

## Homework 3 Partial Solutions

### Section 3.1

8. This questions is about arbitrary vectors, these could be vectors in  $\mathbb{R}^n$  but it could also be the space of matrices  $\mathbb{R}^{n \times m}$ , could be the space of continuous functions on the unit interval into  $\mathbb{R}$ ,  $C([0, 1], \mathbb{R})$ , etc. So you must argue generally using axioms of vector spaces.

$$x + y = x + z$$

$$(-x) + (x + y) = (-x) + (x + z) \quad (\text{A4})$$

$$(-x + x) + y = (-x + x) + z \quad (\text{A2})$$

$$0 + y = 0 + z \quad (\text{A4})$$

$$y = z \quad (\text{A3})$$

13. There are various ways to see that this is not a vector space. One way is to notice that there is no 0 element!

What element  $a$  of  $\mathbb{R}$  would satisfy  $\max(a, r) = r$  for all  $r \in \mathbb{R}$ ? For  $r \geq 0$ ,  $a = 0$  would suffice, but what would work for  $r < 0$ ? If  $a \oplus r = r$  for  $r < 0$ , then  $a < r$ . But then  $a < r$  for all  $r \in \mathbb{R}$ !

14. Let  $V = \mathbb{Z}$  and define scalar multiplication by

$$\alpha \cdot_V n = \lfloor \alpha \rfloor \cdot n \quad (1)$$

$$n +_V m = n + m \quad (2)$$

Is this a vector space?

All the additive axioms clearly hold since these are true of integer arithmetic.

The problem here is  $\alpha \cdot_V (\beta \cdot_V n) = (\alpha \cdot \beta) \cdot_V n$ . For example:

$$.5 \cdot_V (2 \cdot_V n) = 0 \cdot (2 \cdot n) = 0$$

while

$$(.5 \cdot 2) \cdot_V n = 1 \cdot_V n = 1 \cdot n = n$$

## Section 3.2

2.

(a) This is not a subspace because  $(0, 0)^T \notin S$ .

(b) This is a subspace.

- If  $(a, b, c) \in S$ , then  $\alpha(a, b, c)^T \in S$ , since,  $a = b = c$  implies  $\alpha a = \alpha b = \alpha c$ .
- If  $(a, b, c)^T, (A, B, C)^T \in S$ , then  $a + A = b + B = c + C$ , so  $(a, b, c)^T + (A, B, C)^T \in S$ .

Thus  $S$  is closed under scalar multiplication and addition and is a subspace.

(c) This is a subspace. Do just like (b), but use the property  $x_1 = x_2 + x_3$ . Another way is to notice that  $S = NS(A)$  where  $A = \begin{bmatrix} 1 & -1 & -1 \end{bmatrix}$ . (We could have done this with (b) as well.)

(d) This is not a subspace  $(1, 2, 1)^T$  and  $(4, 1, 1)^T$  are in  $S$ , but the sum  $(5, 3, 2)^T \notin S$

4.

(a)  $\text{rref}(A) = I_2$  so  $NS(A) = \text{span}\{\mathbf{0}\}$ .

(b)  $\text{rref}(A) = \begin{bmatrix} 1 & 2 & -3 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$  so  $A\mathbf{x} = \mathbf{0}$  is equivalent to

$$\begin{aligned} x_1 + 2x_2 - 3x_3 &= 0 \\ x_4 &= 0 \end{aligned}$$

Let  $x_2 = s$  and  $x_3 = t$ , then we have:

$$\begin{aligned} x_1 &= -2s + 3t \\ x_2 &= s \\ x_3 &= t \\ x_4 &= 0 \end{aligned}$$

which is the same as

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = s \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} 3 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

So  $NS(A) = \text{span}\{(-1, 1, 0, 0)^T, (3, 0, 1, 0)^T\}$ .

