

# MATH1241 Problem Set Solutions - Algebra

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## Chapter 6

# Vector Spaces

### 6.4 Problem 4

#### Closure under Addition

For any vectors  $\mathbf{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}, \mathbf{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \in \mathbb{C}^n$ , because  $\mathbb{C}$  is closed under addition, we have

$$\mathbf{u} + \mathbf{v} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix} + \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} = \begin{pmatrix} u_1 + v_1 \\ \vdots \\ u_n + v_n \end{pmatrix} \in \mathbb{C}^n.$$

Thus,  $\mathbb{C}^n$  is closed under addition.

#### Associative Law of Addition

For any vectors  $\mathbf{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}, \mathbf{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}, \mathbf{w} = \begin{pmatrix} w_1 \\ \vdots \\ w_n \end{pmatrix} \in \mathbb{C}^n$ , because  $\mathbb{C}$  is associative, we have

$$\begin{aligned} (\mathbf{u} + \mathbf{v}) + \mathbf{w} &= \left( \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix} + \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \right) + \begin{pmatrix} w_1 \\ \vdots \\ w_n \end{pmatrix} = \begin{pmatrix} u_1 + v_1 + w_1 \\ \vdots \\ u_n + v_n + w_n \end{pmatrix} = \begin{pmatrix} u_1 + (v_1 + w_1) \\ \vdots \\ u_n + (v_n + w_n) \end{pmatrix} \\ &= \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix} + \begin{pmatrix} v_1 + w_1 \\ \vdots \\ v_n + w_n \end{pmatrix} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix} + \left( \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} + \begin{pmatrix} w_1 \\ \vdots \\ w_n \end{pmatrix} \right) = \mathbf{u} + (\mathbf{v} + \mathbf{w}). \end{aligned}$$

Thus,  $\mathbb{C}^n$  is associative.

## Closure under Multiplication by a Scalar

For any vector  $\mathbf{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \in \mathbb{C}^n$  and any scalar  $\lambda \in \mathbb{C}$ , since  $\mathbb{C}$  is closed under multiplication by a scalar, we have

$$\lambda \mathbf{v} = \lambda \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} = \begin{pmatrix} \lambda v_1 \\ \vdots \\ \lambda v_n \end{pmatrix} \in \mathbb{C}^n.$$

Thus,  $\mathbb{C}^n$  is closed under multiplication by a scalar.

## Scalar Distributive Law

For any vector  $\mathbf{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} \in \mathbb{C}^n$  and any scalar  $\lambda, \mu \in \mathbb{C}$ , due to the scalar distributive law of  $\mathbb{C}$ , we have

$$(\lambda + \mu)\mathbf{v} = (\lambda + \mu) \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix} = \begin{pmatrix} (\lambda + \mu)v_1 \\ \vdots \\ (\lambda + \mu)v_n \end{pmatrix} \in \mathbb{C}^n.$$

Thus, the scalar distributive law holds for  $\mathbb{C}^n$ .

## 6.5 Problem 5

Denote the  $ij$ th entry of  $M$  by  $[M]_{ij}$  where  $1 \leq i \leq m$  and  $1 \leq j \leq n$ . For any matrices  $A, B \in M_{mn}(\mathbb{C})$  and scalars  $\lambda \in \mathbb{C}$ .

For axiom 1, because  $\mathbb{C}$  is closed under addition, we have

$$[A]_{ij} + [B]_{ij} \in \mathbb{C}, \quad \text{for all } i, j.$$

Hence,  $A + B \in M_{mn}(\mathbb{C})$ , which shows that axiom 1 is satisfied.

For axiom 3, because  $\mathbb{C}$  is commutative, we have

$$[A]_{ij} + [B]_{ij} = [B]_{ij} + [A]_{ij}, \quad \text{for all } i, j.$$

Hence,  $A + B = B + A$ , which shows that axiom 3 is satisfied.

For axiom 6, because  $\mathbb{C}$  is closed under scalar multiplication, we have

$$\lambda[A]_{ij} \in \mathbb{C}, \quad \text{for all } i, j.$$

Hence,  $\lambda A \in M_{mn}(\mathbb{C})$ , which shows that axiom 6 is satisfied.

For axiom 10, because of the distributive law in  $\mathbb{C}$ , we have

$$\lambda([A]_{ij} + [B]_{ij}) = \lambda[A]_{ij} + \lambda[B]_{ij}, \quad \text{for all } i, j.$$

Hence,  $\lambda(A + B) = \lambda A + \lambda B$ , which shows that axiom 10 is satisfied.

## 6.6 Problem 6

It is easy to prove that  $(\mathbb{C}^n, +, *, \mathbb{R})$  is a vector space because  $\mathbb{R}$  is a subfield of  $\mathbb{C}$ . It is also easy to see that  $(\mathbb{R}^n, +, *, \mathbb{C})$  is not a vector space because, for example, the closure under multiplication by a scalar does not hold.

## 6.7 Problem 7

This system is not a vector space.

## 6.8 Problem 8

a)

$$2\mathbf{v} = (1 + 1)\mathbf{v} = 1\mathbf{v} + 1\mathbf{v} = \mathbf{v} + \mathbf{v}.$$

b)

This can be proved by induction.

## 6.9 Problem 9

### Multiplication of the Zero Vector

$$\lambda\mathbf{0} + \mathbf{0} = \lambda\mathbf{0} = \lambda(\mathbf{0} + \mathbf{0}) = \lambda\mathbf{0} + \lambda\mathbf{0}.$$

By the cancellation property, we obtain  $\lambda\mathbf{0} = \mathbf{0}$ .

### Zero Products

If  $\lambda = 0$ , by the property of multiplication of the zero vector, we have  $\lambda\mathbf{v} = \mathbf{0}$ . If  $\lambda \neq 0$ , then  $\lambda^{-1} \neq 0$ , and hence,

$$\mathbf{v} = (\lambda^{-1}\lambda)\mathbf{v} = \lambda^{-1}(\lambda\mathbf{v}) = \lambda^{-1}\mathbf{0} = \mathbf{0}.$$

### Cancellation Property

If  $\lambda\mathbf{v} = \mu\mathbf{v}$ , then  $(\lambda - \mu)\mathbf{v} = \mathbf{0}$ . Since  $\mathbf{v} \neq \mathbf{0}$ , by the property of zero products, we obtain  $\lambda - \mu = 0$ , that is,  $\lambda = \mu$ .

## 6.23 Problem 23

No, because the zero polynomial of  $\mathbb{P}_3$  is not in  $S$ .

## 6.27 Problem 27

a)

Let  $W'$  be the intersection of  $\{W_k : 1 \leq k \leq m + 1\}$ . We prove this by induction.

For  $m = 1$ , we have  $W = W_1$ , which is a subspace of  $V$ .

Suppose that for  $m > 1$ ,  $W$  is a subspace of  $V$ . Then, for  $m + 1$ , since  $W \leq V$  and  $W_{m+1} \leq V$ , we have  $\mathbf{0} \in W$  and  $\mathbf{0} \in W_{m+1}$ , and hence,  $\mathbf{0} \in W'$ . For any vectors  $\mathbf{u}, \mathbf{v} \in W'$  and scalars  $\lambda, \mu \in \mathbb{F}$ , since  $W' = W \cap W_{m+1}$ ,  $\mathbf{u}$  and  $\mathbf{v}$  must be in both  $W$  and  $W_{m+1}$ . Also, since  $W$  and  $W_{m+1}$  are subspaces of  $V$ , they are closed under addition and multiplication by scalars from  $\mathbb{F}$ . Thus,  $\lambda\mathbf{u} + \mu\mathbf{v}$  must be in both  $W$  and  $W_{m+1}$ , and hence in  $W'$ . By the alternative Subspace Theorem,  $W'$  is a subspace of  $V$ .

Therefore, by induction,  $W$  is a subspace of  $V$ .

b)

Suppose that  $W$  is not the set of finite linear combinations of vectors from  $S$ . Then,  $\exists \mathbf{x} \in W$  such that  $\mathbf{x} \notin \text{span}(S)$ . However, for any  $V_i \leq V$  and  $V_i \supseteq S$ , we have  $\text{span}(S) \leq V_i$ , implying that  $\mathbf{x} \notin V_i$ , and hence  $\mathbf{x} \notin W$ , which is a contradiction. Therefore,  $W$  is the set of finite linear combinations of vectors from  $S$ .

### 6.36 Problem 36

This problem is equivalent to proving that  $\text{span}(S)$  is a subspace of  $V$  over field  $\mathbb{F}$ .

