## Chapter 9: Functions of Several Variables

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**Exercise 9.1.** If S is a nonempty subset of a vector space X, prove (as asserted in Section 9.1) that the span of S is a vector space.

Denote the span of S by span(S).

Proof.

- (1) Since  $S \neq \emptyset$ , there is  $\mathbf{z} \in S$ . So  $1\mathbf{z} = \mathbf{z} \in \text{span}(S) \neq \emptyset$ . (In fact,  $\text{span}(S) \supseteq S$ .)
- (2) If  $\mathbf{x}, \mathbf{y} \in \text{span}(S)$ , then there exist elements  $\mathbf{x}_1, \dots, \mathbf{x}_m, \mathbf{y}_1, \dots, \mathbf{y}_n \in S$  and scalars  $a_1, \dots, a_m, b_1, \dots, b_n$  such that

$$\mathbf{x} = a_1 \mathbf{x}_1 + \dots + a_m \mathbf{x}_m,$$
  
$$\mathbf{y} = b_1 \mathbf{y}_1 + \dots + b_n \mathbf{y}_n.$$

Then

$$\mathbf{x} + \mathbf{y} = a_1 \mathbf{x}_1 + \dots + a_m \mathbf{x}_m + b_1 \mathbf{y}_1 + \dots + b_n \mathbf{y}_n$$

is a linear combination of the elements of S. For any scalar c,

$$c\mathbf{x} = (ca_1)\mathbf{x}_1 + \dots + (ca_m)\mathbf{x}_m$$

is again linear combination of the elements of S.

(3) By (1)(2), span(S) is a vector space.

*Note.* Any subspace of X that contains S must also contain span(S).

**Exercise 9.2.** Prove (as asserted in Section 9.6) that BA is linear if A and B are linear transformations. Prove also that  $A^{-1}$  is linear and invertible if A is invertible.

*Proof.* Use the notation in Definitions 9.6.

(1) Show that BA is linear if A and B are linear transformations. Let X, Y, Z be vector spaces,  $A \in L(X, Y)$  and  $B \in L(Y, Z)$ .

(a) Given any  $\mathbf{x}_1, \mathbf{x}_2 \in X$ .

$$(BA)(\mathbf{x}_1 + \mathbf{x}_2) = B(A(\mathbf{x}_1 + \mathbf{x}_2))$$
  
=  $B(A\mathbf{x}_1 + A\mathbf{x}_2)$  (A is a linear transformation)  
=  $B(A\mathbf{x}_1) + B(A\mathbf{x}_2)$  (B is a linear transformation)  
=  $(BA)\mathbf{x}_1 + (BA)\mathbf{x}_2$ .

(b) For any  $\mathbf{x} \in X$  and scalar c,

$$(BA)(c\mathbf{x}) = B(A(c\mathbf{x}))$$
  
=  $B(cA\mathbf{x})$  (A is a linear transformation)  
=  $cB(A\mathbf{x})$  (B is a linear transformation)  
=  $c(BA)\mathbf{x}$ .

By (a)(b),  $BA \in L(X, Z)$ .

- (2) Show that  $A^{-1}$  is linear if A is invertible.
  - (a) Given any  $\mathbf{y}_1, \mathbf{y}_2 \in X$ . Since A is surjective, there exist  $\mathbf{x}_1, \mathbf{x}_2 \in X$  such that

$$\mathbf{y}_1 = A\mathbf{x}_1$$
$$\mathbf{y}_2 = A\mathbf{x}_2.$$

So

$$A^{-1}\mathbf{y}_1 = A^{-1}(A\mathbf{x}_1) = \mathbf{x}_1$$
  
 $A^{-1}\mathbf{y}_2 = A^{-1}(A\mathbf{x}_2) = \mathbf{x}_2$ 

(by Definitions 9.4). Hence

$$A^{-1}(\mathbf{y}_1 + \mathbf{y}_2) = A^{-1}(A\mathbf{x}_1 + A\mathbf{x}_2)$$

$$= A^{-1}(A(\mathbf{x}_1 + \mathbf{x}_2)) \quad (A \text{ is a linear transformation})$$

