



# MATH1231/41 Algebra Part 1

## Revision Session 2019 T2 Solutions

August 11, 2019

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### Example 1: 1231 2015 Q1.v

*Prove that*

$$S = \left\{ \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \in \mathbb{R}^3 : x_1 - 2x_2 + 4x_3 = 0 \right\}$$

*is a subspace of  $\mathbb{R}^3$ .*

Clearly,  $S \subseteq \mathbb{R}^3$  where  $\mathbb{R}^3$  is a known vector space. Since  $0 - 2(0) + 4(0) = 0$ ,  $\mathbf{0} \in S$ . So  $S$  contains a zero element.

Now suppose that  $\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \in S$  and  $\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \in S$ . Then

$$x_1 - 2x_2 + 4x_3 = 0, \quad (1)$$

$$y_1 - 2y_2 + 4y_3 = 0. \quad (2)$$

(1) + (2) gives us  $(x_1 + y_1) - 2(x_2 + y_2) + 4(x_3 + y_3) = 0$ , i.e.  $\begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \\ x_3 + y_3 \end{pmatrix} \in S$ . Hence  $(\mathbf{x} + \mathbf{y}) \in S$

and so  $S$  is closed under vector addition.

If  $\lambda \in \mathbb{R}$  and  $\mathbf{x} \in S$  then  $\lambda \times (1)$  gives us  $\lambda x_1 - 2\lambda x_2 + 4\lambda x_3 = 0$ , i.e.  $\begin{pmatrix} \lambda x_1 \\ \lambda x_2 \\ \lambda x_3 \end{pmatrix} \in S$ . Hence  $\lambda \mathbf{x} \in S$

so  $S$  is closed under scalar multiplication.

Since  $S$  is a subset of  $\mathbb{R}^3$  and contains a zero element, is closed under vector addition, and is closed under scalar multiplication, then by the Subspace Theorem  $S$  is a subspace of  $\mathbb{R}^3$ .

## Example 2: 1231 2013 Q1.i

Let

$$S = \left\{ \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \in \mathbb{R}^3 : x_1^3 + x_2^3 + x_3^3 = 0 \right\}.$$

a) Prove that  $S$  is closed under scalar multiplication.

b) Show that  $S$  is **not** a subspace of  $\mathbb{R}^3$ .

Suppose that  $\mathbf{x} \in S$  and  $\lambda \in \mathbb{R}$ . Then we have

$$x_1^3 + x_2^3 + x_3^3 = 0.$$

Multiplying by  $\lambda^3$ ,

$$(\lambda x_1)^3 + (\lambda x_2)^3 + (\lambda x_3)^3 = 0.$$

Hence  $\lambda \mathbf{x} \in S$ , i.e.  $S$  is closed under scalar multiplication.

Note that  $\mathbf{0}$  is an element of  $S$ , so to prove that  $S$  is not a subspace we will show that  $S$  is not closed under vector addition. Take  $\mathbf{x} = (1, -1, 0)^T$  and  $\mathbf{y} = (-2, 0, 2)^T$ . Clearly  $\mathbf{x}, \mathbf{y} \in S$ , but

$$\mathbf{x} + \mathbf{y} = \begin{pmatrix} -1 \\ -1 \\ 2 \end{pmatrix} \notin S.$$

Hence  $S$  is not closed under vector addition. By the Subspace Theorem,  $S$  is not a subspace of  $\mathbb{R}^3$ .

### Example 3: 1231 2015 Q1.vi

Consider the vectors in  $\mathbb{R}^3$ ,

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ -1 \\ 2 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} 1 \\ 2 \\ 5 \end{pmatrix}, \mathbf{v}_3 = \begin{pmatrix} 2 \\ -3 \\ 3 \end{pmatrix}, \mathbf{b} = \begin{pmatrix} -1 \\ 6 \\ 3 \end{pmatrix}.$$

Prove that  $\mathbf{b} \in \text{span}(\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3)$ .

We want examine the nature of solutions  $(x_1, x_2, x_3)^T$  to

$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + x_3\mathbf{v}_3 = \mathbf{b}.$$

$\mathbf{b} \in \text{span}(\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3)$  if there is at least one solution. Notice that our equation can be written in the form  $A\mathbf{x} = \mathbf{b}$ :

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

Applying Gaussian elimination, we have

$$\left( \begin{array}{ccc|c} 1 & 1 & 2 & -1 \\ -1 & 2 & -3 & 6 \\ 2 & 5 & 3 & 3 \end{array} \right) \sim \left( \begin{array}{ccc|c} 1 & 1 & 2 & -1 \\ 0 & 3 & -1 & 5 \\ 0 & 0 & 0 & 0 \end{array} \right).$$

Clearly there are infinitely many solutions, so  $\mathbf{b} \in \text{span}(\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3)$ .

### Example 4

Let  $\mathbb{P}_2$  be the vector space of all real polynomials of degree at most 2. Find three polynomials  $f_1, f_2, f_3$  in  $\mathbb{P}_2$  such that  $f_i(0) = 1$  for  $i = 1, 2, 3$  and  $\{f_1, f_2, f_3\}$  is linearly independent.

