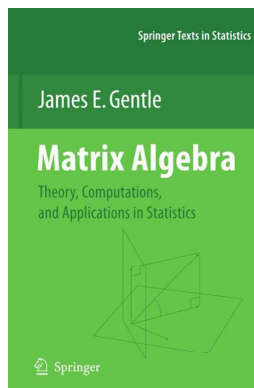


# Theory of Matrices

## Chapter 2. Vectors and Vector Spaces

### 2.1. Operations on Vectors—Proofs of Theorems



## Theorem 2.1.1(A1)

### Theorem 2.1.1. Properties of Vector Algebra in $\mathbb{R}^n$ .

Let  $x, y, z \in \mathbb{R}^n$ . Then:

**A1.**  $(x + y) + z = x + (y + z)$  (Associativity of Vector Addition)

**Proof.** Let  $x, y, z \in \mathbb{R}^n$  be  $x = [x_1, x_2, \dots, x_n]$ ,  $y = [y_1, y_2, \dots, y_n]$ , and  $z = [z_1, z_2, \dots, z_n]$ . Then:

$$\begin{aligned}(x + y) + z &= ([x_1, x_2, \dots, x_n] + [y_1, y_2, \dots, y_n]) + [z_1, z_2, \dots, z_n] \\ &= [x_1 + y_1, x_2 + y_2, \dots, x_n + y_n] + [z_1, z_2, \dots, z_n] \\ &\quad \text{by the definition of vector addition} \\ &= [(x_1 + y_1) + z_1, (x_2 + y_2) + z_2, \dots, (x_n + y_n) + z_n] \\ &\quad \text{by the definition of vector addition} \\ &= [x_1 + (y_1 + z_1), x_2 + (y_2 + z_2), \dots, x_n + (y_n + z_n)] \\ &\quad \text{since addition in } \mathbb{R} \text{ is associative}\end{aligned}$$

## Theorem 2.1.1(A1) (continued)

### Theorem 2.1.1. Properties of Vector Algebra in $\mathbb{R}^n$ .

Let  $x, y, z \in \mathbb{R}^n$ . Then:

**A1.**  $(x + y) + z = x + (y + z)$  (Associativity of Vector Addition)

**Proof (continued).**

$$\begin{aligned}(x + y) + z &= [x_1 + (y_1 + z_1), x_2 + (y_2 + z_2), \dots, x_n + (y_n + z_n)] \\ &\quad \text{since addition in } \mathbb{R} \text{ is associative} \\ &= [x_1, x_2, \dots, x_n] + [y_1 + z_1, y_2 + z_2, \dots, y_n + z_n] \\ &\quad \text{by the definition of vector addition} \\ &= [x_1, x_2, \dots, x_n] + ([y_1, y_2, \dots, y_n] + [z_1, z_2, \dots, z_n]) \\ &\quad \text{by the definition of vector addition} \\ &= x + (y + z).\end{aligned}$$

□

## Theorem 2.1.2

**Theorem 2.1.2.** Let  $V_1$  and  $V_2$  be vector spaces of  $n$ -vectors. Then  $V_1 \cap V_2$  is a vector space.

**Proof.** By our definition of “vector space,” we only need to prove that  $V_1 \cap V_2$  is closed under linear combinations. Let  $x, y \in V_1 \cap V_2$  and  $a, b \in \mathbb{R}$ . Since  $V_1$  is a vector space then it is closed under linear combinations and so  $ax + by \in V_1$ . Similarly,  $ax + by \in V_2$ . So  $ax + by \in V_1 \cap V_2$ . Since  $x$  and  $y$  are arbitrary elements of  $V_1 \cap V_2$  and  $a, b \in \mathbb{R}$  are arbitrary scalars, then  $V_1 \cap V_2$  is closed under linear combinations. That is,  $V_1 \cap V_2$  is a vector space. □

## Theorem 2.1.3

**Theorem 2.1.3.** If  $V_1$  and  $V_2$  are vector spaces of  $n$ -vectors, then  $V_1 + V_2$  is a vector space.

**Proof.** By our definition of “vector space,” we must show that  $V_1 + V_2$  is closed under linear combinations. Let  $x, y \in V_1 + V_2$  and let  $a, b \in \mathbb{R}$ . Then  $x = x_1 + x_2$  and  $y = y_1 + y_2$  for some  $x_1, y_1 \in V_1$  and  $x_2, y_2 \in V_2$ . Since  $V_1$  and  $V_2$  are vector spaces, then they are closed under linear combinations and so  $ax_1 + by_1 \in V_1$  and  $ax_2 + by_2 \in V_2$ . Therefore

$$ax + by = a(x_1 + x_2) + b(y_1 + y_2) = (ax_1 + by_1) + (ax_2 + by_2) \in V_1 + V_2.$$