(c)  $\text{rref}(A) = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{bmatrix}$  so this has  $x_3$  as a free variable. Let  $x_3 = t$ , then

$$\begin{aligned} x_1 &= t \\ x_2 &= t \end{aligned}$$

is the resulting system so an element of  $\text{NS}(A)$  is of the form

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} t \\ t \\ t \end{bmatrix} = t \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

so  $\text{NS}(A) = \text{span}\{(1, 1, 1)^T\}$ .

(d) Just as an example of using MATLAB

```
1 A=[1 1 -1 2; 2 2 -3 1; -1 -1 0 -5]
2 rref(A)
```

$\text{rref}(A) = \begin{bmatrix} 1 & 1 & 0 & 5 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 0 \end{bmatrix}$  so  $x_2$  and  $x_4$  are the non-pivot, hence free variables. Let  $x_2 = s$  and  $x_4 = t$ , then the system becomes

$$\begin{aligned} x_1 &= -s - 5t \\ x_3 &= -3t \end{aligned}$$

So we have  $\mathbf{x} \in \text{NS}(A)$  iff

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -s - 5t \\ s \\ -3t \\ t \end{bmatrix} = s \begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} -5 \\ 0 \\ -3 \\ 1 \end{bmatrix}$$

and thus

$$\text{NS}(A) = \text{span} \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -5 \\ 0 \\ -3 \\ 1 \end{bmatrix} \right\}$$

8.  $A$  is fixed.

- $\mathbf{0}A = A\mathbf{0}$  so  $\mathbf{0} \in S$
- Let  $B, C \in S$ , then  $BA = AB$  and  $CA = AC$  so  $(B + C)A = BA + CA = AB + AC = A(B + C)$  and hence  $B + C \in S$ .
- Let  $B \in S$ , then  $(\alpha B)A = \alpha(BA) = \alpha(AB) = A(\alpha B)$ , so  $\alpha B \in S$ .

11. Just put the vectors in as columns, or rows, of a matrix  $A$ . Find  $\text{rref}(A)$ . If there are two non-zero rows, that is  $\text{rank}(A) = 2$ , then the set is a basis. for example, given  $B = \{(2, 1)^T, (3, 2)^T\}$  for  $A = \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix}$  (I put the vectors in as columns).  $\text{rref}(A) = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ , so  $B$  spans  $\mathbb{R}^2$ . (You could just compute  $\text{rank}(A)$  in MATLAB.

13. If  $A = [\mathbf{x}_1 \quad \mathbf{x}_2]$ , then  $\mathbf{x} \in \text{span}\{\mathbf{x}_1, \mathbf{x}_2\}$  iff  $A\mathbf{z} = \mathbf{x}$  has a solution, similar for  $\mathbf{y}$ . So for  $\mathbf{x}$  just try to solve

$$\begin{bmatrix} -1 & 3 \\ 2 & 4 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 6 \\ 6 \end{bmatrix}$$

Since

$$\text{rref} \left( \begin{bmatrix} -1 & 3 & 2 \\ 2 & 4 & 6 \\ 3 & 2 & 6 \end{bmatrix} \right) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

This has no solution. Recall this was an augmented matrix, and the last row means  $0z_1 + 0z_2 = 1$ , which is nonsense.

**17.** Here is what this question is getting at. Suppose you take  $\mathbf{b} \in \text{CS}(A) = \text{Img}(A)$  so  $\mathbf{b} = A\mathbf{x}$  for some  $\mathbf{x}$ . Then if  $A\mathbf{x}' = \mathbf{b}$  also, we see that  $A\mathbf{x} - A\mathbf{x}' = A(\mathbf{x} - \mathbf{x}') = \mathbf{0}$  so  $\mathbf{x} - \mathbf{x}' \in \text{NS}(A)$ .

It is also clear that if  $\mathbf{z} \in \text{NS}(A)$  and  $A\mathbf{x} = \mathbf{b}$ , then  $A(\mathbf{x} + \mathbf{z}) = A\mathbf{x} + A\mathbf{z} = \mathbf{b} + \mathbf{0} = \mathbf{b}$ .

