

# MA1522 Put It All Together!

## Vector Space

### 2014-2015-Semester 2 Find subspace

#### Question 2 [25 marks]

(a) [11 marks]

Let  $V = \{(a, a, a, 0) \mid a \in \mathbb{R}\}$ .

- (i) Find a basis for  $V$  and determine  $\dim(V)$ .
- (ii) Find a subspace  $W$  of  $\mathbb{R}^4$  such that  $\dim(W) = 3$  and  $\dim(W \cap V) = 1$ . Justify your answer.
- (iii) Let  $U = \{\mathbf{u} \in \mathbb{R}^4 \mid \mathbf{u} \cdot \mathbf{v} = 0 \text{ for all } \mathbf{v} \in V\}$ . Find a basis for and determine the dimension of  $U$ .

### 2015-2016-Semester 2

Let  $V = \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4\}$  be a vector space such that  $\mathbf{v}_i$  are unit vectors for all  $i$  and  $\mathbf{v}_i \cdot \mathbf{v}_j < 0$  if  $i \neq j$ .

- (i) Show that no two vectors among  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4\}$  are linearly dependent.
- (ii) Prove that  $\dim V \geq 3$ .

Clearly  $\mathbf{v}_i \neq 0$ .

(i) Assume that  $\mathbf{v}_1, \mathbf{v}_2$  are linearly dependent.

Let  $\mathbf{v}_1 = c\mathbf{v}_2$ . Then  $0 > \mathbf{v}_1 \cdot \mathbf{v}_2 = c(\mathbf{v}_1 \cdot \mathbf{v}_1) = c$ . We would have

$$0 > \mathbf{v}_1 \cdot \mathbf{v}_3 = c(\mathbf{v}_2 \cdot \mathbf{v}_3) > 0,$$

which is a contradiction.

Hence,  $\mathbf{v}_1$  and  $\mathbf{v}_2$  are linearly independent. Similarly for any other two vectors.

(ii) Assume that  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  are linearly dependent. Then any vector in  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  is a linear combination of the other two. Write  $\mathbf{v}_3 = c_1\mathbf{v}_1 + c_2\mathbf{v}_2$ .

$$0 > \mathbf{v}_1 \cdot \mathbf{v}_3 = c_1 + c_2(\mathbf{v}_1 \cdot \mathbf{v}_2) \quad \text{and} \quad 0 > \mathbf{v}_2 \cdot \mathbf{v}_3 = c_1(\mathbf{v}_1 \cdot \mathbf{v}_2) + c_2.$$

If  $c_1 > 0$ , then  $c_2(\mathbf{v}_1 \cdot \mathbf{v}_2) < -c_1 < 0$ , and thus  $c_2 > 0$ .

By Cauchy-Schwarz inequality,  $0 < -\mathbf{v}_1 \cdot \mathbf{v}_2 \leq \|\mathbf{v}_1\| \|\mathbf{v}_2\| = 1$ . We would have

$$c_1 < c_2(-\mathbf{v}_1 \cdot \mathbf{v}_2) \leq c_2 \quad \text{and} \quad c_2 < c_1(-\mathbf{v}_1 \cdot \mathbf{v}_2) \leq c_1$$

which is a contradiction (based on our assumption that  $c_1 > 0$ ).

Therefore,  $c_1 < 0$  and hence  $c_2 < 0$ .

However, this would imply that  $\mathbf{v}_3 \cdot \mathbf{v}_4 = c_1(\mathbf{v}_1 \cdot \mathbf{v}_4) + c_2(\mathbf{v}_2 \cdot \mathbf{v}_4) > 0$ , a contradiction.

Hence,  $\mathbf{v}_1, \mathbf{v}_2$  and  $\mathbf{v}_3$  are linearly independent. So  $\dim V \geq 3$ .

## Rank/Column Space/Row Space

### 2014-2015-Semester 2

- (iii) Without performing Gaussian elimination, can you tell whether the system

$$\begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & -1 \\ 0 & 1 & 1 \\ 1 & -1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 5 \\ -2 \\ 2 \\ 3 \end{pmatrix}$$

has no solution, exactly one solution, or infinitely many solutions? Why?