Since  $x$  and  $y$  are arbitrary vectors in  $V_1 + V_2$  and  $a, b \in \mathbb{R}$  are arbitrary scalars, then we have that  $V_1 + V_2$  is closed under linear combinations. That is,  $V_1 + V_2$  is a vector space.  $\square$

## Theorem 2.1.4

**Theorem 2.1.4.** If vector spaces  $V_1$  and  $V_2$  are essentially disjoint then every element of  $V_1 \oplus V_2$  can be written as  $v_1 + v_2$ , where  $v_1 \in V_1$  and  $v_2 \in V_2$ , in a unique way.

**Proof.** Let  $V_1$  and  $V_2$  be essentially disjoint vector spaces of  $n$ -vectors; that is,  $V_1 \cap V_2 = \{0\}$ . Suppose some  $v \in V_1 \oplus V_2$  is of the form  $v = v_1 + v_2 = v'_1 + v'_2$  where  $v_1, v'_1 \in V_1$  and  $v_2, v'_2 \in V_2$ . Then  $v_1 - v'_1 = v'_2 - v_2$ . So  $v_1 - v'_1 \in V_1$  and  $v'_2 - v_2 \in V_2$  since  $V_1$  and  $V_2$  are vector space. But then  $v_1 - v'_1, v'_2 - v_2 \in V_1 \cap V_2$  and so  $v_1 - v'_1 = 0$  and  $v'_2 - v_2 = 0$ . That is,  $v_1 = v'_1$  and  $v_2 = v'_2$ . So the representation of  $v \in V_1 \oplus V_2$  as a sum of an element of  $V_1$  and an element of  $V_2$  is unique, as claimed.  $\square$

## Theorem 2.1.5

**Theorem 2.1.5.** If  $\{v_1, v_2, \dots, v_k\}$  is a basis for a vector space  $V$ , then each element can be uniquely represented as a linear combination of the basis vectors.

**Proof.** Suppose that

$$x = b_1 v_1 + b_2 v_2 + \dots + b_k v_k = c_1 v_1 + c_2 v_2 + \dots + c_k v_k. \text{ Then}$$

$$\begin{aligned} 0 &= x - x = (b_1 v_1 + b_2 v_2 + \dots + b_k v_k) - (c_1 v_1 + c_2 v_2 + \dots + c_k v_k) \\ &= (b_1 - c_1)v_1 + (b_2 - c_2)v_2 + \dots + (b_k - c_k)v_k. \end{aligned}$$

Since the basis consists (by definition) of a linearly independent set of vectors, then  $b_1 - c_1 = b_2 - c_2 = \dots = b_k - c_k = 0$ ; that is,  $b_1 = c_1, b_2 = c_2, \dots, b_k = c_k$ . Therefore, the representation of  $x$  is unique. Since  $x$  is an arbitrary vector in  $V$ , the claim follows.  $\square$

## Theorem 2.1.6(3)

**Theorem 2.1.6. Properties of Inner Products.**

Let  $x, y \in \mathbb{R}^n$  and let  $a \in \mathbb{R}$ . Then:

**3.**  $a\langle x, y \rangle = \langle ax, y \rangle$  (Factoring of Scalar Multiplication in Inner Products).

**Proof.** Let  $x, y \in \mathbb{R}^n$  be  $x = [x_1, x_2, \dots, x_n]$  and  $y = [y_1, y_2, \dots, y_n]$ . Then

$$\begin{aligned} a\langle x, y \rangle &= a\langle [x_1, x_2, \dots, x_n], [y_1, y_2, \dots, y_n] \rangle \\ &= a(x_1 y_1 + x_2 y_2 + \dots + x_n y_n) \text{ by the definition of } \langle x, y \rangle \\ &= a(x_1 y_1) + a(x_2 y_2) + \dots + a(x_n y_n) \\ &\quad \text{by distribution property of multiplication over addition in } \mathbb{R} \\ &= (ax_1)y_1 + (ax_2)y_2 + \dots + (ax_n)y_n \\ &\quad \text{by associativity for multiplication in } \mathbb{R} \\ &= \langle [ax_1, ax_2, \dots, ax_n], [y_1, y_2, \dots, y_n] \rangle \\ &\quad \text{by the definition of inner product} \\ &= \langle ax, y \rangle. \quad \square \end{aligned}$$

## Theorem 2.1.7

**Theorem 2.1.7. Schwarz Inequality.**

For any  $x, y \in \mathbb{R}^n$  we have  $|\langle x, y \rangle| \leq \langle x, x \rangle^{1/2} \langle y, y \rangle^{1/2}$ .