Let  $S = \{\mathbf{v}_1, \dots, \mathbf{v}_n\} \subseteq V$ .

The zero vector of  $V$  is in  $\text{span}(S)$  because  $\mathbf{0} = 0\mathbf{v}_1 + \dots + 0\mathbf{v}_n$ .

For any vectors  $\mathbf{u} = \lambda_1\mathbf{v}_1 + \dots + \lambda_n\mathbf{v}_n$ ,  $\mathbf{v} = \mu_1\mathbf{v}_1 + \dots + \mu_n\mathbf{v}_n \in \text{span}(S)$  where  $\lambda_1, \dots, \lambda_n, \mu_1, \dots, \mu_n \in \mathbb{F}$  and any scalar  $\lambda \in \mathbb{F}$ , we have

$$\begin{aligned}\mathbf{u} + \mathbf{v} &= (\lambda_1\mathbf{v}_1 + \dots + \lambda_n\mathbf{v}_n) + (\mu_1\mathbf{v}_1 + \dots + \mu_n\mathbf{v}_n) \\ &= (\lambda_1 + \mu_1)\mathbf{v}_1 + \dots + (\lambda_n + \mu_n)\mathbf{v}_n,\end{aligned}$$

where  $(\lambda_1 + \mu_1), \dots, (\lambda_n + \mu_n) \in \mathbb{F}$ . Thus,  $(\mathbf{u} + \mathbf{v}) \in \text{span}(S)$ , which implies that  $\text{span}(S)$  is closed under addition.

Also, since

$$\begin{aligned}\lambda\mathbf{v} &= \lambda(\mu_1\mathbf{v}_1 + \dots + \mu_n\mathbf{v}_n) \\ &= (\lambda\mu_1)\mathbf{v}_1 + \dots + (\lambda\mu_n)\mathbf{v}_n,\end{aligned}$$

where  $(\lambda\mu_1), \dots, (\lambda\mu_n) \in \mathbb{F}$ , we have  $(\lambda\mathbf{v}) \in \text{span}(S)$ , which implies that  $\text{span}(S)$  is also closed under multiplication by a scalar.

Therefore, by the Subspace Theorem,  $\text{span}(S)$  is a subspace of  $V$ , and hence, the original statement is proved.

### 6.37 Problem 37

We prove this by induction.

For  $n = 1$ ,  $\sum_{k=1}^1 \lambda_k \mathbf{v}_k = \lambda_1 \mathbf{v}_1$ .

Suppose that for  $n > 1$ ,  $\sum_{k=1}^n \lambda_k \mathbf{v}_k = \lambda_1 \mathbf{v}_1 + \dots + \lambda_n \mathbf{v}_n$  holds regardless of the order. By the closures

under addition and multiplication by a scalar,  $\sum_{k=1}^n \lambda_k \mathbf{v}_k$  is in the vector space. Hence, for  $n + 1$ , by

the commutative law of addition, we have  $\sum_{k=1}^{n+1} \lambda_k \mathbf{v}_k = \sum_{k=1}^n \lambda_k \mathbf{v}_k + \lambda_{n+1} \mathbf{v}_{n+1} = \lambda_{n+1} \mathbf{v}_{n+1} + \sum_{k=1}^n \lambda_k \mathbf{v}_k$

regardless of the order.

Therefore, by induction, we proved that we do not need to use brackets when writing down linear combinations.

### 6.46 Problem 46

For

$$\lambda_1 \mathbf{v}_1 + \dots + \lambda_m \mathbf{v}_m = \mathbf{0},$$

multiplying  $\frac{\mathbf{v}_i}{\|\mathbf{v}_i\|^2}$  on both sides of the equation, since  $S$  is orthogonal, we have

$$\lambda_i = \lambda_i \mathbf{v}_i \frac{\mathbf{v}_i}{\|\mathbf{v}_i\|^2} = (\lambda_1 \mathbf{v}_1 + \cdots + \lambda_m \mathbf{v}_m) \frac{\mathbf{v}_i}{\|\mathbf{v}_i\|^2} = \mathbf{0} \cdot \frac{\mathbf{v}_i}{\|\mathbf{v}_i\|^2} = 0, \quad 1 \leq i \leq m,$$

which implies that  $\lambda_1, \dots, \lambda_m$  are all zero. Hence,  $S$  is a linearly independent set.

## 6.52 Problem 52

Because by the Rank-nullity Theorem, the rank cannot exceed the number of columns.

## 6.59 Problem 59

Performing Gaussian Elimination on the matrix whose columns are the vector representations of the polynomials, we have

$$\begin{pmatrix} 1 & 2 & 5 & 0 \\ 1 & -1 & -4 & 0 \\ -1 & 0 & 1 & 1 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & 5 & 0 \\ 0 & -3 & 9 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}.$$

Hence,  $S$  is a linearly dependent spanning set for  $\mathbb{P}_2$ . A subset of  $S$  which is a basis for  $\mathbb{P}_2$  can be  $\{p_1, p_2\}$ .

## 6.60 Problem 60

Since  $\mathbf{w} \in \text{span}(S)$ ,  $\mathbf{w}$  is some linear combination of  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ . Hence, by the definition of linear dependence, the set  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{w}\}$  is a linearly dependent set.

## 6.61 Problem 61

Here gives a rather informal proof. For any subspace  $V$  of  $\mathbb{R}^4$ , we must have  $0 \leq \dim V \leq 4$ . However, the given subspaces have already covered all the subspaces of dimensions from 0 to 4. Hence, the given subspaces are the only subspaces of  $\mathbb{R}^4$ .

## 6.62 Problem 62

$\det(A) = 3$  gives that the columns of  $A$  are linearly independent, and hence form a basis for  $\mathbb{R}^4$ .

Finding the coordinate vector of  $\mathbf{v}$  is equivalent to solving  $A\mathbf{x} = \mathbf{v}$  for  $\mathbf{x}$ . Performing Gaussian elimination on the augmented matrix  $[A|\mathbf{v}]$ , we have

$$\begin{pmatrix} 1 & 2 & -1 & 1 & -2 \\ 3 & 2 & 0 & -2 & -6 \\ 0 & 1 & -1 & 1 & -4 \\ 5 & 3 & 0 & -1 & -2 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & -1 & 1 & -2 \\ 0 & -4 & 3 & -5 & 0 \\ 0 & 0 & 1 & 1 & 16 \\ 0 & 0 & 0 & 1 & 4 \end{pmatrix}.$$

Using back substitution, we obtain

$$[\mathbf{v}]_{\mathcal{B}} = \begin{pmatrix} -2 \\ 4 \\ 12 \\ 4 \end{pmatrix}.$$



### 6.63 Problem 63

$$\mathbf{v} = A \begin{pmatrix} 1 \\ 6 \\ -1 \\ 4 \end{pmatrix} = \begin{pmatrix} 1 & 2 & -1 & 1 \\ 3 & 2 & 0 & -2 \\ 0 & 1 & -1 & 1 \\ 5 & 3 & 0 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 6 \\ -1 \\ 4 \end{pmatrix} = \begin{pmatrix} 18 \\ 7 \\ 11 \\ 19 \end{pmatrix}.$$

### 6.64 Problem 64

$$\mathbf{v} = \begin{pmatrix} 1 & 3 & 2 \\ 2 & 7 & 4 \\ -2 & -5 & 9 \end{pmatrix} \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 10 \end{pmatrix}.$$

### 6.65 Problem 65

a)

Solving

$$\mathbf{v} = \begin{pmatrix} 1 & 1 & -2 \\ 0 & 1 & 0 \\ 1 & 1 & -1 \end{pmatrix} [\mathbf{v}]_{\mathcal{B}} = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix},$$

for  $[\mathbf{v}]_{\mathcal{B}}$ , we obtain

$$[\mathbf{v}]_{\mathcal{B}} = \begin{pmatrix} 3 \\ 2 \\ 2 \end{pmatrix}.$$

b)

Solving

$$\mathbf{v} = \begin{pmatrix} 1 & 1 & -2 \\ 0 & 1 & 0 \\ 1 & 1 & -1 \end{pmatrix} [\mathbf{v}]_{\mathcal{B}} = \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix},$$

for  $[\mathbf{v}]_{\mathcal{B}}$ , we obtain

$$[\mathbf{v}]_{\mathcal{B}} = \begin{pmatrix} -a_1 - a_2 + 2a_3 \\ a_2 \\ -a_1 + a_3 \end{pmatrix}.$$

