# Topics in Linear Algebra: Homework 2

## March 18, 2022

In the following, "linearly independent" is used interchangeably with "independent", and "linearly dependent" is used interchangeably with "dependent".

# Solution 1.2.1.

#### 1. True.

Suppose that any three of the four given subspaces are linearly independent, but the four subspaces are not. Then assume

$$v_i \in V_i, i = 1, 2, 3, 4$$

Since the four subspaces are dependent, W.L.O.G. assume  $\dim(V_4 \cap (V_1 + V_2 + V_3)) > 0$ , then

$$v_4 = -(v_1 + v_2 + v_3)$$

So  $||v_1|| + ||v_2|| + ||v_3|| > 0$ , that means  $V_4$  has intersection with at least one subspace, contradicting the assumption that  $V_4$  and any other two of the three spaces are linearly independent.

#### 2. False.

Consider

$$V_1 = \operatorname{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\}, V_2 = \operatorname{span} \left\{ \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}, V_3 = \operatorname{span} \left\{ \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}, V_4 = \operatorname{span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \right\}$$

Here  $V_4 = V_1 + V_2$ , so  $V_1, V_2, V_4$  are linearly dependent.

## 3. True.

Assume

$$v_i \in V_i, i = 1, 2, 3, 4.$$

If the four subspaces are dependent, then

$$v_1 + v_2 + v_3 + v_4 = 0$$

has non-trivial solutions. Say

$$v_4 = -(v_1 + v_2 + v_3)$$

Hence  $||v_1|| + ||v_2|| + ||v_3|| > 0$ . So at least one of the three vectors is non-zero. If  $v_3$  is non-zero, then it contradicts that " $V_3$ ,  $V_4$  are independent". If  $v_1$  or  $v_2$  is non-zero, then

- (a) If  $v_1 + v_2 = 0$ , then  $v_1 = -v_2$ , and both of which are non-zero. It violates that " $V_1, V_2$  are independent".
- (b) Else, since  $v_4 \in (V_3 + V_4)$ , and  $v_1 + v_2 \neq 0$ , then  $(V_1 + V_2) \cap (V_3 + V_4)$  is a non-trivial space, which is another contradiction.

Switch  $v_4$  to the vectors from the three other subspaces, and the same conclusion would be drawn.

# Solution 1.2.2.

By definition,

$$T:A\mapsto A^T$$

and

$$T^2(A) = (A^T)^T = A.$$

By observation,

A is symmetric 
$$\Leftrightarrow T(A) = A$$

A is skew-symmetric 
$$\Leftrightarrow T(A) = -A$$
.

T is bijective, so if a matrix  $A_0$  is both symmetric and anti-symmetric, then

$$T(A_0) + T(A_0) = A_0 + (-A_0) = 0 \Rightarrow A_0 = 0$$

Thus, the space of symmetric and skew-symmetric matrices are independent. Denote the former as Sym, and the latter as Skew. Consider

$$A = A_1 + A_2,$$

where

$$A_1 = \frac{1}{2}(A + T(A)), A_2 = \frac{1}{2}(A - T(A))$$

$$T(A_1) = \frac{1}{2}T(A + T(A)) = \frac{1}{2}(T(A) + T^2(A)) = \frac{1}{2}(T(A) + A) \in \text{Sym}$$

$$T(A_2) = \frac{1}{2}T(A - T(A)) = \frac{1}{2}(T(A) - T^2(A)) = \frac{1}{2}(T(A) - A) \in \text{Skew}$$

So every matrix can be expressed as a linear combination of two matrices, one from Sym, and one from Skew. Hence

$$V = \operatorname{Sym} \bigoplus \operatorname{Skew}$$

In particular, the former has dimension  $\frac{1}{2}n(n+1)$ , and the latter has  $\frac{1}{2}n(n-1)$ .

With regards to block form of T,

$$T = \left[ \begin{array}{cc} e_1 & \\ & -e_2 \end{array} \right],$$

where  $e_1$  is the block representing the transformation:

 $T_1: \operatorname{Sym} \to \operatorname{Sym}, T_1$  is identity transformation

 $e_2$  is the block representing the transformation:

 $T_2: \text{Skew} \to \text{Skew}, T_2 \text{ is identity transformation}$ 

## Solution 1.2.3.

1. Ker(B):

 $\forall v \in \text{Ker}(B),$ 

$$B(Av) = BAv = ABv = A(Bv) = A \cdot 0 = 0,$$

so  $v \in \text{Ker}(B) \Rightarrow Av \in \text{Ker}(B) \square$ 

Ran(B):

 $\forall w \in \text{Ran}(B), \exists v \text{ s.t. } Bv = w.$ 

$$A(w) = A(Bv) = ABv = BAv = B(Av),$$

so  $w \in \text{Ran}(B) \Rightarrow Aw \in \text{Ran}(B)$ .

2.

$$Ap(A) = A \cdot \sum_{i=0}^{n-1} p_i A^i = \sum_{i=0}^{n-1} p_i A^{i+1} = (\sum_{i=0}^{n-1} p_i A^i) A = p(A) A$$

3. Let k be the smallest positive integer satisfies  $\operatorname{Dim}(\operatorname{Ker}((A-\lambda I)^k))=n$ , where  $k\leq n$ . Then  $\operatorname{Ker}((A-\lambda I)^k)=N_\infty(A-\lambda I)$ ,  $\operatorname{Ran}((A-\lambda I)^k)=R_\infty(A-\lambda I)$ . As

$$(A - \lambda I)^{k} = \sum_{i=0}^{k} \binom{k}{i} A^{i} (-\lambda)^{k-i}$$

is a polynomial of A, by (2)  $A(A - \lambda I)^k = (A - \lambda I)^k A$ , and by (1), and by the fact that  $\lambda \in \mathbb{C}$  has no restriction, the conclusion is drawn.

# Solution 1.2.4.

1. Suppose W is an A-invariant subspace, but none of the eigenvectors of A belongs to W. Then  $\forall \lambda \in \mathbb{C}, \forall v \in W, v \neq 0$ 

$$Av \neq \lambda v \Leftrightarrow (A - \lambda I)v \neq 0$$
  
  $\Leftrightarrow |A|_W - \lambda I| \neq 0$ 

Hence

$$p_{A|_W}(\lambda)$$

has no roots. By the fundamental theorem of algebra,  $\deg p_{A|_W}=0$ . Since  $p_A(\lambda)=p_{A|_W}(\lambda)\cdot p_{A|_{W^\perp}}(\lambda)$ ,  $\dim(W)=\deg p_{A|_W}=0 \Rightarrow W=\{0\}$ , which contradicts the assumption that  $v\neq 0$ .  $\square$ 

2.

$$B(A - \lambda I) = BA - \lambda B = AB - \lambda B = (A - \lambda I)B$$
,

so by 1.2.3.1.  $\forall \lambda \in \mathbf{C} : \text{Ker}(A - \lambda I)$  is B-invariant.  $\square$ 

3. In (1), let  $W = \text{Ker}(A - \lambda_1 I)$ , and fix  $\lambda_1$  as an eigenvalue of A, then B has an eigenvector in  $\text{Ker}(A - \lambda_1 I)$ , which is the  $\lambda_1$ -eigenspace of A, hence that eigenvector of B is also an eigenvector of A.