2015-2016-Semester 2

Let  $\mathbf{A}$  be a  $3 \times 2$  matrix and  $\mathbf{B}$  be a  $2 \times 3$  matrix such that

$$\mathbf{AB} = \begin{pmatrix} -2 & -14 & 14 \\ 5 & 15 & -10 \\ 4 & 8 & -3 \end{pmatrix}.$$

- (i) [3 marks] Find a basis for the row space of  $\mathbf{AB}$  and state the rank of  $\mathbf{AB}$ .
- (ii) [2 marks] Show that  $(\mathbf{AB})^2 = 5\mathbf{AB}$ .
- (iii) [2 marks] What is the rank of  $\mathbf{BA}$ ? Justify your answer.
- (iv) [3 marks] Find  $\mathbf{BA}$ . Show clearly how you derive your answer.

2016-2017-Semester 2

Let  $\mathbf{A}$  and  $\mathbf{B}$  be square matrices of the same order.

- (i) Prove that the nullspace of  $\mathbf{B}$  is a subspace of the nullspace of  $\mathbf{AB}$ .
- (ii) Using (i) prove that

$$\text{nullity}(\mathbf{A}) + \text{nullity}(\mathbf{B}) \geq \text{nullity}(\mathbf{AB}).$$

If  $\mathbf{B}\mathbf{v} = \mathbf{0}$ , then  $\mathbf{AB}\mathbf{v} = \mathbf{0}$ . So

$$\text{nullspace of } \mathbf{B} \subseteq \text{nullspace of } \mathbf{AB}.$$

Let  $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  be a basis for the nullspace of  $\mathbf{B}$ . Extend it to a basis for the nullspace of  $\mathbf{AB}$ :

$$\{\mathbf{v}_1, \dots, \mathbf{v}_k, \mathbf{v}_{k+1}, \dots, \mathbf{v}_m\}.$$

Then  $\mathbf{B}\mathbf{v}_{k+1}, \dots, \mathbf{B}\mathbf{v}_m \neq \mathbf{0}$ , and they belong to the nullspace of  $\mathbf{A}$ . Suppose

$$c_{k+1}\mathbf{B}\mathbf{v}_{k+1} + \dots + c_m\mathbf{B}\mathbf{v}_m = \mathbf{0}.$$

Then  $\mathbf{B}(c_{k+1}\mathbf{v}_{k+1} + \dots + c_m\mathbf{v}_m) = \mathbf{0}$ , which implies

$$c_{k+1}\mathbf{v}_{k+1} + \dots + c_m\mathbf{v}_m \in \text{nullspace of } \mathbf{B} = \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_k\}.$$

So  $c_{k+1} = \dots = c_m = 0$ . Hence,  $\mathbf{B}\mathbf{v}_{k+1}, \dots, \mathbf{B}\mathbf{v}_m$  are linearly independent. Then

$$\text{nullity}(\mathbf{A}) \geq m - k = \text{nullity}(\mathbf{AB}) - \text{nullity}(\mathbf{B}).$$

So

$$\text{nullity}(\mathbf{A}) + \text{nullity}(\mathbf{B}) \geq \text{nullity}(\mathbf{AB}).$$

**Orthogonal Matrix**2014-2015-Semester 2

True or false: Given any  $2 \times 3$  matrix  $\mathbf{M}$  and  $2 \times 1$  column vector  $\mathbf{c}$ , the linear system  $\mathbf{M}\mathbf{x} = \mathbf{c}$  always has infinitely many least squares solutions. Justify your answer.

2015-2016-Semester 2 Use projection to extends orthogonal basis

Let  $V = \text{span}\{(1, 1, 0, 1), (3, 2, 1, 1), (-1, 0, 2, -2)\}$ .

- (i) [3 marks] Using the Gram-Schmidt process, find an orthogonal basis for  $V$ .
- (ii) [2 marks] Using the result in (i), find the projection of  $(2, -2, -2, 3)$  onto  $V$ .
- (iii) [2 marks] Extend your basis in (i) to an orthogonal basis for  $\mathbb{R}^4$ .