We want three polynomials such that each has a different span. The easiest way to do this is

to consider the set of functions

$$\begin{aligned}f_1 &= a_1, \\f_2 &= a_2 + b_2x, \\f_3 &= a_3 + b_3x + c_3x^2.\end{aligned}$$

Since  $f_i(0) = 1$  then  $a_i = 1$ . The other coefficients are arbitrary constants, so set all other constants to 1:

$$\begin{aligned}f_1 &= 1, \\f_2 &= 1 + x, \\f_3 &= 1 + x + x^2.\end{aligned}$$

### Example 5: 1241 2016 Q3.iii

The field  $\mathbb{F} = GF(4)$  has elements  $\{0, 1, \alpha, \beta\}$  with addition and multiplication defined by the following tables. For the vectors

+	0	1	$\alpha$	$\beta$
0	0	1	$\alpha$	$\beta$
1	1	0	$\beta$	$\alpha$
$\alpha$	$\alpha$	$\beta$	0	1
$\beta$	$\beta$	$\alpha$	1	0

$\times$	0	1	$\alpha$	$\beta$
0	0	0	0	0
1	0	1	$\alpha$	$\beta$
$\alpha$	0	$\alpha$	$\beta$	1
$\beta$	0	$\beta$	1	$\alpha$

$$\mathbf{b}_1 = \begin{pmatrix} 1 \\ \alpha \\ \beta \end{pmatrix}, \mathbf{b}_2 = \begin{pmatrix} \beta \\ 1 \\ 1 \end{pmatrix}, \mathbf{b}_3 = \begin{pmatrix} 1 \\ 0 \\ \alpha \end{pmatrix},$$

- a) show that  $\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$  is a basis for  $\mathbb{F}^3$ ;  
b) explain why  $\{\mathbf{b}_1, \mathbf{b}_1 + \mathbf{b}_2, \mathbf{b}_2 + \mathbf{b}_3, \mathbf{b}_3\}$  is a spanning set but not a basis for  $\mathbb{F}^3$ .

First we prove two important results. Suppose  $\mathbf{x} \in \mathbb{F}^3$ . Since  $a + a = 0 \ \forall a \in \mathbb{F}$ , then

$$\mathbf{x} + \mathbf{x} = \mathbf{0} \quad \forall \mathbf{x} \in \mathbb{F}^3. \quad (*)$$

Also, since  $a + 0 = a \ \forall a \in \mathbb{F}$ , then

$$\mathbf{x} + \mathbf{0} = \mathbf{x} \quad \forall \mathbf{x} \in \mathbb{F}^3. \quad (**)$$

Now, note that  $\dim \mathbb{F}^3 = 3$ . If we can show that  $\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$  is a linearly independent set then

we can apply the Dimension Theorem. For  $x, y, z \in \mathbb{F}$ , if

$$x\mathbf{b}_1 + y\mathbf{b}_2 + z\mathbf{b}_3 = \mathbf{0}$$

then we can represent this in the form  $A\mathbf{x} = \mathbf{b}$ :

$$\begin{pmatrix} 1 & \beta & 1 \\ \alpha & 1 & 0 \\ \beta & 1 & \alpha \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}.$$

Carefully applying Gaussian elimination:

$$\begin{aligned} \left( \begin{array}{ccc|c} 1 & \beta & 1 & 0 \\ \alpha & 1 & 0 & 0 \\ \beta & 1 & \alpha & 0 \end{array} \right) &\xrightarrow{R1=\alpha R1} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & 0 \\ \alpha & 1 & 0 & 0 \\ \beta & 1 & \alpha & 0 \end{array} \right) \xrightarrow{R2=R2+R1} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & 0 \\ 0 & 0 & \alpha & 0 \\ \beta & 1 & \alpha & 0 \end{array} \right) \\ &\xrightarrow{R3=\beta R3} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & 0 \\ 0 & 0 & \alpha & 0 \\ \alpha & \beta & 1 & 0 \end{array} \right) \xrightarrow{R2 \leftrightarrow R3} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & 0 \\ \alpha & \beta & 1 & 0 \\ 0 & 0 & \alpha & 0 \end{array} \right) \\ &\xrightarrow{R2=R2+R1} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & 0 \\ 0 & \alpha & \beta & 0 \\ 0 & 0 & \alpha & 0 \end{array} \right). \end{aligned}$$

Clearly, our solution  $(x, y, z)^T = (0, 0, 0)^T$ . So the only solution is the trivial solution, hence  $\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$  is a linearly independent set. Therefore  $\dim(\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}) = \dim(\mathbb{F}^3) = 3$ , and so by the Dimension Theorem  $\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$  is a basis for  $\mathbb{F}^3$ .

Now consider the second set of vectors in  $\mathbb{F}^3$ .

$$\begin{aligned} (\mathbf{b}_1 + \mathbf{b}_2) + (\mathbf{b}_2 + \mathbf{b}_3) + (\mathbf{b}_3) &= \mathbf{b}_1 + (\mathbf{b}_2 + \mathbf{b}_2) + (\mathbf{b}_3 + \mathbf{b}_3) && \text{(associative law)} \\ &= \mathbf{b}_1 + \mathbf{0} + \mathbf{0} && \text{(Using (*))} \\ &= \mathbf{b}_1. && \text{(Using (**))} \end{aligned}$$

Since we have written  $\mathbf{b}_1$  as a linear combination of  $\mathbf{b}_1 + \mathbf{b}_2$ ,  $\mathbf{b}_2 + \mathbf{b}_3$  and  $\mathbf{b}_3$ , then

$$\{\mathbf{b}_1, \mathbf{b}_1 + \mathbf{b}_2, \mathbf{b}_2 + \mathbf{b}_3, \mathbf{b}_3\}$$

is a linearly dependent set. Hence the set cannot be a basis for  $\mathbb{F}^3$ . However since

$$\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3 \in \text{span}(\{\mathbf{b}_1, \mathbf{b}_1 + \mathbf{b}_2, \mathbf{b}_2 + \mathbf{b}_3, \mathbf{b}_3\})$$

and  $\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$  is a spanning set for  $\mathbb{F}^3$ , then  $\{\mathbf{b}_1, \mathbf{b}_1 + \mathbf{b}_2, \mathbf{b}_2 + \mathbf{b}_3, \mathbf{b}_3\}$  is a spanning set for  $\mathbb{F}^3$ .