### 6.66 Problem 66

a)

$$\mathbf{v} = B[\mathbf{v}]_B = \begin{pmatrix} 1 & 3 & -2 \\ -1 & 4 & 3 \\ 2 & 6 & -3 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \\ -3 \end{pmatrix} = \begin{pmatrix} 12 \\ -8 \\ 21 \end{pmatrix}.$$

b)

Solving

$$\mathbf{v} = \begin{pmatrix} 1 & 3 & -2 \\ -1 & 4 & 3 \\ 2 & 6 & -3 \end{pmatrix} [\mathbf{w}]_B = \begin{pmatrix} 7 \\ -3 \\ 11 \end{pmatrix}$$

for  $[\mathbf{w}]_B$ , we obtain

$$[\mathbf{w}]_B = \begin{pmatrix} -2 \\ 1 \\ -3 \end{pmatrix}.$$

## 6.67 Problem 67

a)

It is easy to see that  $\|\mathbf{v}_1\| = \|\mathbf{v}_2\| = \|\mathbf{v}_3\| = 1$  and  $\mathbf{v}_1 \cdot \mathbf{v}_2 = \mathbf{v}_2 \cdot \mathbf{v}_3 = \mathbf{v}_3 \cdot \mathbf{v}_1 = 0$ . Hence,  $S$  is an orthonormal set of vectors in  $\mathbb{R}^3$ .

b)

Let  $A$  be a 3 by 3 matrix whose columns are the vectors in  $S$ . Then,  $S$  is linearly independent because  $\det(A) = 1/6 \neq 0$ . Also, since  $|S| = 3$ , we have that  $S$  is a basis for  $\mathbb{R}^3$ .

c)

Let

$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + x_3\mathbf{v}_3 = \begin{pmatrix} -1 \\ 3 \\ 4 \end{pmatrix}, \quad x_1, x_2, x_3 \in \mathbb{R}.$$

Since  $S$  is an orthonormal set, we have

$$\begin{aligned} x_1 &= (x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + x_3\mathbf{v}_3) \cdot \mathbf{v}_1 = \begin{pmatrix} -1 \\ 3 \\ 4 \end{pmatrix} \cdot \mathbf{v}_1 = -2\sqrt{2}, \\ x_2 &= (x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + x_3\mathbf{v}_3) \cdot \mathbf{v}_2 = \begin{pmatrix} -1 \\ 3 \\ 4 \end{pmatrix} \cdot \mathbf{v}_2 = 2\sqrt{3}, \\ x_3 &= (x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + x_3\mathbf{v}_3) \cdot \mathbf{v}_3 = \begin{pmatrix} -1 \\ 3 \\ 4 \end{pmatrix} \cdot \mathbf{v}_3 = \sqrt{6}. \end{aligned}$$

Hence,

$$\left[ \begin{pmatrix} -1 \\ 3 \\ 4 \end{pmatrix} \right]_S = \begin{pmatrix} -2\sqrt{2} \\ 2\sqrt{3} \\ \sqrt{6} \end{pmatrix}.$$

## 6.68 Problem 68

Suppose that

$$x_1\mathbf{u}_1 + \cdots + x_n\mathbf{u}_n = \mathbf{0}, \quad x_1, \dots, x_n \in \mathbb{R}. \quad (*)$$

Since  $S$  is an orthonormal set, for any  $x_j \in \{x_1, \dots, x_n\}$ , we have

$$x_j = (x_1 \mathbf{u}_1 + \dots + x_n \mathbf{u}_n) \cdot \mathbf{u}_j = \mathbf{0} \cdot \mathbf{u}_j = 0.$$

Hence, (\*) holds only when  $x_1 = \dots = x_n = 0$ , which implies that  $S$  is linearly independent. Also, since  $|S| = n$ , we have that  $S$  is a basis for  $\mathbb{R}^n$ . Further,

$$[\mathbf{v}]_S = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \quad x_j = \mathbf{u}_j \cdot \mathbf{v}.$$

## 6.69 Problem 69

a)

$$\begin{pmatrix} 1 & 0 \\ 0 & 4 \end{pmatrix}, \quad \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}, \quad \begin{pmatrix} 3 & 0 \\ 0 & 2 \end{pmatrix}.$$

b)

No. Because the zero matrix  $\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$  is not in  $S$ .

## 6.70 Problem 70

a)

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \quad \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}, \quad \begin{pmatrix} 2 & 0 \\ 0 & -2 \end{pmatrix}.$$

b)

Yes. This can be easily proved using the Subspace Theorem.

## 6.71 Problem 71

Suppose that

$$\lambda_1 \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + \lambda_2 \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + \lambda_3 \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + \lambda_4 \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix},$$

where  $\lambda_1, \lambda_2, \lambda_3, \lambda_4 \in \mathbb{R}$ . Equating the coefficients gives

$$\begin{aligned} \lambda_1 &= 0, \\ \lambda_2 &= 0, \\ \lambda_3 &= 0, \\ \lambda_4 &= 0. \end{aligned}$$

This is equivalent to

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \lambda_2 \\ \lambda_3 \\ \lambda_4 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \end{pmatrix},$$

which has unique solution  $\lambda_1 = \lambda_2 = \lambda_3 = \lambda_4 = 0$ . Therefore, the four matrices are linearly independent, and hence form a basis for  $M_{22}(\mathbb{R})$ .

## 6.72 Problem 72

The proof is the same as Problem 71. Because  $\mathbb{R}$  is a subfield of  $\mathbb{C}$ , all properties we used for  $\mathbb{R}$  hold automatically for  $\mathbb{C}$ .

## 6.73 Problem 73

Suppose that

$$\lambda_1 \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + \lambda_2 \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} + \lambda_3 \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} + \lambda_4 \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix},$$

where  $\lambda_1, \lambda_2, \lambda_3, \lambda_4 \in \mathbb{C}$ . Equating the coefficients gives

$$\begin{aligned} \lambda_1 + \lambda_4 &= 0, \\ \lambda_2 - i\lambda_3 &= 0, \\ \lambda_2 + i\lambda_3 &= 0, \\ \lambda_1 - \lambda_4 &= 0. \end{aligned}$$

Performing Gaussian elimination on the coefficient matrix, we have

$$\begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & -i & 0 \\ 0 & 1 & i & 0 \\ 1 & 0 & 0 & -1 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & -i & 0 \\ 0 & 0 & 2i & 0 \\ 0 & 0 & 0 & -2 \end{pmatrix},$$

where all columns are leading columns. Thus, the system of equations has unique solution  $\lambda_1 = \lambda_2 = \lambda_3 = \lambda_4 = 0$ . Therefore, the four matrices are linearly independent, and hence form a basis for  $M_{22}(\mathbb{C})$ .

## 6.74 Problem 74

a)

$$[A]_{\mathcal{B}} = \begin{pmatrix} a_{11} \\ a_{12} \\ a_{21} \\ a_{22} \end{pmatrix}.$$

b)

This problem is equivalent to solving the system of equations

$$\begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & -i & 0 \\ 0 & 1 & i & 0 \\ 1 & 0 & 0 & -1 \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \lambda_2 \\ \lambda_3 \\ \lambda_4 \end{pmatrix} = \begin{pmatrix} a_{11} \\ a_{12} \\ a_{21} \\ a_{22} \end{pmatrix}.$$

Performing Gaussian elimination on the augmented matrix and solving for the unknowns, we obtain

$$[A]_{\mathcal{P}} = \begin{pmatrix} (a_{11} + a_{22})/2 \\ (a_{12} + a_{21})/2 \\ (-a_{12} + a_{21})/(2i) \\ (a_{11} - a_{22}/2) \end{pmatrix}.$$

## 6.75 Problem 75

a)

$$\begin{pmatrix} -4 & 2 \\ -1 & -3 \end{pmatrix} = -4 \begin{pmatrix} 1 & 0 \\ -2 & 0 \end{pmatrix} + 2 \begin{pmatrix} 0 & 1 \\ 3 & 0 \end{pmatrix} - 3 \begin{pmatrix} 0 & 0 \\ 5 & 1 \end{pmatrix}.$$

b)

No. Because  $|R| = 3 < 4 = \dim M_{22}(\mathbb{R})$ .

## 6.77 Problem 77

Here gives the proof of axioms 2 and 4, the other two can be proved in a similar manner.