From these two facts we see that if  $\mathbf{x}$  is **any** solution to the system  $A\mathbf{x} = \mathbf{b}$ , then the set of **all** solutions is

$$\mathbf{x} + \text{NS}(A) = \{\mathbf{x} + \mathbf{z} \mid A\mathbf{z} = \mathbf{0}\}$$

**18.**

(a) Adding a vector to a spanning set leaves it a spanning set. This is clear since if  $S \subset S' \subset V$  are sets of vectors in a vector space  $V$ , then clearly  $\text{span}(S) \subset \text{span}(S')$ . But if  $\text{span}(S) = V$ , i.e.,  $S$  is a spanning set, then  $V \subset \text{span}(S) \subset \text{span}(S') \subset V$  so these must all be the same.

(b) Removing a vector from a spanning set may, or may not, leave it as a spanning set. If it is a minimal spanning set (a basis), then removing a vector will mean that what is left is no longer spanning.

## Section 3.3

**2.** Again just write these vectors down as the rows of a matrix  $A$ . If  $\text{rref}(A)$  has any 0 rows, then the vectors are not independent, otherwise they are. For example:

$$\text{rref} \left( \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 2 & 3 \end{bmatrix} \right) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

So these vectors are not independent.

**5.** (This is sort of the opposite of the spanning case.)

(a) Adding vectors to a linearly independent set can obviously mess up independence. (Just add a linear combination of the original vectors.) For example, if  $S \subset \mathbb{R}^n$  is linearly independent, then  $S \cup \{\mathbf{0}\}$  is not.

(b) Clearly removing a vector from a linearly independent set cannot mess up linear independence.

Specifically if  $S = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  and  $S' \subset S$ , say  $S' = \{\mathbf{v}_{i_1}, \dots, \mathbf{v}_{i_k}\}$  and  $c_{i_1}\mathbf{v}_{i_1} + \dots + c_{i_k}\mathbf{v}_{i_k} = \mathbf{0}$  is a linear combination of elements of  $S'$ , then this is trivially also a linear combination of elements of  $S$  and hence by the independence of  $S$  we have  $c_{i_1} = \dots = c_{i_k} = 0$ . So  $S'$  is linearly independent.

8. Determine whether the following are independent in  $P_3$ .

(a)  $\{1, x^2, x^2 - 2\}$  is not independent as  $x^2 - 2 = -2 \cdot 1 + 1 \cdot x^2$ , so  $x^2 - 2$  is a linear combination of 1 and  $x^2$ .

(c)  $\{x + 2, x + 1, x^2 - 1\}$  relative to the standard (ordered) basis for  $P_3$ ,  $\{1, x, x^2\}$ , this is equivalent to asking if  $\{(2, 1, 0), (1, 1, 0), (-1, 0, 1)\}$  is linearly independent. Clearly,

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

so  $\{x + 2, x + 1, x^2 - 1\}$  is linearly independent.

(d)  $\{x + 2, x^2 - 1\}$  is independent since  $\{x + 2, x + 1, x^2 - 1\}$  is linearly independent, by (c).

9. Show the following sets are linearly independent in  $C([0, 1])$

(a)  $\sin(\pi x)$  and  $\cos(\pi x)$

One interesting way here is to note that  $\langle f, g \rangle = \int_0^1 f \cdot g \, dx$  is an inner-product on  $C([0, 1])$  and  $\langle \sin(\pi x), \cos(\pi x) \rangle = 0$ , so actually, these two functions are orthogonal!

A less interesting way is to note that if  $a \sin(\pi x) + b \cos(\pi x) = 0$  (the 0 function), then letting  $x = 0$  gives  $a \sin(0) + b \cos(0) = b = 0$  and letting  $x = 1/2$  gives  $a \sin(\pi/2) + b \cos(\pi/2) = a = 0$  so  $a = b = 0$  and hence the two functions are independent.