#### 2017-2018-Semester 2

Suppose  $\mathbf{v}_1 = (1, 1, 1, 1)$ ,  $\mathbf{v}_2 = (1, 2, 4, 5)$ , and  $\mathbf{v}_3 = (10, -30, -40, -20)$ . Let  $S = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  and let  $U = \text{span}(S)$ .

- (a) (6 points) Use the Gram-Schmidt process to find an orthonormal basis  $T$  for  $U$ .

(4 points) Find the transition matrix from  $S$  to  $T$ . You may leave the entries of your answer in terms of square roots.

#### **Diagonalizable**

#### 2014-2015-Semester 2

Prove the following:

Let  $\mathbf{M}$  and  $\mathbf{N}$  be two  $n \times n$  matrices. Suppose  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  is a set of linearly independent eigenvectors for both  $\mathbf{M}$  and  $\mathbf{N}$ . Then  $\mathbf{MN} = \mathbf{NM}$ .

#### 2017-2018-Semester 1

Suppose  $\mathbf{A}$  is diagonalizable and the eigenvalues of  $\mathbf{A}$  are  $\lambda_1, \lambda_2, \dots, \lambda_k$ .

- (i) If  $\mathbf{v}$  is an eigenvector of  $\mathbf{A}$ , say,  $\mathbf{Av} = \lambda_i \mathbf{v}$  for some  $i$ , show that  $(\mathbf{A} - \lambda_1 \mathbf{I})(\mathbf{A} - \lambda_2 \mathbf{I}) \cdots (\mathbf{A} - \lambda_k \mathbf{I})\mathbf{v} = \mathbf{0}$ .  
(Hint: First, show that  $(\mathbf{A} - \lambda_i \mathbf{I})\mathbf{v} = \mathbf{0}$  and then use the result in part (b).)

Since  $\mathbf{Av} = \lambda_i \mathbf{v} = \lambda_i \mathbf{Iv}$ ,

$$(\mathbf{A} - \lambda_i \mathbf{I})\mathbf{v} = \mathbf{Av} - \lambda_i \mathbf{Iv} = \mathbf{0}.$$

Applying the result in part (b) repeatedly yields

$$(\mathbf{A} - \lambda_1 \mathbf{I}) \cdots (\mathbf{A} - \lambda_i \mathbf{I}) \cdots (\mathbf{A} - \lambda_k \mathbf{I}) = (\mathbf{A} - \lambda_1 \mathbf{I}) \cdots (\mathbf{A} - \lambda_k \mathbf{I})(\mathbf{A} - \lambda_i \mathbf{I})$$

Therefore,

$$\begin{aligned} (\mathbf{A} - \lambda_1 \mathbf{I}) \cdots (\mathbf{A} - \lambda_i \mathbf{I}) \cdots (\mathbf{A} - \lambda_k \mathbf{I})\mathbf{v} &= (\mathbf{A} - \lambda_1 \mathbf{I}) \cdots (\mathbf{A} - \lambda_k \mathbf{I})(\mathbf{A} - \lambda_i \mathbf{I})\mathbf{v} \\ &= (\mathbf{A} - \lambda_1 \mathbf{I}) \cdots (\mathbf{A} - \lambda_k \mathbf{I})(\mathbf{0}) = \mathbf{0} \end{aligned}$$

- (ii) Define  $S = T_{\lambda_1} \circ T_{\lambda_2} \circ \cdots \circ T_{\lambda_k}$ . Prove that  $S$  is the zero transformation.

Since  $A$  is diagonalizable, by Theorem 6.2.3,  $A$  has  $n$  linearly independent eigenvectors, which will span  $\mathbb{R}^n$ . Let  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be one such basis.