### Associative Law of Addition

For any functions  $f, g, h \in \mathcal{C}[X]$ , we have

$$\begin{aligned} ((f + g) + h)(x) &= (f + g)(x) + h(x) \\ &= f(x) + g(x) + h(x) \\ &= f(x) + (g(x) + h(x)) \\ &= f(x) + (g + h)(x) \\ &= (f + (g + h))(x). \end{aligned}$$

Thus,  $\mathcal{C}[X]$  is associative.

### Existence of the Zero

The zero function is  $f(x) = 0, x \in X$ , such that for any function  $g \in \mathcal{C}[X]$ , we have

$$(f + g)(x) = (g + f)(x) = g(x), \quad x \in X.$$

## 6.78 Problem 78

This can be easily proved using the Subspace Theorem.

## 6.79 Problem 79

No. Because the zero function  $y(x) = 0$  of  $\mathcal{R}[\mathbb{R}]$  is not in  $S$ .

## 6.80 Problem 80

Firstly, we notice that  $\mathcal{C}^{(k)}[\mathbb{R}]$  is a subset of the vector space  $\mathcal{R}[\mathbb{R}]$ . The zero function  $f(x) = 0, x \in \mathbb{C}$  is in  $\mathcal{C}^{(k)}[\mathbb{R}]$ . In addition, for any functions  $f, g \in \mathcal{C}^{(k)}[\mathbb{R}]$  and any scalars  $\lambda, \mu \in \mathbb{R}$ , we have that for  $(\lambda f + \mu g)$ , the first  $k$  derivatives exist and is continuous. Hence, by the alternative Subspace Theorem,  $\mathcal{C}^{(k)}[\mathbb{R}]$  is a subspace of  $\mathcal{R}[\mathbb{R}]$ .

## 6.82 Problem 82

It is easy to verify that the zero function  $f(x) = 0, x \in [-\pi, \pi]$  is in  $S$ .

For any functions  $f, g \in S$  and any scalars  $\lambda, \mu \in \mathbb{R}$ , since  $\mathcal{R}[-\pi, \pi]$  is a vector space, we have  $(\lambda f + \mu g) \in \mathcal{R}[-\pi, \pi]$ , and

$$\begin{aligned} \int_{-\pi}^{\pi} \cos(x+t)((\lambda f + \mu g)(t))dt &= \int_{-\pi}^{\pi} \cos(x+t)(\lambda f + \mu g)(t)dt \\ &= \int_{-\pi}^{\pi} \cos(x+t)(\lambda f(t) + \mu g(t))dt \\ &= \lambda \int_{-\pi}^{\pi} \cos(x+t)f(t)dt + \mu \int_{-\pi}^{\pi} \cos(x+t)g(t)dt \\ &= \lambda 0 + \mu 0 \\ &= 0, \end{aligned}$$

which implies that  $(\lambda f + \mu g) \in S$ . Therefore, by the alternative Subspace Theorem,  $S$  is a subspace of  $\mathcal{R}[-\pi, \pi]$ .

## 6.83 Problem 83

Suppose that

$$\lambda_1 f_1(x) + \cdots + \lambda_n f_n(x) = 0, \quad (*)$$

where  $\lambda_1, \dots, \lambda_n \in \mathbb{R}$ . Multiplying  $f_i(x)$  on both sides of (\*), where  $1 \leq i \leq n$ , we have

$$(\lambda_1 f_1(x) + \cdots + \lambda_n f_n(x))f_i(x) = (0)f_i(x),$$

that is,

$$\lambda_1 f_1(x)f_i(x) + \cdots + \lambda_i f_i^2(x) + \cdots + \lambda_n f_n(x)f_i(x) = 0.$$

Integrating both sides from  $a$  to  $b$  with respect to  $x$  gives

$$\int_a^b (\lambda_1 f_1(x)f_i(x) + \cdots + \lambda_i f_i^2(x) + \cdots + \lambda_n f_n(x)f_i(x))dx = \int_a^b 0 dx = 0.$$

Noticing that the left-hand side can be further reduced to

$$\lambda_1 \int_a^b f_1(x)f_i(x)dx + \cdots + \lambda_i \int_a^b f_i^2(x)dx + \cdots + \lambda_n \int_a^b f_n(x)f_i(x)dx = \lambda_i,$$

we have  $\lambda_i = 0$  for  $1 \leq i \leq n$ . Therefore, (\*) has unique solution  $\lambda_1 = \cdots = \lambda_n = 0$ , which implies that  $S$  is a linearly independent set.

### 6.84 Problem 84

This can be easily proved using the Subspace Theorem. I may come back and update this at some point.

### 6.85 Problem 85

No, because the zero polynomial is not in  $S$ .

### 6.86 Problem 86

This can be easily proved using the Subspace Theorem. I may come back and update this at some point.

### 6.87 Problem 87

This problem is equivalent to check the solvability of

$$\begin{pmatrix} 1 & -4 & -5 \\ 2 & -1 & -1 \\ 3 & 9 & 12 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -6 \\ 2 \\ 30 \end{pmatrix}.$$

Performing Gaussian elimination on the augmented matrix, we have

$$\begin{pmatrix} 1 & -4 & -5 & -6 \\ 2 & -1 & -1 & 2 \\ 3 & 9 & 12 & 30 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & -4 & -5 & -6 \\ 0 & 7 & 9 & 14 \\ 0 & 0 & 0 & 6 \end{pmatrix},$$

where the right-most column is a leading column. Therefore, the system of equations has no solution, that is,  $p \notin \text{span}(p_1, p_2, p_3)$ .

### 6.88 Problem 88

Let  $p(z) = a_0 + a_1x + a_2x^2$ . This problem is equivalent to finding the conditions for  $a_0, a_1, a_2$  such that the following system of equations has a solution:

$$\begin{pmatrix} 3 & -3 & -6 \\ 2 & -2 & -4 \\ 0 & 5 & 15 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} a_2 \\ a_1 \\ a_0 \end{pmatrix}.$$

Performing Gaussian elimination on the augmented matrix, we have

$$\begin{pmatrix} 3 & -3 & -6 & a_2 \\ 2 & -2 & -4 & a_1 \\ 0 & 5 & 15 & a_0 \end{pmatrix} \longrightarrow \begin{pmatrix} 3 & -3 & -6 & a_2 \\ 0 & 1 & 3 & \frac{1}{5}a_0 \\ 0 & 0 & 0 & a_1 - \frac{2}{3}a_2 \end{pmatrix}.$$

Hence, the condition  $a_1 - \frac{2}{3}a_2 = 0$  must be satisfied so that  $p$  can be a linear combination of  $p_1, p_2$  and  $p_3$ .

### 6.89 Problem 89

No for both problems. Because  $\text{rank}(A) = 2 < 3 = \dim \mathbb{P}_2$ .

### 6.90 Problem 90

$$\begin{pmatrix} 1 & 2 & 5 \\ 1 & -1 & -4 \\ -1 & 0 & 1 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & 5 \\ 0 & 1 & 3 \\ 0 & 0 & 0 \end{pmatrix},$$

where the third column is not leading. Therefore,  $S$  is not a linearly independent set.

To find an expression of  $p_3$  in terms of  $p_1$  and  $p_2$ , we can treat the row echelon form as an augmented matrix and use back substitution, which gives

$$p_3 = -p_1 + 3p_2.$$

### 6.92 Problem 92

This problem is equivalent to solving the system of equations  $A\mathbf{x} = \mathbf{b}$ . The steps have been omitted for simplicity. The answer is  $\begin{pmatrix} 2 \\ -1 \\ 0 \end{pmatrix}$ .