## Example 6: 1241 2016 Q3.iii

Consider the field  $\mathbb{F} = GF(4)$ , as defined in the previous example. Let  $\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3$  be the vectors from the previous example. Set

$$\mathbf{v} = \begin{pmatrix} \alpha \\ 0 \\ 0 \end{pmatrix}.$$

c) Find the coordinate vector of  $\mathbf{v}$  with respect to the ordered basis  $B = \{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$ .

The coordinate vector  $\mathbf{x} = (x, y, z)^T$  of  $\mathbf{v}$  will satisfy the equation

$$x\mathbf{b}_1 + y\mathbf{b}_2 + z\mathbf{b}_3 = \mathbf{v}.$$

Writing this in the form  $A\mathbf{x} = \mathbf{b}$ , we have

$$\begin{pmatrix} 1 & \beta & 1 \\ \alpha & 1 & 0 \\ \beta & 1 & \alpha \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} \alpha \\ 0 \\ 0 \end{pmatrix}.$$

Carefully applying Gaussian elimination:

$$\begin{aligned} \left( \begin{array}{ccc|c} 1 & \beta & 1 & \alpha \\ \alpha & 1 & 0 & 0 \\ \beta & 1 & \alpha & 0 \end{array} \right) &\xrightarrow{R1=\alpha R1} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & \beta \\ \alpha & 1 & 0 & 0 \\ \beta & 1 & \alpha & 0 \end{array} \right) \xrightarrow{R2=R2+R1} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & \beta \\ 0 & 0 & \alpha & \beta \\ \beta & 1 & \alpha & 0 \end{array} \right) \\ &\xrightarrow{R3=\beta R3} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & \beta \\ 0 & 0 & \alpha & \beta \\ \alpha & \beta & 1 & 0 \end{array} \right) \xrightarrow{R2 \leftrightarrow R3} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & \beta \\ \alpha & \beta & 1 & 0 \\ 0 & 0 & \alpha & \beta \end{array} \right) \\ &\xrightarrow{R2=R2+R1} \left( \begin{array}{ccc|c} \alpha & 1 & \alpha & \beta \\ 0 & \alpha & \beta & \beta \\ 0 & 0 & \alpha & \beta \end{array} \right). \end{aligned}$$

From  $R3$  we have  $z = \alpha$  since  $\alpha \times \alpha = \beta$ . In  $R2$  we have

$$\begin{aligned} \alpha y + \beta \times \alpha &= \beta \\ \alpha y + 1 &= \beta && (\beta \times \alpha = 1) \\ \alpha y + 1 + 1 &= \beta + 1 && (\text{Adding 1 to both sides}) \\ \alpha y &= \alpha && (1 + 1 = 0 \text{ and } \beta + 1 = \alpha) \\ y &= 1. && (\text{Since } \alpha \times 1 = \alpha) \end{aligned}$$

In  $R1$  we have

$$\begin{array}{ll}
 \alpha x + 1 \times 1 + \alpha \times \alpha = \beta & \\
 \alpha x + 1 + \beta = \beta & (1 \times 1 = 1 \text{ and } \alpha \times \alpha = \beta) \\
 \alpha x + (1 + \beta) + (1 + \beta) = \beta + 1 + \beta & (\text{Adding } (1 + \beta) \text{ to both sides}) \\
 \alpha x = \alpha + \beta & ((1 + \beta) + (1 + \beta) = 0 \text{ and } \beta + 1 = \alpha) \\
 \alpha x = 1 & (\alpha + \beta = 1) \\
 x = \beta & (\text{Since } \alpha \times \beta = 1)
 \end{array}$$

Hence the coordinate vector of  $\mathbf{v}$  with respect to the basis  $B$  is

$$\mathbf{x} = \begin{pmatrix} \beta \\ 1 \\ \alpha \end{pmatrix}.$$

### Example 7: 1241 2014 S2 Q3.i

Prove that the function  $T : \mathbb{P}(\mathbb{R}) \rightarrow \mathbb{R}^2$  defined by

$$T(p) = \begin{pmatrix} p(0) \\ p(1) \end{pmatrix}, \text{ for all polynomials } p \in \mathbb{P}(\mathbb{R}),$$

is a linear transformation.