$\forall \mathbf{v} \in \mathbb{R}^n, \exists a_1, \dots, a_n \in \mathbb{R}$ , such that  $\mathbf{v} = a_1 \mathbf{v}_1 + \dots + a_n \mathbf{v}_n = \sum_{k=1}^n a_k \mathbf{v}_k$ .

$$\begin{aligned} S(\mathbf{v}) &= S\left(\sum_{k=1}^n a_k \mathbf{v}_k\right) \\ &= \sum_{k=1}^n a_k S(\mathbf{v}_k) \\ &= \sum_{k=1}^n a_k \mathbf{0} \quad \text{from part (i)} \\ &= \mathbf{0} \end{aligned}$$

$R(S) = \{\mathbf{0}\}$ , and therefore,  $S$  is the zero transformation.

### 2017-2018-Semester 2

(8 points) Assume that  $\mathbf{A}$  is a symmetric matrix. Show that if  $m > 0$  and  $\mathbf{A}^m = \mathbf{I}$ , then  $\mathbf{A}^2 = \mathbf{I}$ .

**Solution:** Symmetric matrices are diagonalizable. There exist real numbers  $\lambda_1, \dots, \lambda_n$  and an invertible matrix  $\mathbf{P}$  such that

$$\mathbf{P}^{-1} \mathbf{A} \mathbf{P} = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} = \mathbf{B}.$$

Note that each  $\lambda_i$  is an eigenvalue of  $\mathbf{A}$ . Hence by Part (b), each  $\lambda_i$  is either 1 or  $-1$ . Therefore  $\lambda_i^2 = 1$ , for each  $1 \leq i \leq n$ . So we have

$$\mathbf{B}^2 = \begin{pmatrix} \lambda_1^2 & & \\ & \ddots & \\ & & \lambda_n^2 \end{pmatrix} = \begin{pmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{pmatrix} = \mathbf{I}.$$

So by Part (c),  $\mathbf{A}^2 = \mathbf{I}$ .

### **Linear Transformation**

#### 2016-2017-Semester 2

(a) Let  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  be a linear transformation such that

$$T\left(\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}\right) = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \quad T\left(\begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}\right) = \begin{pmatrix} 2 \\ 1 \\ -1 \end{pmatrix}, \quad T\left(\begin{pmatrix} 3 \\ 3 \\ 2 \end{pmatrix}\right) = \begin{pmatrix} -2 \\ 1 \\ 3 \end{pmatrix}.$$

(i) Find the standard matrix for  $T$ .

(ii) Find  $\text{rank}(T)$  and  $\text{nullity}(T)$ .

#### 2015-2016-Semester 2

Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be a linear operator such that its standard matrix is diagonalisable. Prove that  $R(T) = R(T \circ T)$  and  $\text{Ker}(T) = \text{Ker}(T \circ T)$ .

Let  $\mathbf{A}$  be the standard matrix for  $T$ .

Then  $\mathbf{A}$  has  $n$  linearly independent eigenvectors, say  $\mathbf{v}_1, \dots, \mathbf{v}_n$ , associated to eigenvalues  $\lambda_1, \dots, \lambda_n$ , respectively.

Suppose that  $\lambda_1 = \dots = \lambda_k = 0$ , and  $\lambda_i \neq 0$  if  $i > k$ .

For each  $i$ ,  $\mathbf{A}\mathbf{v}_i = \lambda\mathbf{v}_i$ , and thus  $\mathbf{A}^2\mathbf{v}_i = \lambda^2\mathbf{v}_i$ ; so  $\mathbf{v}_1, \dots, \mathbf{v}_n$  are the eigenvectors of  $\mathbf{A}^2$  associated to eigenvalues  $\lambda_1^2, \dots, \lambda_n^2$ .

Note that  $\lambda_i^2 = 0$  if  $i \leq k$  and  $\lambda_i^2 \neq 0$  if  $i > k$ . Then

$$\begin{aligned}\text{Ker}(T) &= \text{nullspace of } \mathbf{A} = \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_k\} \\ &= \text{nullspace of } \mathbf{A}^2 = \text{Ker}(T \circ T).\end{aligned}$$

$R(T \circ T) = \text{column space of } \mathbf{A}^2 \subseteq \text{column space of } \mathbf{A} = R(T)$ , and

$$\dim R(T \circ T) = n - \dim \text{Ker}(T) = n - \dim \text{Ker}(T \circ T) = \dim R(T).$$

We conclude that  $R(T \circ T) = R(T)$ .