### 6.95 Problem 95

Suppose that

$$x_1 p_1(x) + \cdots + x_n p_n(x) = 0, \quad (*)$$

where  $x_1, \dots, x_n \in \mathbb{R}$ . Multiplying  $p_i(x)$  on both sides of  $(*)$ , where  $1 \leq i \leq n$ , we have

$$(x_1 p_1(x) + \cdots + x_n p_n(x)) p_i(x) = (0) p_i(x),$$

that is,

$$x_1 p_1(x) p_i(x) + \cdots + x_i p_i^2(x) + \cdots + x_n p_n(x) p_i(x) = 0.$$

Integrating both sides from  $a$  to  $b$  with respect to  $x$  gives

$$\int_a^b (x_1 p_1(x) p_i(x) + \cdots + x_i p_i^2(x) + \cdots + x_n p_n(x) p_i(x)) dx = \int_a^b 0 dx = 0.$$

Noticing that the left-hand side can be further reduced to

$$x_1 \int_a^b p_1(x) p_i(x) dx + \cdots + x_i \int_a^b p_i^2(x) dx + \cdots + x_n \int_a^b p_n(x) p_i(x) dx = x_i,$$

we have  $x_i = 0$  for  $1 \leq i \leq n$ . Therefore,  $(*)$  has unique solution  $x_1 = \cdots = x_n = 0$ , which implies that  $S$  is a linearly independent set. Also, since  $|S| = n = \dim \mathbb{P}_{n-1}(\mathbb{R})$ ,  $S$  is a basis for  $\mathbb{P}_{n-1}(\mathbb{R})$ , and hence, by the definition of coordinate vector,

$$[p]_S = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}, \quad \text{where} \quad x = \int_a^b p_i(x) p(x) dx.$$



## Chapter 7

# Linear Transformations

### 7.3 Problem 3

a)

$T : \mathbb{C} \rightarrow \mathbb{R}$  is a linear map because

$$\begin{aligned} T(\lambda z_1 + \mu z_2) &= \operatorname{Re}(\lambda z_1 + \mu z_2) \\ &= \lambda \operatorname{Re}(z_1) + \mu \operatorname{Re}(z_2) \\ &= \lambda T(z_1) + \mu T(z_2) \end{aligned}$$

for any  $z_1, z_2 \in \mathbb{C}$  and any  $\lambda, \mu \in \mathbb{R}$ .

b)

$T : \mathbb{C} \rightarrow \mathbb{R}$  is a linear map because

$$\begin{aligned} T(\lambda z_1 + \mu z_2) &= \operatorname{Im}(\lambda z_1 + \mu z_2) \\ &= \lambda \operatorname{Im}(z_1) + \mu \operatorname{Im}(z_2) \\ &= \lambda T(z_1) + \mu T(z_2) \end{aligned}$$

for any  $z_1, z_2 \in \mathbb{C}$  and any  $\lambda, \mu \in \mathbb{R}$ .

c)

$T : \mathbb{C} \rightarrow [0, \infty)$  is not a linear map because  $T(-1) \neq -T(1)$ .

d)

$T : \mathbb{C} \setminus \{0\} \rightarrow (-\pi, \pi]$  is not a linear map because neither the domain nor the codomain is a vector space.

e)

$T : \mathbb{C} \rightarrow \mathbb{C}$  is a linear map because it preserves linear combination.

## 7.5 Problem 5

If  $n = 1$ , then  $T(\lambda_1 \mathbf{v}_1) = \lambda_1 T(\mathbf{v}_1)$ .

Suppose that for  $n > 1$ ,

$$T(\lambda_1 \mathbf{v}_1 + \cdots + \lambda_n \mathbf{v}_n) = \lambda_1 T(\mathbf{v}_1) + \cdots + \lambda_n T(\mathbf{v}_n)$$

holds. Then,

$$\begin{aligned} T(\lambda_1 \mathbf{v}_1 + \cdots + \lambda_{n+1} \mathbf{v}_{n+1}) &= T((\lambda_1 \mathbf{v}_1 + \cdots + \lambda_n \mathbf{v}_n) + \lambda_{n+1} \mathbf{v}_{n+1}) \\ &= 1 \cdot T(\lambda_1 \mathbf{v}_1 + \cdots + \lambda_n \mathbf{v}_n) + \lambda_{n+1} T(\mathbf{v}_{n+1}) \\ &= \lambda_1 T(\mathbf{v}_1) + \cdots + \lambda_{n+1} T(\mathbf{v}_{n+1}). \end{aligned}$$

## 7.10 Problem 10

If  $T$  is a linear map, then

$$\begin{aligned} T \begin{pmatrix} 1 \\ 7 \\ 13 \end{pmatrix} &= T \left( 3 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} + \begin{pmatrix} -2 \\ 1 \\ 4 \end{pmatrix} \right) = 3T \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} + T \begin{pmatrix} -2 \\ 1 \\ 4 \end{pmatrix} \\ &= 3 \begin{pmatrix} 4 \\ 1 \end{pmatrix} + \begin{pmatrix} -1 \\ 2 \end{pmatrix} = \begin{pmatrix} 11 \\ 5 \end{pmatrix} \neq \begin{pmatrix} 4 \\ -2 \end{pmatrix}. \end{aligned}$$

Therefore,  $T$  is not a linear map.

## 7.16 Problem 16

$$T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -x_1 \\ x_2 \end{pmatrix}.$$

Since

$$T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = x_1 \begin{pmatrix} -1 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix},$$

we have

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

## 7.17 Problem 17

$$T \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ -x_3 \end{pmatrix}.$$

Since

$$T \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = x_1 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + x_2 \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + x_3 \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix},$$

we have

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

## 7.18 Problem 18

Here gives an intuitive but slightly cumbersome solution.

We notice that  $T(\mathbf{p}) = \mathbf{q}$  satisfies

$$\mathbf{p} + \mathbf{q} = \lambda \mathbf{d} \quad \text{for some } \lambda \in \mathbb{R} \quad (*)$$

and

$$\text{proj}_{\mathbf{x}} \mathbf{p} = \text{proj}_{\mathbf{x}} \mathbf{q}. \quad (**)$$

(\*) implies that

$$q_i = \lambda d_i - p_i \quad \text{for } 1 \leq i \leq n.$$

(\*\*) implies that

$$\left( \frac{\mathbf{p} \cdot \mathbf{d}}{\|\mathbf{d}\|^2} \right) \mathbf{d} = \left( \frac{\mathbf{q} \cdot \mathbf{d}}{\|\mathbf{d}\|^2} \right) \mathbf{d},$$

that is,

$$\left( \sum_{j=1}^n p_j d_j \right) \mathbf{d} = \left( \sum_{j=1}^n q_j d_j \right) \mathbf{d} = \left( \sum_{j=1}^n (\lambda d_j - p_j) d_j \right) \mathbf{d},$$

where we have cancelled  $1/\|\mathbf{d}\|^2$  on both sides. Rearranging the equation, we obtain

$$\lambda = \frac{2 \sum_{j=1}^n p_j d_j}{\sum_{k=1}^n d_k^2} = \frac{2 \sum_{j=1}^n p_j d_j}{\|\mathbf{d}\|^2}.$$

Plugging this back into (\*), we have

$$q_i = \frac{2 \sum_{j=1}^n p_j d_j d_i}{\|\mathbf{d}\|^2} - p_i = p_i \left( \frac{2d_i^2}{\|\mathbf{d}\|^2} - 1 \right) + \sum_{j=1, j \neq i}^n p_j \frac{2d_i d_j}{\|\mathbf{d}\|^2}.$$

Hence, the matrix  $A$  is in the form of

$$[A]_{ij} = \begin{cases} \frac{2d_i^2}{\|\mathbf{d}\|^2} - 1 & \text{if } i = j \\ \frac{2d_i d_j}{\|\mathbf{d}\|^2} & \text{if } i \neq j \end{cases}.$$

## 7.20 Problem 20

This can be generalized to  $\mathbb{R}^n$ .

$T$  is a linear map because for any vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^n$  and any scalars  $\lambda, \mu \in \mathbb{R}$ , we have

$$\begin{aligned} T(\lambda \mathbf{u} + \mu \mathbf{v}) &= \frac{(\lambda \mathbf{u} + \mu \mathbf{v}) \cdot \mathbf{b}}{|\mathbf{b}|^2} \mathbf{b} \\ &= \lambda \frac{\mathbf{u} \cdot \mathbf{b}}{|\mathbf{b}|^2} + \mu \frac{\mathbf{v} \cdot \mathbf{b}}{|\mathbf{b}|^2} \\ &= \lambda T(\mathbf{u}) + \mu T(\mathbf{v}). \end{aligned}$$

Hence,  $T$  is a linear map. Further,

$$T(\mathbf{a}) = \frac{\mathbf{a} \cdot \mathbf{b}}{|\mathbf{b}|^2} \mathbf{b} = \frac{\mathbf{b} \mathbf{b}^T}{\mathbf{b}^T \mathbf{b}} \mathbf{a}$$

implies that  $A = \frac{\mathbf{b} \mathbf{b}^T}{\mathbf{b}^T \mathbf{b}}$ .

### 7.21 Problem 21

$S$  is not a linear map because for any  $\mathbf{b} \neq \mathbf{0}$ ,  $S(-\mathbf{b}) = S(\mathbf{b}) \neq -S(\mathbf{b})$ .