For the map  $T$  to be linear, we need to show that  $T$  preserves addition and scalar multiplication. First consider addition. For any  $p, q \in \mathbb{P}(\mathbb{R})$ ,

$$\begin{aligned}
 T(p + q) &= \begin{pmatrix} (p + q)(0) \\ (p + q)(1) \end{pmatrix} = \begin{pmatrix} p(0) + q(0) \\ p(1) + q(1) \end{pmatrix} = \begin{pmatrix} p(0) \\ p(1) \end{pmatrix} + \begin{pmatrix} q(0) \\ q(1) \end{pmatrix} \\
 &= T(p) + T(q).
 \end{aligned}$$

So  $T$  preserves addition. Now consider scalar multiplication. For any  $p \in \mathbb{P}(\mathbb{R})$  and  $\lambda \in \mathbb{R}$ ,

$$\begin{aligned}
 T(\lambda p) &= \begin{pmatrix} (\lambda p)(0) \\ (\lambda p)(1) \end{pmatrix} = \begin{pmatrix} \lambda p(0) \\ \lambda p(1) \end{pmatrix} = \lambda \begin{pmatrix} p(0) \\ p(1) \end{pmatrix} \\
 &= \lambda T(p).
 \end{aligned}$$

Hence  $T$  preserves scalar multiplication. Therefore since  $T$  preserves addition and scalar multiplication, then  $T$  is linear.

### Example 8: 1241 2016 Q3.ii

Let  $V$  and  $W$  be vector spaces, let  $T : V \rightarrow W$  be a linear transformation, and let  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  be vectors in  $V$ .

- a) Prove that if  $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_m)$  are linearly independent, then  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  are linearly independent.
- b) Suppose that  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  are linearly independent. Is  $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_m)$  linearly independent?

Suppose that  $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_m)$  are linearly independent, and assume  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  are linearly dependent. Then for some  $i \in \{1, 2, \dots, m\}$  and constants  $\lambda_j$ , we have

$$\mathbf{v}_i = \sum_{j \neq i} \lambda_j \mathbf{v}_j.$$

So then

$$\begin{aligned} T(\mathbf{v}_i) &= T\left(\sum_{j \neq i} \lambda_j \mathbf{v}_j\right) \\ &= \sum_{j \neq i} T(\lambda_j \mathbf{v}_j) && \text{(Since } T \text{ preserves addition)} \\ &= \sum_{j \neq i} \lambda_j T(\mathbf{v}_j). && \text{(Since } T \text{ preserves scalar multiplication)} \end{aligned}$$

Hence we have a contradiction, since  $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_m)$  are linearly independent. Hence our assumption is incorrect, i.e.  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  are not linearly dependent. So we have proven, by contradiction, that  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  are linearly independent.

However, linear independence of  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_m$  does not imply linear independence of  $T(\mathbf{v}_1), T(\mathbf{v}_2), \dots, T(\mathbf{v}_m)$ . Consider, for example, the linear map  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  defined by

$$T(\mathbf{x}) = \mathbf{0} \quad \forall \mathbf{x} \in \mathbb{R}^2.$$

The standard basis vectors  $\mathbf{e}_1, \mathbf{e}_2$  are linearly independent, however  $xT(\mathbf{e}_1) + yT(\mathbf{e}_2) = \mathbf{0}$  for any choice of  $x, y \in \mathbb{R}$ . So  $T(\mathbf{e}_1), T(\mathbf{e}_2)$  are not linearly independent. Interestingly enough, part b would be true if  $T$  were injective.

### Example 9: 1231 2013 Q2,iv

Let  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the linear map which rotates a vector  $\mathbf{x}$  about the origin through  $\frac{\pi}{6}$  anti-clockwise and doubles its length.



- a) Show that  $T(\mathbf{e}_1) = \begin{pmatrix} \sqrt{3} \\ 1 \end{pmatrix}$ , where  $\mathbf{e}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ .
- b) Find the matrix  $A$  such that  $T(\mathbf{x}) = A\mathbf{x}$  for all  $\mathbf{x} \in \mathbb{R}^2$ .

Since  $T$  rotates a vector  $(x, y)^T$  anticlockwise by  $\frac{\pi}{6}$ , we know that for  $x > 0$  and  $y \geq 0$ ,

$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} L \cos \left( \frac{\pi}{6} + \tan^{-1} \left( \frac{y}{x} \right) \right) \\ L \sin \left( \frac{\pi}{6} + \tan^{-1} \left( \frac{y}{x} \right) \right) \end{pmatrix}.$$

Since  $T$  also doubles the length of a vector  $(x, y)^T$ , then  $L = 2\sqrt{x^2 + y^2}$ . Hence

$$T \begin{pmatrix} x \\ y \end{pmatrix} = 2\sqrt{x^2 + y^2} \begin{pmatrix} \cos \left( \frac{\pi}{6} + \tan^{-1} \left( \frac{y}{x} \right) \right) \\ \sin \left( \frac{\pi}{6} + \tan^{-1} \left( \frac{y}{x} \right) \right) \end{pmatrix}.$$

So

$$\begin{aligned} T \begin{pmatrix} 1 \\ 0 \end{pmatrix} &= 2\sqrt{1+0} \begin{pmatrix} \cos \left( \frac{\pi}{6} + \tan^{-1} 0 \right) \\ \sin \left( \frac{\pi}{6} + \tan^{-1} 0 \right) \end{pmatrix} = 2 \begin{pmatrix} \sqrt{3}/2 \\ 1/2 \end{pmatrix} \\ &= \begin{pmatrix} \sqrt{3} \\ 1 \end{pmatrix}. \end{aligned}$$