(b)  $x^{3/2}$  and  $x^{5/2}$

Suppose  $ax^{3/2} + bx^{5/2} = 0$  for all  $x \in [0, 1]$ , then for  $x = 1$  we have  $a + b = 0$  and for  $x = 1/4$  we have  $a(1/2)^3 + b(1/2)^5 = 0$  so  $a + b(1/2)^2 = 0$  hence  $a + b/4 = 0$  or equivalently  $4a + b = 0$ . Solving

$$\begin{aligned} 4a + b &= 0 \\ a + b &= 0 \end{aligned}$$

gives  $a = b = 0$ . So These are independent.

(c)  $1, x^x - e^{-x}$  and  $e^x + e^{-x}$

Again suppose  $h(x) = a + b(e^x - e^{-x}) + c(e^x + e^{-x}) = 0$ . It is easy to see  $h(0) = a + 2c = 0$ ,  $h'(0) = 2b = 0$  and  $h''(0) = 2c = 0$ . So clearly,  $a = b = c = 0$  as desired.

(d)  $e^x, e^{-x}$  and  $e^{2x}$

This is like (c), Assume  $h(x) = ae^x + be^{-x} + ce^{2x}$ , then  $h'(x) = ae^x - be^{-x} + 2ce^{2x}$  and  $h''(x) = ae^x + be^{-x} + 4e^{2x}$  and so

$$\begin{aligned} h(0) &= a + b + c = 0 \\ h'(0) &= a - b + 2c = 0 \\ h''(0) &= a + b + 4c = 0 \end{aligned}$$

It is easy to check that this has the unique solution  $a = b = c = 0$ .

**10.** It turns out here that  $1$ ,  $\cos(x)$ , and  $\sin^2(x/2)$  are linearly dependent and this is from one of the half-angle formulas,

$$\cos(x) = \cos^2(x/2) - \sin^2(x/2) = 1 - 2\sin^2(x/2)$$

.

**16.** Show that the columns of  $A$  are linearly independent iff  $\text{NS}(A) = \{\mathbf{0}\}$ .

Suppose  $A$  is  $m \times n$  so  $A = [\mathbf{a}_1 \ \cdots \ \mathbf{a}_n]$  with  $\mathbf{a}_i \in \mathbb{R}^m$  the  $i^{\text{th}}$  column of  $A$ . Then

$$A\mathbf{x} = x_1\mathbf{a}_1 + \cdots + x_n\mathbf{a}_n$$

is an arbitrary linear combination of the columns of  $A$  and so.

(if) Assume  $\text{NS}(A) = \{\mathbf{0}\}$ , then  $x_1\mathbf{a}_1 + \cdots + x_n\mathbf{a}_n = \mathbf{0}$  iff  $A\mathbf{x} = \mathbf{0}$  iff  $\mathbf{x} = \mathbf{0}$ , that is  $x_1 = x_2 = \cdots = x_n = 0$ . So the columns of  $A$  are linearly independent since the only linear combination giving  $\mathbf{0}$  is the trivial combination.

(only-if) Assume the columns of  $A$  are linearly independent, then  $A\mathbf{x} = \mathbf{0}$  would mean the  $x_1\mathbf{a}_1 + \cdots + x_n\mathbf{a}_n = \mathbf{0}$  so by linear independence,  $x_1 = x_2 = \cdots = 0$  and hence  $A\mathbf{x} = \mathbf{0}$  implies  $\mathbf{x} = \mathbf{0}$  so  $\text{NS}(A) = \{\mathbf{0}\}$ .