### 7.22 Problem 22

$$\begin{aligned} A_\theta A_\phi &= \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{pmatrix} \\ &= \begin{pmatrix} \cos \theta \cos \phi - \sin \theta \sin \phi & -(\sin \theta \cos \phi + \cos \theta \sin \phi) \\ \sin \theta \cos \phi + \cos \theta \sin \phi & \cos \theta \cos \phi - \sin \theta \sin \phi \end{pmatrix} \\ &= \begin{pmatrix} \cos(\phi + \theta) & -\sin(\phi + \theta) \\ \sin(\phi + \theta) & \cos(\phi + \theta) \end{pmatrix} \\ &= A_{\phi + \theta}, \end{aligned}$$

which says that rotating in the plane by angles  $\phi$  and  $\theta$  consecutively is equivalent to rotating by angle  $\phi + \theta$ .

### 7.23 Problem 23

$$R_\alpha(\mathbf{i}) = \begin{pmatrix} \cos \alpha \\ 0 \\ -\sin \alpha \end{pmatrix}, \quad R_\alpha(\mathbf{j}) = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \quad R_\alpha(\mathbf{k}) = \begin{pmatrix} \sin \alpha \\ 0 \\ \cos \alpha \end{pmatrix}.$$

Hence,

$$A = \begin{pmatrix} \cos \alpha & 0 & \sin \alpha \\ 0 & 1 & 0 \\ -\sin \alpha & 0 & \cos \alpha \end{pmatrix},$$

which is obviously a linear map.

### 7.27 Problem 27

By the definition of kernel, matrix  $A$  can be

$$\begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \end{pmatrix}.$$

### 7.30 Problem 30

a)

For any vectors  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^3$  and any scalars  $\lambda, \mu \in \mathbb{R}$ , we have

$$T(\lambda \mathbf{u} + \mu \mathbf{v}) = \mathbf{b} \times (\lambda \mathbf{u} + \mu \mathbf{v}) = \lambda(\mathbf{b} \times \mathbf{u}) + \mu(\mathbf{b} \times \mathbf{v}) = \lambda T(\mathbf{u}) + \mu T(\mathbf{v}).$$

Hence,  $T$  is a linear map.

b)

$$T(\mathbf{x}) = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \times \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -3x_2 + 2x_3 \\ 3x_1 - x_3 \\ -2x_1 + x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

gives the system of equations

$$\begin{aligned} -3x_2 + 2x_3 &= 0 \\ 3x_1 - x_3 &= 0 \\ -2x_1 + x_2 &= 0. \end{aligned}$$

Performing Gaussian elimination on the coefficient matrix, we obtain

$$\begin{pmatrix} 0 & -3 & 2 \\ 3 & 0 & -1 \\ -2 & 1 & 0 \end{pmatrix} \longrightarrow \begin{pmatrix} 3 & 0 & 1 \\ 0 & 1 & -\frac{2}{3} \\ 0 & 0 & 0 \end{pmatrix},$$

which implies that the nullity of  $T$  is 1.

### 7.39 Problem 39

If a) holds, then  $A$  is invertible, that is,  $\det(A) \neq 0$ , which implies that the nullity is 0, and hence b) and c) hold. Then, by the Rank-nullity Theorem for matrix, we have  $\text{rank}(A) = n - \text{nullity}(A) = n - 0 = n$ . Thus, d) and e) hold, and the  $n$  columns of  $A$  are linearly independent, which implies that f) holds. Finally, if f) holds, we have that the columns of  $A$  are linearly independent, meaning that  $A$  is invertible, and hence a) holds.

### 7.40 Problem 40

The rank of  $A$  is  $r$  means that after Gaussian elimination, only the top  $r$  rows are not all zeroes.

### 7.41 Problem 41

Because by the Rank-nullity Theorem for matrix, the rank of  $A$  is  $n - \nu$ . Hence, there are  $m - (n - \nu) = m - n + \nu$  zero rows after Gaussian elimination.

### 7.42 Problem 42

In this case,  $A$  is a full rank matrix that has no zero rows after Gaussian elimination.

### 7.43 Problem 43

All the statements are the properties of a linear isomorphism.

### 7.44 Problem 44

For any vectors  $\mathbf{a}, \mathbf{b} \in \mathbb{R}^4$  and any scalars  $\lambda, \mu \in \mathbb{R}$ , we have

$$T(\lambda\mathbf{a} + \mu\mathbf{b}) = \begin{pmatrix} \lambda a_1 + \mu b_1 & \lambda a_2 + \mu b_2 \\ \lambda a_3 + \mu b_3 & \lambda a_4 + \mu b_4 \end{pmatrix} = \lambda \begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix} + \mu \begin{pmatrix} b_1 & b_2 \\ b_3 & b_4 \end{pmatrix} = \lambda T(\mathbf{a}) + \mu T(\mathbf{b}).$$

Hence,  $T$  is a linear map.

### 7.45 Problem 45

For any vectors  $\mathbf{a}, \mathbf{b} \in \mathbb{R}^4$  and any scalars  $\lambda, \mu \in \mathbb{R}$ , we have

$$\begin{aligned} T(\lambda\mathbf{a} + \mu\mathbf{b}) &= \begin{pmatrix} 3(\lambda a_1 + \mu b_1) - 2(\lambda a_4 + \mu b_4) & (\lambda a_4 + \mu b_4) + 2(\lambda a_3 + \mu b_3) \\ -5(\lambda a_2 + \mu b_2) + 3(\lambda a_3 + \mu b_3) & \lambda a_1 + \mu b_1 \end{pmatrix} \\ &= \lambda \begin{pmatrix} 3a_1 - 2a_4 & a_4 + 2a_3 \\ -5a_2 + 3a_3 & a_1 \end{pmatrix} + \mu \begin{pmatrix} 3b_1 - 2b_4 & b_4 + 2b_3 \\ -5b_2 + 3b_3 & b_1 \end{pmatrix} \\ &= \lambda T(\mathbf{a}) + \mu T(\mathbf{b}). \end{aligned}$$

Hence,  $T$  is a linear map.

### 7.46 Problem 46

Yes.

### 7.48 Problem 48

It is easy to prove that  $T$  is a linear map, and we can also find the matrix  $A$  by

$$T \begin{pmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{pmatrix} = \begin{pmatrix} a_1 - 3a_2 \\ 2a_3 - 3a_4 \\ a_2 \\ 3a_1 - a_2 + 2a_3 + 4a_4 \end{pmatrix} = \underbrace{\begin{pmatrix} 1 & -3 & 0 & 0 \\ 0 & 0 & 2 & -3 \\ 0 & 1 & 0 & 0 \\ 3 & -1 & 2 & 4 \end{pmatrix}}_A \begin{pmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{pmatrix}.$$

### 7.51 Problem 51

For any  $p_1, p_2 \in \mathbb{P}_3$  and any  $\lambda_1, \lambda_2 \in \mathbb{R}$ , we have

$$\begin{aligned} (T(\lambda_1 p_1 + \lambda_2 p_2))(x) &= \int_0^x (\lambda_1 p_1 + \lambda_2 p_2)(t) dt \\ &= \int_0^x (\lambda_1 p_1)(t) dt + \int_0^x (\lambda_2 p_2)(t) dt \\ &= \lambda_1 \int_0^x p_1(t) dt + \lambda_2 \int_0^x p_2(t) dt \\ &= \lambda_1 (T(p_1))(x) + \lambda_2 (T(p_2))(x), \end{aligned}$$

which implies that  $T(\lambda_1 p_1 + \lambda_2 p_2) = \lambda_1 T(p_1) + \lambda_2 T(p_2)$ . Hence,  $T$  is a linear map.

## 7.52 Problem 52

The zero function of  $\mathcal{R}[\mathbb{R}]$  is in  $V$ .

For any functions  $f_1, f_2 \in V$  and any scalars  $\lambda_1, \lambda_2 \in \mathbb{R}$ , since  $\mathcal{R}[\mathbb{R}]$  is a vector space, it is closed under linear combination, so we have  $(\lambda_1 f_1 + \lambda_2 f_2) \in \mathcal{R}[\mathbb{R}]$ , and hence,  $\int_0^x (\lambda_1 f_1 + \lambda_2 f_2)(t) dt$  also exists for all  $x \in \mathbb{R}$ , which implies that  $(\lambda_1 f_1 + \lambda_2 f_2) \in V$ .

Therefore, by the alternative Subspace Theorem,  $V$  is a subspace.