Also,

$$\begin{aligned} T \begin{pmatrix} 0 \\ 1 \end{pmatrix} &= 2\sqrt{0+1} \begin{pmatrix} \cos \left( \frac{\pi}{6} + \frac{\pi}{2} \right) \\ \sin \left( \frac{\pi}{6} + \frac{\pi}{2} \right) \end{pmatrix} = 2 \begin{pmatrix} -1/2 \\ \sqrt{3}/2 \end{pmatrix} \\ &= \begin{pmatrix} -1 \\ \sqrt{3} \end{pmatrix}. \end{aligned}$$

Using the Matrix Representation Theorem,  $T(\mathbf{x}) = A\mathbf{x}$  where

$$A = \begin{pmatrix} \sqrt{3} & -1 \\ 1 & \sqrt{3} \end{pmatrix}.$$

### Example 10: 1231 2018 Q1.iv

Consider the matrix  $M = \begin{pmatrix} 1 & 1 \\ 2 & 2 \end{pmatrix}$ .

a) Find a basis for  $\ker(M)$ .

b) Find a basis for  $\text{im}(M^T)$ .

c) Give a geometric description of  $\ker(M)$  and  $\text{im}(M)$  as subspaces of  $\mathbb{R}^2$ .

If  $\mathbf{x} \in \ker(M)$  then  $M\mathbf{x} = \mathbf{0}$ . Hence

$$\begin{pmatrix} 1 & 1 \\ 2 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$
$$\begin{pmatrix} x + y \\ 2x + 2y \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}.$$

So  $x + y = 0$ , i.e.  $y = -x$ . So

$$\mathbf{x} = x \begin{pmatrix} 1 \\ -1 \end{pmatrix}.$$

Therefore

$$\ker(M) = \text{span} \left\{ \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\},$$

so a basis for  $\ker(M)$  is

$$\left\{ \begin{pmatrix} 1 \\ -1 \end{pmatrix} \right\}.$$

Now, consider  $\mathbf{y} \in \text{im}(M^T)$ . Then

$$\mathbf{y} = \begin{pmatrix} 1 & 2 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x + 2y \\ x + 2y \end{pmatrix} = (x + 2y) \begin{pmatrix} 1 \\ 1 \end{pmatrix}.$$

Therefore

$$\text{im}(M^T) = \text{span} \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\},$$

so a basis for  $\text{im}(M^T)$  is

$$\left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}.$$

The kernel of  $M$  is the line in  $\mathbb{R}^2$ , in the direction  $(1, -1)^T$ . The image of  $M$  is the line in  $\mathbb{R}^2$ , in the direction  $(1, 2)^T$ .

### Example 11: 1241 2015 Q3.ii

Consider the mapping  $T : \mathbb{P}_2 \rightarrow \mathbb{P}_3$  defined by

$$T(p)(x) = (x^2 + 1)p'(x) - 2xp(x).$$

Assuming  $T$  is linear, find the rank and nullity of  $T$ .

Let  $p(x) = ax^2 + bx + c$ . Then  $p'(x) = 2ax + b$ , and so

$$\begin{aligned} T(p)(x) &= (x^2 + 1)(2ax + b) - 2x(ax^2 + bx + c) \\ &= -bx^2 + 2(a - c)x + b. \end{aligned}$$

Hence

$$\begin{aligned} T(p) &= \begin{pmatrix} -b \\ 2a - 2c \\ b \end{pmatrix} = a \begin{pmatrix} 0 \\ 2 \\ 0 \end{pmatrix} + b \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix} + c \begin{pmatrix} 0 \\ -2 \\ 0 \end{pmatrix} \\ &= 2(a - c) \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + b \begin{pmatrix} -1 \\ 0 \\ 1 \end{pmatrix}. \end{aligned}$$

So if  $q \in \text{im}(T)$  then

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

So  $\text{rank}(T) = 2$ . Since  $\dim(\mathbb{P}_2) = 3$  (standard basis is  $\{1, x, x^2\}$ ), then by the Rank-Nullity Theorem,  $\text{nullity}(T) = 1$ .



# MATH1231/41 Algebra Part 2

## Revision Session 2019 T2 Solutions

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### Eigenvalues and Eigenvectors

QUESTION 1. A linear transformation  $P : V \rightarrow V$  is said to be idempotent if  $P(P(\mathbf{v})) = P(\mathbf{v})$  for all  $\mathbf{v} \in V$  (in other words  $P^2 = P$ ).

- (a) Show that the only possible eigenvalues for an idempotent linear transformation are 0 and 1.
- (b) Show that if  $P$  is idempotent and  $P$  is neither the zero nor the identity transformation on  $V$ , then both 0 and 1 are eigenvalues.

*Solution:*

- (a) Suppose  $\lambda$  is an eigenvalue of  $P$ , then  $P(\mathbf{v}) = \lambda\mathbf{v}$  where  $\mathbf{v}$  is nonzero. By the definition of idempotent,

$$\begin{aligned}
P(P(\mathbf{v})) &= P(\mathbf{v}) \\
P(\lambda\mathbf{v}) &= \lambda\mathbf{v} && \text{(since } P\mathbf{v} = \lambda\mathbf{v}) \\
\lambda P(\mathbf{v}) &= \lambda\mathbf{v} && \text{(P is linear)} \\
\lambda(\lambda\mathbf{v}) &= \lambda\mathbf{v} && \text{(Since } P\mathbf{v} = \lambda\mathbf{v}) \\
\lambda^2\mathbf{v} &= \lambda\mathbf{v} \\
(\lambda^2 - \lambda)\mathbf{v} &= \mathbf{0}.
\end{aligned}$$

Since  $\mathbf{v}$  is nonzero,  $\lambda^2 - \lambda = 0$ , so  $\lambda = 0, 1$ .