**17.** Suppose  $\text{NS}(A) = \{\mathbf{0}\}$  and  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k$  are linearly independent. Suppose also

$$\alpha_1 A\mathbf{x}_1 + \alpha_2 A\mathbf{x}_2 + \cdots + \alpha_k A\mathbf{x}_k = \mathbf{0},$$

then

$$\mathbf{0} = \alpha_1 A\mathbf{x}_1 + \cdots + \alpha_k A\mathbf{x}_k = A(\alpha_1 \mathbf{x}_1 + \cdots + \alpha_k \mathbf{x}_k)$$

so  $\alpha_1 \mathbf{x}_1 + \cdots + \alpha_k \mathbf{x}_k \in \text{NS}(A) = \{\mathbf{0}\}$  and thus

$$\alpha_1 \mathbf{x}_1 + \cdots + \alpha_k \mathbf{x}_k = \mathbf{0}$$

But the  $\mathbf{x}_i$ 's are linearly independent so  $\alpha_1 = \alpha_2 = \cdots = \alpha_k = 0$ . but this is what we needed to see that  $A\mathbf{x}_1, A\mathbf{x}_2, \dots, A\mathbf{x}_k$  is linearly independent.

## Section 3.4

5.

(a) Let  $A$  be the matrix whose columns are the three vectors given

$$A = \begin{bmatrix} 2 & 3 & 2 \\ 1 & -1 & 6 \\ 3 & 4 & 4 \end{bmatrix}$$

The given vectors are linearly independent iff  $\text{NS}(A) = \{\mathbf{0}\}$ , since

$$\text{NS}(A) = \{\mathbf{0}\} \text{ iff } A\mathbf{x} = \mathbf{0} \text{ implies } \mathbf{x} = \mathbf{0},$$

but the right hand side here says precisely that the only linear combination of the columns that yields  $\mathbf{0}$  is the trivial combination, that is all coefficients are 0.

$$\text{rref } A = \begin{bmatrix} 1 & 0 & 4 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{bmatrix}$$

Clearly, this has a nontrivial null space, in fact,

$$\text{NS}(A) = \text{span}\{(-4, 2, 1)\}$$

So  $-4\mathbf{x}_1 + 2\mathbf{x}_2 + \mathbf{x}_3 = \mathbf{0}$ , where these were the given vectors. (Easy for the reader to check. Do it!)

(b) Clearly  $\mathbf{x}_1$  and  $\mathbf{x}_2$  are linearly independent, since there is no  $r \in \mathbb{R}$  such that  $r\mathbf{x}_1 = \mathbf{x}_2$ .

(c) Let  $S = \text{span}\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$ , then (a) and (b) together show  $2 \leq \dim(S) < 3$  so  $\dim(S) = 2$ .

(d) A 2-dimensional subspace of  $\mathbb{R}^3$  is a plane.

**alternate solution**

$$\begin{bmatrix} 3 & -3 & -6 \\ -2 & 2 & 4 \\ 3 & 4 & 4 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & -2 \\ 0 & 0 & 0 \\ 0 & 7 & 10 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & -2 \\ 0 & 7 & 10 \\ 0 & 0 & 0 \end{bmatrix}$$

So a basis for  $V = \text{span}\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$  is given by  $\{\mathbf{x}_2, \mathbf{x}_3\}$ . So  $\dim(V) = 2$  and  $V$  is a plane in  $\mathbb{R}^3$ .

7.  $(a + b, a - b + 2c, b, c) = a(1, 1, 0, 0) + b(1, -1, 1, 0) + c(0, 2, 0, 1)$

It is easy to see that  $\{(1, 1, 0, 0), (1, -1, 1, 0), (0, 2, 0, 1)\}$  is independent so  $\dim(S) = 3$ .

8.