$$= \mathbf{x}_1 + \mathbf{x}_2 \qquad (Definitions 9.4)$$

$$= A^{-1}\mathbf{y}_1 + A^{-1}\mathbf{y}_2.$$

(b) For any  $\mathbf{y} \in X$  and scalar c, there is a corresponding  $\mathbf{x} \in X$  such that  $\mathbf{y} = A\mathbf{x}$  since A is surjective. So  $A^{-1}\mathbf{y} = \mathbf{x}$  by Definition 9.4. Hence

$$A^{-1}(c\mathbf{y}) = A^{-1}(cA\mathbf{x})$$
  
=  $A^{-1}(A(c\mathbf{x}))$  (A is a linear transformation)  
=  $c\mathbf{x}$  (Definitions 9.4)  
=  $cA^{-1}\mathbf{y}$ .

By (a)(b),  $A^{-1} \in L(X)$ .

- (3) Show that  $A^{-1}$  is invertible if A is invertible. It suffices to show that  $A^{-1}$  is injective and surjective.
  - (a) Show that  $A^{-1}$  is injective. Given any  $\mathbf{y}_1, \mathbf{y}_2 \in X$ . Since A is surjective, there exist  $\mathbf{x}_1, \mathbf{x}_2 \in X$  such that

$$\mathbf{y}_1 = A\mathbf{x}_1$$
$$\mathbf{y}_2 = A\mathbf{x}_2.$$

Suppose  $A^{-1}\mathbf{y}_1 = A^{-1}\mathbf{y}_2$ . So  $A^{-1}(A\mathbf{x}_1) = A^{-1}(A\mathbf{x}_2)$ , or  $\mathbf{x}_1 = \mathbf{x}_2$ , or  $\mathbf{y}_1 = A\mathbf{x}_1 = A\mathbf{x}_2 = \mathbf{y}_2$ .

(b) Show that  $A^{-1}$  is surjective. For any  $\mathbf{x} \in X$ , there exists  $A\mathbf{x} \in X$  such that  $A^{-1}(A\mathbf{x}) = \mathbf{x}$  by Definitions 9.4.

**Exercise 9.3.** Assume  $A \in L(X,Y)$  and  $A\mathbf{x} = \mathbf{0}$  only when  $\mathbf{x} = \mathbf{0}$ . Prove that A is then 1-1.

*Proof.* Suppose  $A\mathbf{x} = A\mathbf{y}$ . Since A is a linear transformation,  $A(\mathbf{x} - \mathbf{y}) = A\mathbf{x} - A\mathbf{y} = \mathbf{0}$ . By assumption,  $\mathbf{x} - \mathbf{y} = \mathbf{0}$  or  $\mathbf{x} = \mathbf{y}$ .  $\square$ 

Exercise 9.4. Prove (as asserted in Section 9.30) that null spaces and ranges of linear transformations are vector spaces.

*Proof.* Use the notation in Definitions 9.30. Suppose X, Y are vector spaces, and  $A \in L(X,Y)$ , as in Definition 9.6.

- (1) Show that  $\mathcal{N}(A)$  is a vector space in X.
  - (a) Note that  $\mathbf{0} \in X$ . Since  $A\mathbf{0} = \mathbf{0}$ ,  $\mathbf{0} \in \mathcal{N}(A) \neq \emptyset$ .
  - (b) Suppose  $\mathbf{x}_1, \mathbf{x}_2 \in \mathcal{N}(A)$ . Then

$$\begin{aligned} A(\mathbf{x}_1 + \mathbf{x}_2) &= A\mathbf{x}_1 + A\mathbf{x}_2 & \quad (A \text{ is a linear transformation}) \\ &= \mathbf{0} + \mathbf{0} & \quad (\mathbf{x}_1, \mathbf{x}_2 \in \mathcal{N}(A)) \\ &= \mathbf{0}. \end{aligned}$$

So  $\mathbf{x}_1 + \mathbf{x}_2 \in \mathcal{N}(A)$ .