(b) Since  $P$  is not the zero map, there exists a  $\mathbf{v} \in V$  with  $P\mathbf{v}$  nonzero. So we can write  $P(\mathbf{v}) = \mathbf{w}$  where  $\mathbf{w} \in V$  and  $\mathbf{w}$  nonzero.

Applying  $P$  to both sides,

$$\begin{aligned}
P^2\mathbf{v} &= P(\mathbf{w}) \\
P(\mathbf{v}) &= P(\mathbf{w}) && \text{(from definition of idempotent)} \\
\mathbf{w} &= P(\mathbf{w}) && \text{(Since } P\mathbf{v} = \mathbf{w})
\end{aligned}$$

Thus  $P\mathbf{w} = \mathbf{w} = 1 \times \mathbf{w}$ . So 1 is an eigenvalue.

Now we prove that 0 is an eigenvalue.

Since  $P$  is not the identity map, there exists  $\mathbf{u} \in V$  with  $P\mathbf{u}$  not equal to  $\mathbf{u}$ . So we can write  $P\mathbf{u} = \mathbf{u} + \epsilon$  where  $\epsilon \in V$  and  $\epsilon$  nonzero.

Applying  $P$  to both sides,

$$\begin{aligned}
P^2\mathbf{u} &= P(\mathbf{u} + \epsilon) \\
P\mathbf{u} &= P\mathbf{u} + P\epsilon \\
P\epsilon &= \mathbf{0} \\
&= 0\epsilon
\end{aligned}$$

So 0 is an eigenvalue.

QUESTION 2. Find the eigenvalues and eigenvectors of  $A = \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix}$ .

*Solution:*

$$\det(A - \lambda I) = \det \begin{pmatrix} 1 - \lambda & 1 \\ -1 & 1 - \lambda \end{pmatrix} = (1 - \lambda)^2 + 1. \text{ So } \lambda = \frac{2 + \sqrt{-4}}{2} = 1 + i, 1 - i.$$

To find the eigenvectors for  $1 + i$ ,

$\ker \begin{pmatrix} -i & 1 \\ -1 & -i \end{pmatrix} = \ker \begin{pmatrix} 1 & i \\ -1 & -i \end{pmatrix} = \ker \begin{pmatrix} 1 & i \\ 0 & 0 \end{pmatrix}$ . So the eigenvector is  $t \begin{pmatrix} -i \\ 1 \end{pmatrix}$  for  $t$  nonzero.

To find the eigenvectors for  $1 - i$ ,

$\ker \begin{pmatrix} i & 1 \\ -1 & i \end{pmatrix} = \ker \begin{pmatrix} -1 & i \\ -1 & i \end{pmatrix} = \ker \begin{pmatrix} -1 & i \\ 0 & 0 \end{pmatrix}$ . So the eigenvector is  $t \begin{pmatrix} i \\ 1 \end{pmatrix}$  for  $t$  nonzero.

### QUESTION 3.

(a) Find all eigenvalues and eigenvectors of the matrix  $A = \begin{pmatrix} -4 & 5 \\ 1 & 0 \end{pmatrix}$

(b) Is  $A$  diagonalisable? Give reasons.

*Solution:*

(a)

$$\det(A - \lambda I) = \det \begin{pmatrix} -4 - \lambda & 5 \\ 1 & -\lambda \end{pmatrix} = (-4 - \lambda)(-\lambda) - (5)(1) = 0.$$

Therefore  $\lambda = -5, 1$  are the eigenvalues.

To find the eigenvectors for  $\lambda = -5$ ,

$$\ker \begin{pmatrix} 1 & 5 \\ 1 & 5 \end{pmatrix} = \ker \begin{pmatrix} 1 & 5 \\ 0 & 0 \end{pmatrix}. \text{ So the corresponding eigenvector is } t \begin{pmatrix} -5 \\ 1 \end{pmatrix}.$$

To find the eigenvectors for  $\lambda = 1$ ,

$$\ker \begin{pmatrix} -5 & 5 \\ 1 & -1 \end{pmatrix} = \ker \begin{pmatrix} -5 & 5 \\ 0 & 0 \end{pmatrix}. \text{ So the corresponding eigenvector is } t \begin{pmatrix} 1 \\ 1 \end{pmatrix}.$$

(b) Yes as there are 2 distinct eigenvalues.

QUESTION 4. Evaluate  $A^8$  if  $A = \begin{pmatrix} 5 & -8 \\ 1 & -1 \end{pmatrix}$

*Solution:*

We know  $A^8 = MD^8M^{-1}$ , where the columns of  $M$  are the eigenvectors of  $A$  and the diagonal elements of  $D$  are the eigenvalues of  $A$ .