(a) No, two non co-linear vectors span a plane not all of  $\mathbb{R}^3$

(b)  $X$  must be linearly independent. We can be more specific here. If  $A$  has columns  $\mathbf{x}_1 = (1, 1, 1)$ ,  $\mathbf{x}_2 = (3, -1, 4)$ , and  $\mathbf{x}_3 = (a_1, a_2, a_3)$ , then  $X$  is linearly independent iff any of the following hold

- $\text{NS}(A) = \{\mathbf{0}\}$
- $\det(A) = 0$
- $\text{rref}(A) = I_3$

Any one of these can be used to characterize the  $x_3$  that are allowed, but geometrically we know that the set of these vectors is ALL vectors not in the plane spanned by  $\mathbf{x}_1$  and  $\mathbf{x}_2$ .

(c) Any vector not in the plane spanned by  $\mathbf{x}, \mathbf{x}_2$  will work, say  $\mathbf{x}_3 = (1, 0, 0)^T$

13.  $\cos(2x) = 2\cos^2(x) - 1$ , so  $\dim(\text{span}\{\cos(2x), \cos^2(x), 1\}) = 2$ .

## Section 3.5

1. Find the transition matrix from the basis  $\mathcal{U} = \{\mathbf{u}_1, \mathbf{u}_2\}$  to the standard basis. This I would also denote  $[\text{id}]_{\mathcal{U}, \mathcal{E}}$ , where  $\text{id} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  is just the identity transformation.

(a)  $U = [\mathbf{u}_1 \quad \mathbf{u}_2] = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$

(b)  $U = [\mathbf{u}_1 \quad \mathbf{u}_2] = \begin{bmatrix} 1 & 2 \\ 2 & 5 \end{bmatrix}$

(c)  $U = [\mathbf{u}_1 \quad \mathbf{u}_2] = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

2. This is just the opposite of (1), find the transition matrix from the standard basis to the basis  $\mathcal{U} = \{\mathbf{u}_1, \mathbf{u}_2\}$ , that is find  $[\text{id}]_{\mathcal{E}, \mathcal{B}}$ .

Letting  $U$  be the matrix from (1), here the matrix we desire is  $U^{-1}$ , so

(a)  $U^{-1} = [\mathbf{u}_1 \quad \mathbf{u}_2]^{-1} = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}^{-1} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$

(b)  $U^{-1} = [\mathbf{u}_1 \quad \mathbf{u}_2]^{-1} = \begin{bmatrix} 1 & 2 \\ 2 & 5 \end{bmatrix}^{-1} = \begin{bmatrix} 5 & -2 \\ -2 & 1 \end{bmatrix}$

(c)  $U^{-1} = [\mathbf{u}_1 \quad \mathbf{u}_2]^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

3.

(a) The transition matrix for  $\mathcal{V} = \{\mathbf{v}_1, \mathbf{v}_2\} \rightarrow \{\mathbf{e}_1, \mathbf{e}_2\}$  is  $V = \begin{bmatrix} 3 & 4 \\ 2 & 3 \end{bmatrix}$ . So the transformation matrix from  $\mathcal{V} = \{\mathbf{v}_1, \mathbf{v}_2\} \rightarrow \mathcal{U} = \{\mathbf{u}_1, \mathbf{u}_2\}$  is  $U^{-1}V$ , where  $U$  is as in 1.

(a)  $U^{-1}V = \begin{bmatrix} 2.5 & 3.5 \\ -0.5 & -0.5 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 5 & 7 \\ -1 & -1 \end{bmatrix}$

(b)  $U^{-1}V = \begin{bmatrix} 11 & 14 \\ -4 & -5 \end{bmatrix}$

(c)  $U^{-1}V = \begin{bmatrix} 2 & 3 \\ 3 & 4 \end{bmatrix}$

6. Let  $\mathcal{U} = \{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\} = \{(1, 1, 1), (1, 2, 2), (1, 3, 4)\}$  and  $\mathcal{V} = \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} = \{(4, 6, 7), (0, 1, 1), (0, 1, 2)\}$ .

(a) Find transition matrix from  $\mathcal{V}$  to  $\mathcal{U}$ .