For the second part of the problem, for any functions  $f_1, f_2 \in V$  and any scalars  $\lambda_1, \lambda_2 \in \mathbb{R}$ , we have

$$\begin{aligned}(T(\lambda_1 f_1 + \lambda_2 f_2))(x) &= \int_0^x (\lambda_1 f_1 + \lambda_2 f_2)(t) dt \\&= \int_0^x (\lambda_1 f_1)(t) dt + \int_0^x (\lambda_2 f_2)(t) dt \\&= \lambda_1 \int_0^x f_1(t) dt + \lambda_2 \int_0^x f_2(t) dt \\&= \lambda_1 (T(f_1))(x) + \lambda_2 (T(f_2))(x),\end{aligned}$$

which implies that  $T(\lambda_1 f_1 + \lambda_2 f_2) = \lambda_1 T(f_1) + \lambda_2 T(f_2)$ . Hence,  $T$  is a linear map.

## 7.53 Problem 53

$S$  is not a linear map because  $\mathbb{Z}$  is not a vector space. Also,  $S$  is not a linear map either because  $S(1.2 + 1.4) = S(2.6) = 3 \neq 2 = S(1.2) + S(1.4)$ .

## 7.54 Problem 54

Since the Laplace transform is linear, applying it to both sides of the equation, we have

$$\begin{aligned}\text{LHS} &= L(y'' + 4y' + 3y)(s) \\&= L(y'')(s) + 4L(y')(s) + 3L(y)(s) \\&= \int_0^\infty e^{-st} y''(t) dt + 4 \int_0^\infty e^{-st} y'(t) dt + 3y_L(s) \\&= [e^{-st} y'(t)]_0^\infty + (s + 4) \int_0^\infty e^{-st} y'(t) dt + 3y_L(s) \\&= -y'(0) + (s + 4) ([e^{-st} y(t)]_0^\infty + y_L(s)) + 3y_L(s) \\&= -y'(0) + (s + 4)(-y(0)) + (s + 3)(s + 1)y_L(s) \\&= -(s + 6) + (s + 3)(s + 1)y_L(s),\end{aligned}$$

which equals to

$$\text{RHS} = L(e^{-3s}) = \frac{1}{s + 3}.$$

Solving for  $y_L(s)$ , we obtain

$$y_L(s) = \frac{s^2 + 9s + 19}{(s + 3)^2(s + 1)}.$$

## 7.55 Problem 55

a)

For any  $p_1(x), p_2(x) \in \mathbb{P}_3(\mathbb{R})$  and any scalars  $\lambda_1, \lambda_2 \in \mathbb{R}$ , we have

$$\begin{aligned} T(\lambda_1 p_1(x) + \lambda_2 p_2(x)) &= T((\lambda_1 a_1 + \lambda_2 a_2) + (\lambda_1 b_1 + \lambda_2 b_2)x + (\lambda_1 c_1 + \lambda_2 c_2)x^2 + (\lambda_1 d_1 + \lambda_2 d_2)x^3) \\ &= \begin{pmatrix} (\lambda_1 a_1 + \lambda_2 a_2) - (\lambda_1 b_1 + \lambda_2 b_2) \\ (\lambda_1 c_1 + \lambda_2 c_2) - (\lambda_1 d_1 + \lambda_2 d_2) \end{pmatrix} \\ &= \lambda_1 \begin{pmatrix} a_1 - b_1 \\ c_1 - d_1 \end{pmatrix} + \lambda_2 \begin{pmatrix} a_2 - b_2 \\ c_2 - d_2 \end{pmatrix} \\ &= \lambda_1 T(p_1(x)) + \lambda_2 T(p_2(x)). \end{aligned}$$

Hence,  $T$  is linear.

b)

$$T(p(x)) = T\left(\begin{pmatrix} -2 \\ 3 \end{pmatrix} - \begin{pmatrix} -2 \\ 3 \end{pmatrix}\right) = \begin{pmatrix} 0 \\ 0 \end{pmatrix},$$

which implies that  $p(x)$  is in the kernel of  $T$ .

## 7.56 Problem 56

a)

$$T\left(\begin{pmatrix} 1 \\ -3 \\ 2 \\ -4 \end{pmatrix}\right) = (1 - 2(-3)) + (2 - 4)x = 7 - 2x.$$

b)

For any  $\begin{pmatrix} a_1 \\ b_1 \\ c_1 \\ d_1 \end{pmatrix}, \begin{pmatrix} a_2 \\ b_2 \\ c_2 \\ d_2 \end{pmatrix} \in \mathbb{R}^4$  and any scalars  $\lambda_1, \lambda_2 \in \mathbb{R}$ , we have

$$\begin{aligned} T\left(\lambda_1 \begin{pmatrix} a_1 \\ b_1 \\ c_1 \\ d_1 \end{pmatrix} + \lambda_2 \begin{pmatrix} a_2 \\ b_2 \\ c_2 \\ d_2 \end{pmatrix}\right) &= T\begin{pmatrix} \lambda_1 a_1 + \lambda_2 a_2 \\ \lambda_1 b_1 + \lambda_2 b_2 \\ \lambda_1 c_1 + \lambda_2 c_2 \\ \lambda_1 d_1 + \lambda_2 d_2 \end{pmatrix} \\ &= ((\lambda_1 a_1 + \lambda_2 a_2) - 2(\lambda_1 b_1 + \lambda_2 b_2)) + ((\lambda_1 c_1 + \lambda_2 c_2) + (\lambda_1 d_1 + \lambda_2 d_2))x \\ &= \lambda_1((a_1 - 2b_1) + (c_1 + d_1)x) + \lambda_2((a_2 - 2b_2) + (c_2 + d_2)x) \\ &= \lambda_1 T\begin{pmatrix} a_1 \\ b_1 \\ c_1 \\ d_1 \end{pmatrix} + \lambda_2 T\begin{pmatrix} a_2 \\ b_2 \\ c_2 \\ d_2 \end{pmatrix}. \end{aligned}$$

Hence,  $T$  is a linear transformation.



c)

Let  $T \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = 0$ , we can easily come up with a non-zero vector lying in  $\ker(T)$ , for example,  $\begin{pmatrix} 2 \\ 1 \\ 1 \\ -1 \end{pmatrix}$ .

## 7.57 Problem 57

Here gives the solution to  $T \begin{pmatrix} i \\ 2 \\ -1 \end{pmatrix}$ . The other one can be obtained with the same manner.

$$T \begin{pmatrix} i \\ 2 \\ -1 \end{pmatrix} = iT \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + 2T \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} - T \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = (2 + i) + (7 - 4i)z + 2z^2 - 3iz^3.$$

## 7.58 Problem 58

a)

For any  $p(x), q(x) \in \mathbb{P}_3$  and any scalars  $\lambda, \mu \in \mathbb{R}$ , we have

$$\begin{aligned} (T(\lambda p + \mu q))(x) &= \int_0^x (\lambda p + \mu q)(t) dt \\ &= \int_0^x (\lambda p(t) + \mu q(t)) dt \\ &= \lambda \int_0^x p(t) dt + \mu \int_0^x q(t) dt \\ &= \lambda(T(p))(x) + \mu(T(q))(x). \end{aligned}$$

Hence,  $T(\lambda p + \mu q) = \lambda T(p) + \mu T(q)$ , which implies that  $T$  is a linear map.

b)

Since

$$\begin{aligned} T(\mathbf{e}_1) &= \int_0^x 1 dt = x, \\ T(\mathbf{e}_2) &= \int_0^x t dt = \frac{1}{2}x, \\ T(\mathbf{e}_3) &= \int_0^x t^2 dt = \frac{1}{3}x^3, \\ T(\mathbf{e}_4) &= \int_0^x t^3 dt = \frac{1}{4}x^4, \end{aligned}$$

we have

$$A = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{3} & 0 \\ 0 & 0 & 0 & \frac{1}{4} \end{pmatrix}.$$

c)

Noticing that the columns of  $A$  are linearly independent, they can be a basis for  $\text{im}(T)$ .

d)

Since all the columns of  $A$  are linearly independent, we have  $\ker(T) = \{\mathbf{0}\}$ , which implies that the basis for the kernel of  $T$  is the empty set.

### 7.61 Problem 61

Let  $B_V = \left\{ \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\}$  and  $B_W = \left\{ \begin{pmatrix} 1 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right\}$ , then

$$T \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad T \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \end{pmatrix},$$

and hence

$$\left[ T \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right]_{B_W} = \begin{pmatrix} -1 \\ 2 \end{pmatrix}, \quad \left[ T \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right]_{B_W} = \begin{pmatrix} -3 \\ 4 \end{pmatrix}.$$

Therefore, by the Matrix Representation Theorem,  $A = \begin{pmatrix} -1 & -3 \\ 2 & 4 \end{pmatrix}$ .