### 2016-2017-Semester 2

Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be a linear transformation such that  $\text{Ker}(T) = \text{Ker}(T \circ T)$ . Prove that

$$\text{Ker}(T \circ T) = \text{Ker}(T \circ T \circ T).$$

If  $\mathbf{v} \in \text{Ker}(T \circ T)$ , then  $T \circ T(\mathbf{v}) = \mathbf{0}$ ; consequently  $T \circ T \circ T(\mathbf{v}) = T(T \circ T(\mathbf{v})) = T(\mathbf{0}) = \mathbf{0}$ , i.e.,  $\mathbf{v} \in \text{Ker}(T \circ T \circ T)$ .

Let  $\mathbf{v} \in \text{Ker}(T \circ T \circ T)$ . Set  $\mathbf{w} = T(\mathbf{v})$ . Then

$$T \circ T(\mathbf{w}) = T \circ T(T(\mathbf{v})) = T \circ T \circ T(\mathbf{v}) = \mathbf{0}.$$

So  $\mathbf{w} \in \text{Ker}(T \circ T) = \text{Ker}(T)$ . We thus have

$$T \circ T(\mathbf{v}) = T(T(\mathbf{v})) = T(\mathbf{w}) = \mathbf{0};$$

that is,  $\mathbf{v} \in \text{Ker}(T \circ T)$ .

## Subspace

2014-2015-Sem2

(i)

$$V = \{(a, a, a, 0) \mid a \in \mathbb{R}\} = \{a(1, 1, 1, 0) \mid a \in \mathbb{R}\} = \text{span}\{(1, 1, 1, 0)\}.$$

Thus  $\{(1, 1, 1, 0)\}$  is a basis for  $V$  and  $\dim(V) = 1$ .

(ii) Let  $W = \text{span}\{(1, 1, 1, 0), (0, 1, 0, 0), (0, 0, 1, 0)\}$ . Since  $V = \text{span}\{(1, 1, 1, 0)\}$ , we have  $V \subseteq W$  and  $W \cap V = V$  and thus  $\dim(W \cap V) = 1$ . On the other hand,

$$a(1, 1, 1, 0) + b(0, 1, 0, 0) + c(0, 0, 1, 0) = (0, 0, 0, 0) \Rightarrow a = b = c = 0.$$

Thus  $\{(1, 1, 1, 0), (0, 1, 0, 0), (0, 0, 1, 0)\}$  is a linearly independent set and  $\dim(W) = 3$ .

(iii) Since  $V = \text{span}\{(1, 1, 1, 0)\}$ ,  $\mathbf{u} = (u_1, u_2, u_3, u_4) \in U$  if and only if  $(u_1, u_2, u_3, u_4) \cdot (1, 1, 1, 0) = 0$ , which implies  $u_1 + u_2 + u_3 = 0$ . Solving, we have

$$\begin{cases} u_1 = -s - t \\ u_2 = s \\ u_3 = t \\ u_4 = r, \quad s, t, r \in \mathbb{R}. \end{cases}$$

So

$$U = \text{span} \left\{ \begin{pmatrix} -1 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \right\}.$$

$$\text{A basis for } U \text{ is } \left\{ \begin{pmatrix} -1 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \right\} \text{ and } \dim(U) = 3.$$

## Rank Problem

2014-2015-sem2

Yes. The system has exactly one solution. First of all, observe that the 3 columns of the coefficient matrix of the system are precisely  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  and the constant matrix is  $\mathbf{w}$ . From (ii), we know  $\mathbf{w}$  belongs to the column space  $V = \text{span}(S)$  of the matrix, and hence the system is consistent. Furthermore,  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  is a basis for  $V$ , hence  $\mathbf{w}$  can only be expressed as a linear combination of this basis in exactly one way. Hence the system has exactly one solution.