This is

$$[\text{id}]_{\mathcal{V}, \mathcal{U}} = [\text{id} \circ \text{id}]_{\mathcal{V}, \mathcal{U}} = [\text{id}]_{\mathcal{E}, \mathcal{U}} [\text{id}]_{\mathcal{V}, \mathcal{E}} = U^{-1}V = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 3 & 4 \end{bmatrix}^{-1} \begin{bmatrix} 4 & 0 & 0 \\ 6 & 1 & 1 \\ 7 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & -1 & -2 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$$

(b) Find the  $\mathcal{U}$  representation of  $\mathbf{v} = 2\mathbf{v}_1 + 3\mathbf{v}_2 - 4\mathbf{v}_3$ .



We see  $[\mathbf{v}]_{\mathcal{V}} = \begin{bmatrix} 2 \\ 3 \\ -4 \end{bmatrix}$  and

$$[\mathbf{v}]_{\mathcal{U}} = [\text{id}]_{\mathcal{V}, \mathcal{U}} [\mathbf{v}]_{\mathcal{V}} = \begin{bmatrix} 1 & -1 & -2 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \\ -4 \end{bmatrix} = \begin{bmatrix} 7 \\ 5 \\ -2 \end{bmatrix}$$

So  $\mathbf{v} = 7\mathbf{u}_2 + 5\mathbf{u}_3 - 2\mathbf{u}_1$ .

You should check this:

$$\begin{aligned} 2\mathbf{v}_1 + 3\mathbf{v}_2 - 4\mathbf{v}_3 &= \begin{bmatrix} 4 & 0 & 0 \\ 6 & 1 & 1 \\ 7 & 1 & 2 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \\ -4 \end{bmatrix} = \begin{bmatrix} 8 \\ 11 \\ 9 \end{bmatrix} \\ 7\mathbf{u}_2 + 5\mathbf{u}_3 - 2\mathbf{u}_1 &= \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 3 & 4 \end{bmatrix} \begin{bmatrix} 7 \\ 5 \\ -2 \end{bmatrix} = \begin{bmatrix} 8 \\ 11 \\ 9 \end{bmatrix} \end{aligned}$$

**10.** Find transition matrix from the basis  $\mathcal{B} = \{1, x, x^2\}$  for  $\mathbb{P}_3$  to  $\mathcal{C} = \{1, 1+x, 1+x+x^2\}$ . The transformation matrix from  $\mathcal{C}$  to  $\mathcal{B}$  is easy:

$$[1]_{\mathcal{C}} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \quad [1+x]_{\mathcal{C}} = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad [1+x+x^2]_{\mathcal{C}} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

So we have

$$[\text{id}]_{\mathcal{C}, \mathcal{B}} = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

The matrix in the other direction, from  $\mathcal{B}$  to  $\mathcal{C}$  is just the inverse

$$[\text{id}]_{\mathcal{B}, \mathcal{C}} = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix}$$

As an application, write  $p = 3 - 2x + 4x^2$  in the  $\mathcal{C}$  basis.  $[p]_{\mathcal{B}} = (3, -2, 4)$  so

$$[p]_{\mathcal{C}} = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ -2 \\ 4 \end{bmatrix} = \begin{bmatrix} 5 \\ -6 \\ 4 \end{bmatrix}$$

hence

$$3 - 2x + 4x^2 = 5 - 6(1+x) + 4(1+x+x^2)$$

## Section 3.6

1. Let  $A$  denote the matrix given

$$(a) \text{ rref}(A) = \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \text{ so}$$

