neq 0$ . WOLOG, suppose  $a_0 \neq 0$ . If  $a_1 = \ldots = a_n = 0$ . Then  $a_0x = 0$ . We can get either  $a_0 = 0$  or x = 0. Contradiction! Let m be the largest integer s.t.  $a_m \neq 0$ , where  $m \geq 1$ . WOLOG, we can suppose  $a_m = 1$ . Define

$$p(x) = a_0 + a_1 x + \ldots + a_m x^m$$

Then  $b_1, b_2, \ldots, b_m \in \mathbb{C}$  s.t.

$$p(x) = (x - b_m)(x - b_{m-1})\dots(x - b_1)$$

Then, we get

$$0 = a_0 x + a_1 A x + \dots + a_n A^n x$$
  
=  $a_0 x + a_1 A x + \dots + a_m A^m x$   
=  $(a_0 I_n + a_1 A + \dots + a_m A^m) x$  (3)  
=  $p(A) x$   
=  $(A - b_m I_n) (A - b_{m-1} I_n) \dots (A - b_1 I_n) x$ 

Let k be the smallest integer s.t.

$$(A - b_k I_n)(A - b_{k-1} I_n) \dots (A - b_1 I_n)x = 0$$

We know that  $k \leq m$ . Define

$$z = (A - b_{k-1}I_n)\dots(A - b_1I_n)x$$

Then,  $z \neq 0$ . If k = 1, then define z = x, which is still not zero. Then

$$(A - b_k I_n)z = (A - b_k I_n)(A - b_{k-1} I_n)\dots(A - b_1 I_n)x = 0$$

It gives that  $Az = b_k z$ . Since  $z \neq 0$ , it show that A must have at least one eigenvector, as well as eigenvalue.

#### С

In part (a), what are the other eigenvalues of your matrix A? Why do I know them already?

*Proof.* There is no other eigenvalue of matrix A. This is because different eigenvalues should have different eigenvectors, which will be proven in Question 8. Since there is only one eigenvector, then there should be only one eigenvalue.

## Problem 6.

Write each of the following matrices as a linear combination of projection matrices:

a

Proof.

We can get its eigenvalues: 0 and 4. Consider  $\lambda = 4$ , and then eigenvector is  $V = \begin{bmatrix} 1\\1\\1 \end{bmatrix}$ .

 $\mathbf{b}$ 

Proof.

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

We can get its eigenvalues: -2, 0, 2. Consider  $\lambda = -2$ . Its eigenvector  $V_{-2} = \begin{bmatrix} 1 \\ -1 \\ -1 \\ 1 \end{bmatrix}$ . Then

its projection matrix

Hence,  $B = (-2)P_{-2} + 2P_2$ .

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•	

Proof.

$$C = \begin{bmatrix} 1 & 2 & 2 & 3 \\ 2 & 1 & 3 & 2 \\ 2 & 3 & 1 & 2 \\ 3 & 2 & 2 & 1 \end{bmatrix}$$

We can get its eigenvalues: -2, 0, 8. Consider  $\lambda = -2$ , and its eigenspace is  $span \left\{ \begin{bmatrix} 1\\0\\0\\-1 \end{bmatrix}, \begin{bmatrix} 0\\-1\\1\\0 \end{bmatrix} \right\}$ .

Consider

$$V_{-2} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \\ 0 & 1 \\ -1 & 0 \end{bmatrix}$$

Then, projection matrix

$$P_{-2} = V_{-2}(V_{-2}^T V_{-2})^{-1} V_{-2}^T = \frac{1}{2} \begin{bmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & -1 & 0 \\ 0 & -1 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix}$$

#### Problem 7.

Suppose P and Q are  $n \times n$  projection matrices with real entries such that PQ = 0. What can you say about their column spaces? Explain.