Using the same method as in the previous questions, we can find the eigenvalues and corresponding eigenvectors to be  $\lambda = 3, 1$ , and  $t \begin{pmatrix} 4 \\ 1 \end{pmatrix}, t \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ .

$$\text{Therefore } M = \begin{pmatrix} 4 & 2 \\ 1 & 1 \end{pmatrix} \text{ and } D = \begin{pmatrix} 3 & 0 \\ 0 & 1 \end{pmatrix}.$$

$$\text{So, } A^8 = \begin{pmatrix} 4 & 2 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 3 & 0 \\ 0 & 1 \end{pmatrix} \times \frac{1}{2} \begin{pmatrix} 1 & -2 \\ -1 & 4 \end{pmatrix}$$

which we can then simplify.

QUESTION 5. Solve the following system of differential equations:

$$\begin{aligned}\frac{dx}{dt} &= x + 2y \\ \frac{dy}{dt} &= 3x + 2y.\end{aligned}$$

*Solution:*

We put the coefficients in a matrix,  $A = \begin{pmatrix} 1 & 2 \\ 3 & 2 \end{pmatrix}$ ,

then find the eigenvalues and corresponding eigenvectors to be 4, -1 and  $t \begin{pmatrix} 2 \\ 3 \end{pmatrix}, t \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ .

The general solution to  $\mathbf{y} = A\mathbf{y}$  is

$$\mathbf{y}(t) = \alpha_1 e^{4t} \begin{pmatrix} 2 \\ 3 \end{pmatrix} + \alpha_2 e^{-t} \begin{pmatrix} 1 \\ -1 \end{pmatrix} \text{ where } \alpha_1 \text{ and } \alpha_2 \in \mathbb{R}.$$

Therefore  $x(t) = 2\alpha_1 e^{4t} + \alpha_2 e^{-t}$  and  $y(t) = 3\alpha_1 e^{4t} - \alpha_2 e^{-t}$ .

QUESTION 6. Solve the following 2nd order ODE:  $y'' + 4y' - 5y = 0$ .

*Solution:*

We can turn a 2nd order linear ODE with constant coefficients into a system of 1st order equations with constant coefficients, then solve it using the method of the previous question.

Let  $y_1 = y$  and  $y_2 = y'_1 = y'$ .

Then  $y'_2 = y'' = 5y - 4y' = 5y_1 - 4y_2$ .

Therefore we get the following system of equations,

$$\begin{aligned}y'_1 &= y_2 \\ y'_2 &= 5y_1 - 4y_2.\end{aligned}$$

So  $\frac{d\mathbf{y}}{dt} = A\mathbf{y}$  where  $A = \begin{pmatrix} 0 & 1 \\ 5 & -4 \end{pmatrix}$ . We then find the eigenvalues and corresponding eigenvectors

of  $A$  to be -5, 1 and  $t \begin{pmatrix} -1 \\ 5 \end{pmatrix}$  and  $t \begin{pmatrix} 1 \\ 1 \end{pmatrix}$  with  $t$  nonzero.

Therefore  $\mathbf{y}(t) = \alpha_1 e^{-5t} \begin{pmatrix} -1 \\ 5 \end{pmatrix} + \alpha_2 e^t \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ .

Therefore  $y_1(t) = -\alpha_1 e^{-5t} + \alpha_2 e^t$ .

## Probability and Statistics

QUESTION 7. Show that the sequence defined by  $p_k = \frac{7}{10}(\frac{3}{10})^k$  for  $k = 0, 1, 2, \dots$  is a probability distribution.

*Solution:*

Since  $\sum_{k=0}^{\infty} \frac{7}{10}(\frac{3}{10})^k$  is a geometric series with ratio between -1 and 1, we can use the limiting sum formula to simplify it.

Therefore  $S_{\infty} = \frac{a}{1-r} = \frac{0.7}{1-0.3} = 1$ .

Also, since  $p_k \geq 0$ , it is a probability distribution.

QUESTION 8. A certain diagnostic test for a disease is 99% sure of correctly indicating that a person has the disease when they actually do and 98% sure of correctly indicating that a person does not have a disease when they actually do not. Suppose 2% of the population actually have this disease.

- (a) What is the probability that a person doesn't have the disease when they test positive (false positive)?
- (b) What is the probability that a person has the disease when they test negative (false negative)?

*Solution:*

(a) Using Bayes' Rule,

$$\begin{aligned}\mathbb{P}(D^c | T) &= \frac{\mathbb{P}(T | D^c)\mathbb{P}(D^c)}{\mathbb{P}(T | D^c)\mathbb{P}(D^c) + \mathbb{P}(T | D)\mathbb{P}(D)} \\ &= \frac{0.02 \times 0.98}{0.02 \times 0.98 + 0.99 \times 0.02} \\ &= 0.4974.\end{aligned}$$

(b) Again, using Bayes' Rule,

$$\begin{aligned}\mathbb{P}(D | T^c) &= \frac{\mathbb{P}(T^c | D)\mathbb{P}(D)}{\mathbb{P}(T^c | D)\mathbb{P}(D) + \mathbb{P}(T^c | D^c)\mathbb{P}(D^c)} \\ &= \frac{0.01 \times 0.02}{0.01 \times 0.02 + 0.98 \times 0.98} \\ &= 0.00021.\end{aligned}$$

QUESTION 10.

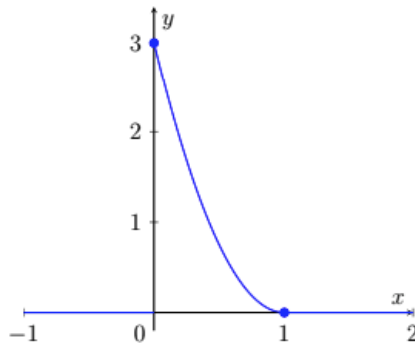
- (a) Sketch the graph of  $y = f(x)$ .