2015-2016-Semester 2

(i) We check that

$$\begin{pmatrix} -2 & -14 & 14 \\ 5 & 15 & -10 \\ 4 & 8 & -3 \end{pmatrix} \xrightarrow{\substack{-\frac{1}{2}R_1 \\ \frac{1}{5}R_2}} \begin{pmatrix} 1 & 7 & -7 \\ 1 & 3 & -2 \\ 4 & 8 & -3 \end{pmatrix} \xrightarrow{\substack{R_2 - R_1 \\ R_3 - 4R_1}} \begin{pmatrix} 1 & 7 & -7 \\ 0 & -4 & 5 \\ 0 & -20 & 25 \end{pmatrix} \xrightarrow{R_3 - 5R_2} \begin{pmatrix} 1 & 7 & -7 \\ 0 & -4 & 5 \\ 0 & 0 & 0 \end{pmatrix}$$

So a basis for the row space is given by  $\{(1, 7, -7), (0, -4, 5)\}$  and  $\text{rank}(\mathbf{AB}) = 2$ .

(ii) One verifies that:

$$(\mathbf{AB})^2 = \begin{pmatrix} -10 & -70 & 70 \\ 25 & 75 & -50 \\ 20 & 40 & -15 \end{pmatrix} = 5\mathbf{AB}.$$

(iii)  $\text{rank}(\mathbf{BA}) \geq \text{rank}(\mathbf{A}(\mathbf{BA})\mathbf{B}) = \text{rank}((\mathbf{AB})^2) = 2$ .

Since  $\mathbf{BA}$  is  $2 \times 2$ , so  $\text{rank}(\mathbf{BA}) = 2$ .

(iv) From (ii),

$$(\mathbf{BA})^3 = \mathbf{BABABA} = \mathbf{B}(\mathbf{AB})^2\mathbf{A} = \mathbf{B}(5\mathbf{AB})\mathbf{A} = 5(\mathbf{BA})^2.$$

From (iii), we have  $\mathbf{BA}$  is full rank and invertible.

$$\text{It follows that } \mathbf{BA} = 5\mathbf{I} = \begin{pmatrix} 5 & 0 \\ 0 & 5 \end{pmatrix}.$$

### Orthogonal Matrix

#### 2014-2015-sem2

True. If  $\mathbf{M}$  is  $2 \times 3$ , the rank of  $\mathbf{M}$  is at most 2, and hence the nullity of  $\mathbf{M}$  is at least 1.

This implies the nullity of  $\mathbf{M}^T\mathbf{M}$  is at least 1. Hence  $\mathbf{M}^T\mathbf{M}$  is not invertible.

Since  $\mathbf{M}^T\mathbf{M}\mathbf{x} = \mathbf{M}^T\mathbf{c}$  is always consistent, this means it will have infinitely many solutions. i.e.  $\mathbf{M}\mathbf{x} = \mathbf{c}$  always has infinitely many least squares solutions.

#### 2015-2016-Sem2

(i) Let  $\mathbf{u}_1 = (1, 1, 0, 1)$ ,  $\mathbf{u}_2 = (3, 2, 1, 1)$  and  $\mathbf{u}_3 = (-1, 0, 2, -2)$ .

$$\begin{aligned} \mathbf{v}_1 &= \mathbf{u}_1 = (1, 1, 0, 1), \\ \mathbf{v}_2 &= \mathbf{u}_2 - \frac{\mathbf{u}_2 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 = (3, 2, 1, 1) - \frac{6}{3}(1, 1, 0, 1) = (1, 0, 1, -1), \\ \mathbf{v}_3 &= \mathbf{u}_3 - \frac{\mathbf{u}_3 \cdot \mathbf{v}_1}{\mathbf{v}_1 \cdot \mathbf{v}_1} \mathbf{v}_1 - \frac{\mathbf{u}_3 \cdot \mathbf{v}_2}{\mathbf{v}_2 \cdot \mathbf{v}_2} \mathbf{v}_2 \\ &= (-1, 0, 2, -2) - \frac{-3}{3}(1, 1, 0, 1) - \frac{3}{3}(1, 0, 1, -1) = (-1, 1, 1, 0). \end{aligned}$$

An orthogonal basis for  $V$  is  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ .

