

# Math 140C: Homework 1

## Merrick Qiu

### Problem 1

1. (  $\implies$  ) Since  $\mathbf{y} \in \text{span}(E)$ , we can write  $\mathbf{y} = c_1\mathbf{v}_1 + \cdots + c_r\mathbf{v}_r$  for some  $c_i \in \mathbb{R}$ . Thus we have that

$$c_1\mathbf{v}_1 + \cdots + c_r\mathbf{v}_r - \mathbf{y} = 0$$

which implies that  $E \cup \{\mathbf{y}\}$  is linearly dependent.

(  $\impliedby$  ) Since  $E \cup \{\mathbf{y}\}$  is linearly dependent, we can write  $c_1\mathbf{v}_1 + \cdots + c_r\mathbf{v}_r + c_{r+1}\mathbf{y} = 0$  for some  $c_i \in \mathbb{R}$ . Thus

$$\mathbf{y} = -\frac{c_1}{c_{r+1}}\mathbf{v}_1 - \cdots - \frac{c_r}{c_{r+1}}\mathbf{v}_r$$

which implies that  $\mathbf{y} \in \text{span}(E)$ .

2. If  $\mathbf{x} \in \text{span}(E)$  then  $\mathbf{x} = a_1\mathbf{v}_1 + \cdots + a_r\mathbf{v}_r$  for some  $a_i \in \mathbb{R}$ . It is also true that  $\mathbf{x} = a_1\mathbf{v}_1 + \cdots + a_r\mathbf{v}_r + 0\mathbf{y}$  so  $\mathbf{x} \in \text{span}(E \cup \{\mathbf{y}\})$ .

If  $\mathbf{x} \in \text{span}(E \cup \{\mathbf{y}\})$  then  $\mathbf{x} = a_1\mathbf{v}_1 + \cdots + a_r\mathbf{v}_r + a_{r+1}\mathbf{y}$  for some  $a_i \in \mathbb{R}$ . Since  $E \cup \{\mathbf{y}\}$  is linearly dependent,  $\mathbf{y} \in \text{span}(E)$  so  $\mathbf{y} = c_1\mathbf{v}_1 + \cdots + c_r\mathbf{v}_r$  for some  $c_i \in \mathbb{R}$ . Thus

$$\mathbf{x} = (a_1 + a_{r+1}c_1)\mathbf{v}_1 + \cdots + (a_r + a_{r+1}c_r)\mathbf{v}_r$$

so  $\mathbf{x} \in \text{span}(E)$

## Rudin 9.1

If  $\mathbf{x} \in \text{span}(S)$  and  $\mathbf{y} \in \text{span } S$  then for some set of  $\mathbf{v}_1, \dots, \mathbf{v}_n \in S$  and constants  $a_i, b_i \in \mathbb{R}$ , we can write

$$\mathbf{x} = \sum_{i=1}^n a_i \mathbf{v}_i \quad \mathbf{y} = \sum_{i=1}^n b_i \mathbf{v}_i$$

$\text{span}(S)$  is a vector space since for all  $c \in \mathbb{R}$  and  $\mathbf{x}, \mathbf{y} \in \text{span}(S)$ ,

$$c\mathbf{x} = \sum_{i=1}^n ca_i \mathbf{v}_i \in \text{span}(S)$$

$$\mathbf{x} + \mathbf{y} = \sum_{i=1}^n (a_i + b_i) \mathbf{v}_i \in \text{span}(S).$$

## Rudin 9.2

If  $A$  and  $B$  are linear transformations in  $X$  then for all  $\mathbf{x}, \mathbf{v}_1, \mathbf{v}_2 \in X$

$$\begin{aligned} BA(\mathbf{v}_1 + \mathbf{v}_2) &= B(A(\mathbf{v}_1 + \mathbf{v}_2)) \\ &= B(A\mathbf{v}_1 + A\mathbf{v}_2) \\ &= BA\mathbf{v}_1 + BA\mathbf{v}_2 \end{aligned}$$

$$BA(c\mathbf{x}) = B(cA\mathbf{x}) = cBA\mathbf{x}$$

Thus  $BA$  is also a linear transformation.