(c) Suppose  $\mathbf{x} \in \mathcal{N}(A)$  and c is a scalar. Then

$$A(c\mathbf{x}) = cA\mathbf{x}$$
 (A is a linear transformation)  
=  $c\mathbf{0}$  ( $\mathbf{x} \in \mathcal{N}(A)$ )  
=  $\mathbf{0}$ .

So  $c\mathbf{x} \in \mathcal{N}(A)$ .

By (a)(b)(c),  $\mathcal{N}(A)$  is a vector space.

- (2) Show that  $\mathcal{R}(A)$  is a vector space in Y.
  - (a) Note that  $\mathbf{0} \in X$ . So  $A\mathbf{0} = \mathbf{0} \in \mathcal{R}(A) \neq \emptyset$ .
  - (b) Suppose  $\mathbf{y}_1, \mathbf{y}_2 \in \mathcal{R}(A)$ . Then there exist  $\mathbf{x}_1, \mathbf{x}_2 \in X$  such that  $A\mathbf{x}_1 = \mathbf{y}_1$  and  $A\mathbf{x}_2 = \mathbf{y}_2$ . Hence

$$\mathbf{y}_1 + \mathbf{y}_2 = A\mathbf{x}_1 + A\mathbf{x}_2$$
  
=  $A(\mathbf{x}_1 + \mathbf{x}_2)$  (A is a linear transformation).

So  $\mathbf{y}_1 + \mathbf{y}_2 \in \mathscr{R}(A)$ .

(c) Suppose  $\mathbf{y} \in \mathcal{R}(A)$  and c is a scalar. Then there exists  $\mathbf{x} \in X$  such that  $A\mathbf{x} = \mathbf{y}$ . Hence

$$c\mathbf{y} = cA\mathbf{x}$$
  
=  $A(c\mathbf{x})$  (A is a linear transformation).

So  $c\mathbf{y} \in \mathcal{R}(A)$ .

By (a)(b)(c),  $\mathcal{R}(A)$  is a vector space.

**Exercise 9.5.** Prove that to every  $A \in L(\mathbb{R}^n, \mathbb{R}^1)$  corresponds a unique  $\mathbf{y} \in \mathbb{R}^n$  such that  $A\mathbf{x} = \mathbf{x} \cdot \mathbf{y}$ . Prove also that  $||A|| = |\mathbf{y}|$ . (Hint: Under certain conditions, equality holds in the Schwarz inequality.)

Proof.

- (1) Recall that  $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$  is the standard basis of  $\mathbb{R}^n$  (Definitions 9.1). Given any  $\mathbf{x} \in \mathbb{R}^n$ , write  $\mathbf{x} = (x_1, \dots, x_n)$  as  $\mathbf{x} = \sum x_i \mathbf{e}_i$ .
- (2) Show that y exists. Since A is a linear transformation,

$$A\mathbf{x} = A\left(\sum x_j \mathbf{e}_j\right)$$

$$= \sum x_j A \mathbf{e}_j$$

$$= (x_1, \dots, x_n) \cdot (A \mathbf{e}_1, \dots, A \mathbf{e}_n)$$

$$= \mathbf{x} \cdot \sum (A \mathbf{e}_j) \mathbf{e}_j.$$

Define  $\mathbf{y} = \sum (A\mathbf{e}_i)\mathbf{e}_i \in \mathbb{R}^n$  so that  $A\mathbf{x} = \mathbf{x} \cdot \mathbf{y}$ .

(3) Show that **y** is unique. Suppose there exists some  $\mathbf{z} \in \mathbb{R}^n$  such that  $A\mathbf{x} = \mathbf{x} \cdot \mathbf{z}$ . So

$$0 = A\mathbf{x} - A\mathbf{x}$$
$$= \mathbf{x} \cdot \mathbf{y} - \mathbf{x} \cdot \mathbf{z}$$
$$= \mathbf{x} \cdot (\mathbf{y} - \mathbf{z})$$

for any  $\mathbf{x} \in \mathbb{R}^n$ . In particular, take  $\mathbf{x} = \mathbf{y} - \mathbf{z} \in \mathbb{R}^n$  to get

$$0 = (\mathbf{y} - \mathbf{z}) \cdot (\mathbf{y} - \mathbf{z}) = |\mathbf{y} - \mathbf{z}|^2$$

or y - z = 0 or y = z.