(ii) The projection of  $\mathbf{v} = (2, -2, -2, 3)$  onto  $V$  is

$$\begin{aligned} \mathbf{p} &= \frac{(\mathbf{v} \cdot \mathbf{v}_1)}{(\mathbf{v}_1 \cdot \mathbf{v}_1)} \mathbf{v}_1 + \frac{(\mathbf{v} \cdot \mathbf{v}_2)}{(\mathbf{v}_2 \cdot \mathbf{v}_2)} \mathbf{v}_2 + \frac{(\mathbf{v} \cdot \mathbf{v}_3)}{(\mathbf{v}_3 \cdot \mathbf{v}_3)} \mathbf{v}_3 \\ &= \frac{3}{3}(1, 1, 0, 1) + \frac{-3}{3}(1, 0, 1, -1) + \frac{-6}{3}(-1, 1, 1, 0) = (2, -1, -3, 2). \end{aligned}$$

(iii) Note that  $\mathbf{v} - \mathbf{p} = (0, -1, 1, 1)$  is orthogonal to each  $\mathbf{v}_i$ . Then

$$\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, (0, -1, 1, 1)\}$$

is an orthogonal basis for  $\mathbb{R}^4$ .

2017-2018-Semester 2

1-(b)

**Solution:**  $[\mathbf{v}_1]_T = \begin{pmatrix} \mathbf{v}_1 \cdot \mathbf{u}_1 \\ \mathbf{v}_1 \cdot \mathbf{u}_2 \\ \mathbf{v}_1 \cdot \mathbf{u}_3 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \\ 0 \end{pmatrix}$ .  $[\mathbf{v}_2]_T = \begin{pmatrix} \mathbf{v}_2 \cdot \mathbf{u}_1 \\ \mathbf{v}_2 \cdot \mathbf{u}_2 \\ \mathbf{v}_2 \cdot \mathbf{u}_3 \end{pmatrix} = \begin{pmatrix} 6 \\ \sqrt{10} \\ 0 \end{pmatrix}$ .

$[\mathbf{v}_3]_T = \begin{pmatrix} \mathbf{v}_3 \cdot \mathbf{u}_1 \\ \mathbf{v}_3 \cdot \mathbf{u}_2 \\ \mathbf{v}_3 \cdot \mathbf{u}_3 \end{pmatrix} = \begin{pmatrix} -40 \\ -7\sqrt{10} \\ \sqrt{910} \end{pmatrix}$ . So the transition matrix from  $S$  to  $T$

is:

$$\mathbf{P} = \begin{pmatrix} 2 & 6 & -40 \\ 0 & \sqrt{10} & -7\sqrt{10} \\ 0 & 0 & \sqrt{910} \end{pmatrix}.$$

## Linear Transformation

2016-2017-Sem2: Use  $\mathbf{A}^{-1}$  to determine the standard matrix!

(i)  $\mathbf{A} \begin{pmatrix} 1 & 0 & 3 \\ 2 & 1 & 3 \\ 3 & 2 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 2 & -2 \\ 1 & 1 & 1 \\ 0 & -1 & 3 \end{pmatrix}$ .  $\mathbf{A} = \begin{pmatrix} 1 & 2 & -2 \\ 1 & 1 & 1 \\ 0 & -1 & 3 \end{pmatrix} \begin{pmatrix} 4 & -6 & 3 \\ -5 & 7 & -3 \\ -1 & 2 & -1 \end{pmatrix} = \begin{pmatrix} -4 & 4 & -1 \\ -2 & 3 & -1 \\ 2 & -1 & 0 \end{pmatrix}$ .

(ii)  $\mathbf{A} \xrightarrow[R_3 + \frac{1}{2}R_1]{R_2 - \frac{1}{2}R_1} \begin{pmatrix} -4 & 4 & -1 \\ 0 & 1 & -\frac{1}{2} \\ 0 & 1 & -\frac{1}{2} \end{pmatrix} \xrightarrow{R_3 - R_2} \begin{pmatrix} -4 & 4 & -1 \\ 0 & 1 & -\frac{1}{2} \\ 0 & 0 & 0 \end{pmatrix}$ .

So  $\text{rank}(T) = 2$  and  $\text{nullity}(T) = 3 - 2 = 1$ .