### 7.62 Problem 62

Noticing that the matrix representing  $G$  with respect to the standard bases in both domain and codomain is just  $A$  itself, we have

$$G(p) = A \begin{pmatrix} -3 \\ 1 \\ 5 \end{pmatrix} = \begin{pmatrix} -6 \\ 4 \\ 0 \end{pmatrix},$$

, which implies that  $G(p) = -6 + 4x$ .

### 7.65 Problem 65

Let  $p(x) = \sum_{k=0}^n a_k x^k$ . Since  $T$  is a linear map,

$$q(x) = \sum_{k=0}^n a_k T(x^k) = \sum_{k=0}^n ((k(k-1)a_k) - (3ka_k) + 3a_k)x^k = \sum_{k=0}^n (k-1)(k-3)a_k x^k, \quad 0 \leq k \leq n.$$

To fit it into the matrix representation, let  $i = k + 1$ , then we end up with a  $(n+1)$  by  $(n+1)$  diagonal matrix with

$$[A]_{ii} = (i-2)(i-4).$$

Hence,  $\ker(T) = \lambda x + \mu x^3$  where  $\lambda, \mu \in \mathbb{R}$  and the nullity of  $T$  is 2.

## 7.66 Problem 66

The proof of the orthonormality of  $B'$  is omitted for simplicity.

Let  $Q$  be the matrix whose columns are  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$  and  $Q'$  be the matrix whose columns are  $\mathbf{u}'_1, \mathbf{u}'_2, \mathbf{u}'_3$ . Since

$$\mathbf{a} = Q \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix} = Q' \begin{pmatrix} a'_1 \\ a'_2 \\ a'_3 \end{pmatrix}$$

and  $Q'$  is orthonormal, there exists a matrix  $E = Q'^{-1}Q$  such that

$$E \begin{pmatrix} a_1 \\ a_2 \\ a_3 \end{pmatrix} = \begin{pmatrix} a'_1 \\ a'_2 \\ a'_3 \end{pmatrix}.$$

Hence,  $T$  is a linear map, and

$$E = Q'^{-1}Q = \left( Q \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & -1 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \end{pmatrix} \right)^{-1} Q = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & -1 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \end{pmatrix}^{-1} = \begin{pmatrix} \frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} \\ 0 & -1 & 0 \end{pmatrix}.$$

Finally, solving the equations for  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$ , we obtain

$$\begin{aligned} \mathbf{u}_1 &= \frac{1}{\sqrt{2}}(\mathbf{u}'_1 - \mathbf{u}'_2), \\ \mathbf{u}_2 &= -\mathbf{u}'_3, \\ \mathbf{u}_3 &= \frac{1}{\sqrt{2}}(\mathbf{u}'_1 + \mathbf{u}'_2). \end{aligned}$$

Hence, for any vector  $\mathbf{v} = \begin{pmatrix} v_1 \\ v_2 \\ v_3 \end{pmatrix} \in V$ , we have

$$\begin{aligned} \mathbf{v} &= v_1\mathbf{u}_1 + v_2\mathbf{u}_2 + v_3\mathbf{u}_3 \\ &= v_1 \frac{1}{\sqrt{2}}(\mathbf{u}'_1 - \mathbf{u}'_2) + v_2(-\mathbf{u}'_3) + v_3 \frac{1}{\sqrt{2}}(\mathbf{u}'_1 + \mathbf{u}'_2) \\ &= \frac{1}{\sqrt{2}}(v_1 + v_3)\mathbf{u}'_1 + \frac{1}{\sqrt{2}}(-v_1 + v_3)\mathbf{u}'_2 + (-v_2)\mathbf{u}'_3, \end{aligned}$$

which implies that

$$[\mathbf{v}]_{B'} = \begin{pmatrix} \frac{1}{\sqrt{2}}(v_1 + v_3) \\ \frac{1}{\sqrt{2}}(-v_1 + v_3) \\ -v_2 \end{pmatrix}.$$

Since this is equal to  $E\mathbf{v}$ , the matrix is a matrix representation of the identity map.

## 7.67 Problem 67

$$\begin{aligned} [(T + S)(\mathbf{v})]_{B_W} &= [T(\mathbf{v}) + S(\mathbf{v})]_{B_W} \\ &= [T(\mathbf{v})]_{B_W} + [S(\mathbf{v})]_{B_W} \\ &= A[\mathbf{v}]_{B_V} + B[\mathbf{v}]_{B_V} \\ &= (A + B)[\mathbf{v}]_{B_V}. \end{aligned}$$

Hence,  $(A + B)$  is the matrix representing  $T + S$  with respect to the bases  $B_V$  and  $B_W$ .

## 7.68 Problem 68

$$\begin{aligned}
 [(S \circ T)(\mathbf{u})]_{B_W} &= [S(T(\mathbf{u}))]_{B_W} \\
 &= [S(A[\mathbf{u}]_{B_U})]_{B_W} \\
 &= B(A[\mathbf{u}]_{B_U}) \\
 &= (BA)[\mathbf{u}]_{B_U}.
 \end{aligned}$$

Hence,  $BA$  is the matrix which represents the composition function  $S \circ T$  with respect to the bases  $B_U$  and  $B_W$ .

## 7.69 Problem 69

a)

Suppose that

$$\begin{aligned}
 \lambda_1 e^x + \lambda_2(x-1)e^x + \lambda_3(x-1)(x-2)e^x &= e^x((\lambda_1 - \lambda_2 + 2\lambda_3) + (\lambda_2 - 3\lambda_3)x + \lambda_3 x^2) \\
 &= 0,
 \end{aligned}$$

then we obtain the system of equations

$$\begin{pmatrix} 1 & -1 & 2 \\ 0 & 1 & -3 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} \lambda_1 \\ \lambda_2 \\ \lambda_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

Since all the columns of the coefficient matrix are leading, we have that the system has only zero solution. Hence,  $B$  is linearly independent.

b)

i)

By the Matrix Representation Theorem, the columns of the matrix are the vectors for

$$\begin{aligned}
 D(e^x) &= e^x, \\
 D((x-1)e^x) &= xe^x, \\
 D((x-1)(x-2)e^x) &= (x^2 - x - 1)e^x.
 \end{aligned}$$

Representing these vectors with respect to  $B$ , for  $e^x$ , we have the system of equations

$$\begin{pmatrix} 1 & -1 & 2 \\ 0 & 1 & -3 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}.$$

Solving for  $x_1, x_2, x_3$ , we obtain the vector representation for  $e^x$  with respect to  $B$ , which is  $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$ .

Similarly, the vector representations for  $xe^x$  and  $(x^2 - x - 1)e^x$  with respect to  $B$  are  $\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$  and

$\begin{pmatrix} -1 \\ 2 \\ 1 \end{pmatrix}$  respectively.

Therefore, the matrix for  $D$  with respect to the ordered basis  $B$  of  $W$  is

$$A = \begin{pmatrix} 1 & 1 & -1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{pmatrix}.$$

ii)

Since  $D$  is a linear map, the composition  $D \circ D$  is also a linear map whose matrix representation is

$$AA = \begin{pmatrix} 1 & 2 & 0 \\ 0 & 1 & 4 \\ 0 & 0 & 1 \end{pmatrix}.$$

iii)

Because all the columns of the product of  $AA$  are leading, the system of equations has unique solution. Hence, the statement holds.

## 7.70 Problem 70

a)

This problem is omitted as similar proofs have been given in the previous problems.

b)

This problem is equivalent to solving the system of equations

$$\begin{pmatrix} 3 & 1 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} a_{11} \\ a_{12} \\ a_{21} \\ a_{22} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix},$$

and we obtain

$$\ker(T) = \lambda \begin{pmatrix} -\frac{1}{3} & 1 \\ 1 & 0 \end{pmatrix}, \quad \lambda \in \mathbb{F}.$$

c)

$\text{rank}(T) = 3$ .

d)

No. Because  $T$  is linear, it is injective if and only if  $\ker(T) = \{\mathbf{0}\}$ , whereas the nullity of  $T$  is 1.

e)

By the Matrix Representation Theorem, the matrix of  $T$  with respect to the standard bases of  $M_{22}(\mathbb{F})$  and  $\mathbb{F}^3$  is

$$\begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & -1 & 0 \\ 3 & 1 & 0 & 0 \end{pmatrix}.$$