So letting  $x_3 = t$  we get  $x_2 = 0$  and  $x_1 = -2t$  and the solutions  $A\mathbf{x} = \mathbf{0}$  are all those

$$\mathbf{x} = t \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$$

$$\text{RS}(A) = \text{span}\{(1, 0, 2)^T, (0, 1, 0)^T\}$$

$$\text{CS}(A) = \text{span}\{(1, 2, 4)^T, (3, 1, 7)^T\}$$

$$\text{NS}(A) = \text{span}\{(-2, 0, 1)^T\}$$

Remarks: The non zero rows of  $\text{rref}(A)$  are a basis for  $\text{RS}(A)$ . The columns of  $A$  that are pivot columns of  $\text{rref}(A)$  are a basis for  $\text{CS}(A)$ .

$$(b) \text{ rref}(A) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & -2/7 \\ 0 & 0 & 1 & 10/7 \end{bmatrix}$$

So setting  $x_4 = t$  we get  $A\mathbf{x} = \mathbf{0}$  at  $\mathbf{x} = t \begin{bmatrix} 0 \\ 2/7 \\ -10/7 \\ 1 \end{bmatrix}$  and so

$$\text{RS}(A) = \text{span}\{(1, 0, 0, 0)^T, (0, 1, 0, -2/7)^T, (0, 0, 1, 10/7)^T\}$$

$$\text{CS}(A) = \text{span}\{(-3, 1, 3)^T, (1, 2, 4)^T, (3, -1, 5)^T\}$$

$$\text{NS}(A) = \text{span}\{(0, 2/7, -10/7, 1)^T\}$$

6. If  $\mathbf{b}$  is in  $\text{CS}(A)$  and the columns of  $A$  are independent, then  $A\mathbf{x} = \mathbf{b}$  has a solution, since this is what  $\mathbf{b} \in \text{CS}(A)$  means and the solution is unique. For is  $A\mathbf{x}' = A\mathbf{x} = \mathbf{b}$ , then  $A\mathbf{x}' - A\mathbf{x} = A(\mathbf{x}' - \mathbf{x}) = \mathbf{0}$ , but this means  $\mathbf{x}' - \mathbf{x} = \mathbf{0}$ , since the columns of  $A$  are independent and hence  $\mathbf{x} = \mathbf{x}'$ .

9.

(a) If  $A$  is  $6 \times 5$  and  $\dim(\text{NS}(A)) = 2$ , then since  $\mathbb{R}^5 = \text{RS}(A) \oplus \text{NS}(A)$  we have  $5 = \dim \text{RS}(A) + 2$  so  $\dim \text{RS}(A) = 3$ .

(b) If  $B$  is  $6 \times 5$ , then as above  $5 = \dim \text{NS}(A) + \dim \text{RS}(A) = \dim \text{NS}(A) + \text{rank}(A) = \dim \text{NS}(A) + 4$ , so  $\dim \text{NS}(A) = 1$ .

14. From  $U$  read off the solutions to  $A\mathbf{x} = 0$ , i.e.  $\text{NS}(A) = \text{NS}(U)$  as

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -2s - t \\ -s - 4t \\ s \\ t \end{bmatrix} = s \begin{bmatrix} -2 \\ -1 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -1 \\ -4 \\ 0 \\ 1 \end{bmatrix} = s \cdot \mathbf{u}_1 + t \cdot \mathbf{u}_2$$

Now we know  $A(s\mathbf{u}_1 + t\mathbf{u}_2) = \mathbf{0}$  so in particular,  $A\mathbf{u}_1 = A\mathbf{u}_2 = \mathbf{0}$  and if  $A = \begin{bmatrix} a_1 & a_2 & a_3 & a_4 \end{bmatrix}$ , then

$$A\mathbf{u}_1 = -2\mathbf{a}_1 - \mathbf{a}_2 + \mathbf{a}_3 = \mathbf{0}$$

$$A\mathbf{u}_2 = -\mathbf{a}_1 - 4\mathbf{a}_2 + \mathbf{a}_4 = \mathbf{0}$$

so

$$\mathbf{a}_3 = 2\mathbf{a}_1 + \mathbf{a}_2$$

$$\mathbf{a}_4 = \mathbf{a}_1 + 4\mathbf{a}_2$$