Proof. Denote  $C_A$  as collection of column vectors of A. Then column space of A is  $span(C_A)$ . We claim that  $span(C_P) \oplus span(C_Q) \subseteq \mathcal{M}_{n \times n}(\mathbb{R})$ . This means: 1.  $span(C_P) \cap span(C_Q) = \{0\}$ 

2. Union of two column spaces should be subspace of  $\mathcal{M}_{n \times n}(\mathbb{R})$ .

First, let us prove  $\mathcal{M}_{n\times n}(\mathbb{R}) = span(C_P) \oplus ker(P)$ . Suppose  $x \in span(C_P) \cap ker(P)$ . Then there exists a such that Px = 0, and x = Pa, so  $0 = Px = P^2a = Pa = x$ . Hence,  $span(C_W) \cap ker(W) = \{0\}$ . Meanwhile,  $dim(span(C_W)) + dim(ker(W)) = n$ , and hence  $\mathcal{M}_{n\times n}(\mathbb{R}) = span(C_P) \oplus ker(P)$ . It is obvious that Px = 0 iff  $x \in ker(P)$ . Since PQ = 0, every vector in  $C_Q$  must be in ker(P), and hence,  $span(C_Q) \subseteq ker(P)$ . Hence,  $span(C_P) \oplus span(C_Q) \subseteq \mathcal{M}_{n\times n}(\mathbb{R})$ . Furthermore, if PQ = QP = 0 (which may not be true), by conversing P and Q in the above proof, we can get  $span(C_P) \oplus span(C_Q) = \mathcal{M}_{n\times n}(\mathbb{R})$ .

#### Problem 8.

Let  $\tau$  be a linear operator on vector space V over  $\mathbb{F}$  and suppose  $v_1, \ldots, v_k$  are eigenvectors belonging to distinct eigenvalues:  $\tau v_j = \lambda_j v_j$  with  $\lambda_j \neq \lambda_l$  unless j = l. Prove that the set  $S = \{v_1, \ldots, v_k\}$  is linearly independent in V.

*Proof.* Suppose A is matrix of  $\tau$ . Since  $v_i \in N(A - \lambda_i I)$ , it is suffice to prove for  $\forall i$ 

$$N(A - \lambda_i I) \bigcap \sum_{j \neq i}^k N(A - \lambda_j I) = \{0\}$$

We can prove by induction. First, we can prove that it holds when k = 2. Suppose  $x \in N(A - \lambda_1 I) \cap N(A - \lambda_2 I)$ . Then,

$$0 = (A - \lambda_1 I)x = (A - \lambda_2 I)$$
  

$$\Rightarrow (\lambda_1 - \lambda_2)x = 0$$

$$\Rightarrow x = 0$$
(4)

Last line is because  $\lambda_1 \neq \lambda_2$ . Hence,  $N(A - \lambda_1 I) \cap N(A - \lambda_2 I) = \{0\}$ . Then, suppose that it hold when k = n. When k = n + 1. Suppose  $x \in N(A - \lambda_{n+1}I) \cap \sum_{i=1}^n N(A - \lambda_j I)$ . There exists  $v_1, \ldots, v_n \in \bigcup_{j=1}^n N(A - \lambda_j I)$  such that  $x = v_1 + \ldots + v_n$ . WOLOG, suppose  $v_i \in N(A - \lambda_i I)$ . Then,

$$0 = (A - \lambda_{n+1}I)x = (A - \lambda_{n+1}I)(v_1 + \dots + v_n)$$

Since  $(A - \lambda_i I)v_i = 0$ , for i = 1, 2, ..., n. We can have

$$(\lambda_{n+1} - \lambda_1)v_1 + \ldots + (\lambda_{n+1} - \lambda_n)v_n = 0$$

If there exists some  $v_i \neq 0$ , since  $\{v_1, \ldots, v_n\}$  are linearly independent, there must exist some  $\lambda_i = \lambda_{n+1}$ . Contradiction! Hence, all  $v_i$  is zero, and x = 0. The result holds when k = n+1. Hence,  $N(A - \lambda_i I) \bigcap \sum_{j \neq i}^k N(A - \lambda_j I) = \{0\}$ , for  $\forall i$ . S is linearly independent in V.