- (b) Find  $\mathbb{E}(X)$  and  $\text{Var}(X)$ .
- (c) Find  $\mathbb{P}(\frac{1}{2} < \sin(\pi X) < \frac{1}{\sqrt{2}})$ .
- (d) The median of a distribution is defined to be the real number  $m$  such that  $\mathbb{P}(X \leq m) = \frac{1}{2}$ . Find the median of the above distribution.

*Solution:*

(a) The graph is a parabola between 0 and 1, with vertex  $(1, 0)$  and y intercept 3, and it is 0 otherwise.



(b) To find the expectation, we integrate between 0 and 1 (since the PDF is 0 otherwise),

$$\begin{aligned}\mathbb{E}(X) &= \int_0^1 x(3(1-x)^2) dx \\ &= 3 \int_0^1 x - 2x^2 + x^3 dx \\ &= 3 \left( \frac{1}{2} - \frac{2}{3} + \frac{1}{4} \right) \\ &= \frac{1}{4}\end{aligned}$$

To find the variance, we need to also compute  $\mathbb{E}(X^2)$ ,

$$\begin{aligned}\mathbb{E}(X^2) &= \int_0^1 x^2(3(1-x)^2) dx \\ &= 3 \int_0^1 x^2 - 2x^3 + x^4 dx \\ &= 3 \left( \frac{1}{3} - \frac{2}{4} + \frac{1}{5} \right) \\ &= \frac{1}{10}.\end{aligned}$$

Therefore  $\text{Var}(X) = \mathbb{E}(X^2) - (\mathbb{E}(X))^2 = \frac{1}{10} - \frac{1}{4}^2 = \frac{3}{80}$ .

(c) Note that  $t \in [0, 1]$  since the PDF is 0 otherwise.

$\therefore \frac{1}{2} < \sin(\pi t) < \frac{1}{\sqrt{2}}$ , so  $\frac{1}{6} < t < \frac{1}{4}$  or  $\frac{3}{4} < t < \frac{5}{6}$ , which we get by drawing the sin graph and using a calculator.

Let  $F$  be the CDF of  $X$ . Then,

$$\begin{aligned}\mathbb{P}\left(\frac{1}{2} < \sin(\pi X) < \frac{1}{\sqrt{2}}\right) &= \mathbb{P}\left(\frac{1}{6} < X < \frac{1}{4}\right) + \mathbb{P}\left(\frac{3}{4} < X < \frac{5}{6}\right) \\ &= F\left(\frac{1}{4}\right) - F\left(\frac{1}{6}\right) + F\left(\frac{5}{6}\right) - F\left(\frac{3}{4}\right).\end{aligned}$$

To calculate  $F$ ,

$$\begin{aligned}F &= \int_0^x f(t) dt \\ &= \int_0^x 3(1-t)^2 dt \\ &= -(1-x)^3 - (-(1-0)^3) \\ &= 1 - (1-x)^3.\end{aligned}$$

Therefore,

$$\begin{aligned}\mathbb{P}\left(\frac{1}{2} < \sin(\pi X) < \frac{1}{\sqrt{2}}\right) &= \left(1 - \frac{1}{6}\right)^3 - \left(1 - \frac{1}{4}\right)^3 + \left(1 - \frac{3}{4}\right)^3 - \left(1 - \frac{5}{6}\right)^3 \\ &= \frac{145}{864}.\end{aligned}$$

(d) We already have  $F$  from part c, and we need to solve  $F(m) = \frac{1}{2}$ .

$\therefore 1 - (1-m)^3 = \frac{1}{2}$ , so we can solve for  $m$  to get  $m = 1 - \frac{1}{3\sqrt{2}}$ .

**QUESTION 11.** A 6-sided die, with faces numbered 1 to 6, is suspected of being unfair so that the number 6 will occur more frequently than should happen by chance. During 300 test rolls of the die, the number 6 occurred 68 times.

- Write down an expression for a tail probability that measures the chance of rolling a 6 at least 68 times.
- Use the normal approximation to the binomial to estimate this probability.
- Is this evidence that the die is unfair?

*Solution:*

(a)  $X \sim B(300, \frac{1}{6})$ . Therefore,

$$\begin{aligned}\mathbb{P}(X \geq 68) &= \sum_{k=68}^{300} \mathbb{P}(X = k) \\ &= \sum_{k=68}^{300} \binom{300}{k} \left(\frac{1}{6}\right)^k \left(\frac{5}{6}\right)^{300-k}\end{aligned}$$

(b)  $\mathbb{E}(X) = np = 300 \times \frac{1}{6} = 50$ .

$\text{Var}(X) = np(1-p) = 50 \times \frac{5}{6} = \frac{125}{3}$ . Therefore  $\text{SD}(X) = \sqrt{\text{Var}(X)} = \sqrt{\frac{125}{3}}$ .

$$P(X \geq 68) = P(Y \geq 67.5) = P\left(Z \geq \frac{67.5 - 50}{\sqrt{\frac{125}{3}}}\right) = \mathbb{P}(z \geq 2.71) = 1 - \mathbb{P}(z \leq 2.71) = 0.0034.$$

(c) Since  $0.34\% < 5\%$ , the answer is yes.