(4) Show that  $||A|| = |\mathbf{y}|$ . By the Schwarz inequality (Theorem 1.37(d)),

$$|A\mathbf{x}| = |\mathbf{x} \cdot \mathbf{y}| \le |\mathbf{x}||\mathbf{y}| \le |\mathbf{y}|$$

as  $|\mathbf{x}| \leq 1$ . Take the sup over all  $|\mathbf{x}| \leq 1$  to get

$$||A|| \leq |\mathbf{y}|.$$

If  $\mathbf{y} = \mathbf{0}$ , then  $||A|| = |\mathbf{y}| = 0$ . If  $\mathbf{y} \neq \mathbf{0}$ , then the equality holds when  $\mathbf{x} = \frac{\mathbf{y}}{|\mathbf{y}|} \in \mathbb{R}^n$ . (Here  $|\mathbf{x}| = 1$ .)

**Exercise 9.6.** If f(0,0) = 0 and

$$f(x,y) = \frac{xy}{x^2 + y^2}$$
 if  $(x,y) \neq (0,0)$ ,

prove that  $(D_1f)(x,y)$  and  $(D_2f)(x,y)$  exist at every point of  $\mathbb{R}^2$ , although f is not continuous at (0,0).

Proof.

(1) Show that

$$(D_1 f)(x,y) = \begin{cases} 0 & \text{if } (x,y) = (0,0), \\ \frac{y(y^2 - x^2)}{(x^2 + y^2)^2} & \text{if } (x,y) \neq (0,0). \end{cases}$$

Write

$$(D_1 f)(x,y) = \lim_{t \to 0} \frac{f((x,y) + t(1,0)) - f(x,y)}{t}$$
$$= \lim_{t \to 0} \frac{f(x+t,y) - f(x,y)}{t}.$$

If (x, y) = (0, 0),

$$(D_1 f)(0,0) = \lim_{t \to 0} \frac{f(t,0) - f(0,0)}{t} = \lim_{t \to 0} \frac{0 - 0}{t} = 0.$$

If  $(x, y) \neq (0, 0)$ ,

$$(D_1 f)(x,y) = \lim_{t \to 0} \frac{f(x+t,y) - f(x,y)}{t}$$

$$= \lim_{t \to 0} \frac{\frac{(x+t)y}{(x+t)^2 + y^2} - \frac{xy}{x^2 + y^2}}{t}$$

$$= \lim_{t \to 0} \frac{y(y^2 - x^2) - txy}{((x+t)^2 + y^2)(x^2 + y^2)}$$

$$= \frac{y(y^2 - x^2)}{(x^2 + y^2)^2}.$$

(2) Show that

$$(D_2 f)(x,y) = \begin{cases} 0 & \text{if } (x,y) = (0,0), \\ \frac{x(x^2 - y^2)}{(x^2 + y^2)^2} & \text{if } (x,y) \neq (0,0). \end{cases}$$

Similar to (1).

(3) Show that f is not continuous at (0,0). Note that

$$\lim_{n\to\infty} f\left(\frac{1}{n},\frac{1}{n}\right) = \lim_{n\to\infty} \frac{\frac{1}{n}\cdot\frac{1}{n}}{\frac{1}{n^2}+\frac{1}{n^2}} = \lim_{n\to\infty} \frac{1}{2} = \frac{1}{2}$$

and

$$\lim_{n\to\infty} f\left(\frac{1}{n},0\right) = \lim_{n\to\infty} \frac{0}{\frac{1}{n^2}+0} = \lim_{n\to\infty} 0 = 0.$$

Hence the limit  $\lim_{(x,y)\to(0,0)} f(x,y)$  does not exist.

## Exercise 9.7. ...

Proof.

- (1)
- (2)

## Exercise 9.8. ...

Proof.

- (1)
- (2)

Exercise 9.9
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Exercise 9.10
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Exercise 9.27
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Proof. (1) (2) □  Exercise 9.30  Proof.
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(1)

(2)