If  $A$  is one-to-one from  $X$  onto  $X$  then for all  $\mathbf{x}, \mathbf{v}_1, \mathbf{v}_2 \in X$  we can write

$$\mathbf{x} = A\mathbf{y} \quad \mathbf{v}_1 = A\mathbf{v}_1 \quad \mathbf{v}_2 = A\mathbf{v}_2$$

for some vectors  $\mathbf{y}, \mathbf{v}_1, \mathbf{v}_2 \in X$ .  $A^{-1}$  is a linear operator since

$$\begin{aligned} A^{-1}(\mathbf{v}_1 + \mathbf{v}_2) &= A^{-1}(A\mathbf{v}_1 + A\mathbf{v}_2) \\ &= A^{-1}A(\mathbf{v}_1 + \mathbf{v}_2) \\ &= \mathbf{v}_1 + \mathbf{v}_2 \\ &= A^{-1}\mathbf{v}_1 + A^{-1}\mathbf{v}_2 \end{aligned}$$

$$A^{-1}(c\mathbf{x}) = A^{-1}(cA\mathbf{y}) = A^{-1}A(c\mathbf{y}) = c\mathbf{y} = cA^{-1}\mathbf{x}.$$

The inverse of  $A^{-1}$  is  $A$  since

$$A(A^{-1}\mathbf{x}) = A(A^{-1}A\mathbf{y}) = A\mathbf{y} = \mathbf{x}$$

### Rudin 9.3

Suppose  $A$  is not 1-1. Then for some  $\mathbf{y} \in Y$ , there exists distinct  $\mathbf{v}, \mathbf{w} \in X$  such that  $A\mathbf{v} = \mathbf{y}$  and  $A\mathbf{w} = \mathbf{y}$ . Subtracting these two equations implies that

$$A(\mathbf{v} - \mathbf{w}) = 0$$

which contradicts our assumption that  $A\mathbf{x} = 0$  only when  $\mathbf{x} = 0$ .

## Rudin 9.4

Let  $A \in L(X, Y)$  be a linear transformation. Let  $\mathbf{x}, \mathbf{y} \in \mathcal{N}(A)$ .  $\mathcal{N}(A)$  is a vector space since

$$A(\mathbf{x} + \mathbf{y}) = A\mathbf{x} + A\mathbf{y} = \mathbf{0}$$

$$A(c\mathbf{x}) = cA\mathbf{x} = \mathbf{0}$$

Let  $\mathbf{x}, \mathbf{y} \in \mathcal{R}(A)$ . We can write  $\mathbf{x} = A\mathbf{p}$  and  $\mathbf{y} = A\mathbf{q}$  for some  $\mathbf{p}, \mathbf{q} \in X$ . By the linearity of  $A$ ,

$$\mathbf{x} + \mathbf{y} = A\mathbf{p} + A\mathbf{q} = A(\mathbf{p} + \mathbf{q}) \in \mathcal{R}(A)$$

$$c\mathbf{x} = cA\mathbf{p} = A(c\mathbf{p}) \in \mathcal{R}(A)$$

## Rudin 9.5

Let  $\mathbf{x} = x_1\mathbf{e}_1 + \cdots + x_n\mathbf{e}_n$  for the standard basis vectors  $\mathbf{e}_i$ . If we let  $\mathbf{y} \in \mathbb{R}^n$  with  $\mathbf{y}_i = A\mathbf{e}_i$  then

$$A\mathbf{x} = A(x_1\mathbf{e}_1 + \cdots + x_n\mathbf{e}_n) = c_1\mathbf{y}_1 + \cdots + c_n\mathbf{y}_n = \mathbf{x} \cdot \mathbf{y}.$$

It is unique since if there was  $\mathbf{z}$  such that  $A\mathbf{x} = \mathbf{x} \cdot \mathbf{z}$ , then

$$|\mathbf{y} - \mathbf{z}|^2 = \mathbf{y} \cdot \mathbf{y} - \mathbf{y} \cdot \mathbf{z} - \mathbf{z} \cdot \mathbf{y} - \mathbf{z} \cdot \mathbf{z} = A(\mathbf{y}) - A(\mathbf{y}) - A(\mathbf{z}) + A(\mathbf{z}).$$

By the Schwarz inequality,

$$||A|| = \sup |A\mathbf{x}| = \sup |\mathbf{x} \cdot \mathbf{y}| \leq \sup |\mathbf{x}| |\mathbf{y}|.$$

which implies that  $||A|| \leq |\mathbf{y}|$ . Also note that  $A\left(\frac{\mathbf{y}}{|\mathbf{y}|}\right) = \frac{\mathbf{y}}{|\mathbf{y}|} \cdot \mathbf{y} = |\mathbf{y}|$  so  $||A|| \geq |\mathbf{y}|$ .