**Proof.** Let  $t \in \mathbb{R}$ . Then

$$\begin{aligned} 0 &\leq \langle (tx + y), (tx + y) \rangle \text{ by Theorem 2.1.6(1)} \\ &= t\langle x, tx + y \rangle + \langle y, tx + y \rangle \text{ by linearity in the 1st entry} \\ &= t(t\langle x, x \rangle + \langle x, y \rangle) + (t\langle y, x \rangle + \langle y, y \rangle) \text{ by linearity in the 2nd entry} \\ &= t^2\langle x, x \rangle + 2t\langle x, y \rangle + \langle y, y \rangle \text{ by Theorem 1.2.6(2)} \\ &= at^2 + bt + c \end{aligned}$$

where  $a = \langle x, x \rangle$ ,  $b = 2\langle x, y \rangle$ , and  $c = \langle y, y \rangle$ . As a quadratic in  $t$ ,  $at^2 + bt + c$  cannot have two distinct roots or else we would have  $at^2 + bt + c < 0$  for some  $t$ . This means that the discriminant  $b^2 - 4ac$  in the quadratic equation  $t = (-b \pm \sqrt{b^2 - 4ac})/(2a)$ , must be  $b^2 - 4ac \leq 0$ ; that is,  $(b/2)^2 \leq ac$ . Hence, we have  $(b/2)^2 = \langle x, y \rangle^2 \leq ac = \langle x, x \rangle \langle y, y \rangle$  or  $\sqrt{\langle x, y \rangle^2} = |\langle x, y \rangle| \leq \langle x, x \rangle^{1/2} \langle y, y \rangle^{1/2}$ .  $\square$

()

## Theorem 2.1.8

**Theorem 2.1.8** The basis norm is indeed a norm for any basis  $\{v_1, v_2, \dots, v_k\}$  of vector space  $V$ .

**Proof.** Let  $x = c_1 v_1 + c_2 v_2 + \dots + c_k v_k$  and  $y = d_1 v_1 + d_2 v_2 + \dots + d_k v_k$ . If  $x \neq 0$  then some  $c_i \neq 0$  and so  $\rho(x) > 0$ . Clearly  $\rho(0) = 0$ . So “Nonnegativity and Mapping of the Identity” is satisfied. Next

$$\rho(ax) = \rho(a(c_1 v_1 + c_2 v_2 + \dots + c_k v_k)) = \rho((ac_1)v_1 + (ac_2)v_2 + \dots + (ac_k)v_k)$$

$$= \left\{ \sum_{j=1}^k (ac_j)^2 \right\}^{1/2} = |a| \left\{ \sum_{j=1}^k c_j^2 \right\}^{1/2} = |a| \rho(x)$$

and “Relation of Scalar Multiplication to Real Multiplication” holds.

()

## Theorem 2.1.8 (continued 1)

**Proof (continued).** Finally,

$$\begin{aligned} \rho(x + y)^2 &= \rho((c_1 + d_1)v_1 + (c_2 + d_2)v_2 + \dots + (c_k + d_k)v_k)^2 \\ &= \sum_{j=1}^k (c_j + d_j)^2 = \sum_{j=1}^k (c_j^2 + 2c_j d_j + d_j^2) \\ &= \sum_{j=1}^k c_j^2 + 2 \sum_{j=1}^k c_j d_j + \sum_{j=1}^k d_j^2 \\ &\leq \sum_{j=1}^k c_j^2 + 2 \left\{ \sum_{j=1}^k c_j^2 \right\}^{1/2} \left\{ \sum_{j=1}^k d_j^2 \right\}^{1/2} + \sum_{j=1}^k d_j^2 \\ &\quad \text{by Theorem 2.1.7 (Schwarz's Inequality in } \mathbb{R}^n) \\ &= \left( \left\{ \sum_{j=1}^k c_j^2 \right\}^{1/2} + \left\{ \sum_{j=1}^k d_j^2 \right\}^{1/2} \right)^2 = (\rho(x) + \rho(y))^2. \end{aligned}$$

()

## Theorem 2.1.8 (continued 2)

**Theorem 2.1.8** The basis norm is indeed a norm for any basis  $\{v_1, v_2, \dots, v_k\}$  of vector space  $V$ .

**Proof (continued).** ...

$$\rho(x + y)^2 = (\rho(x) + \rho(y))^2.$$

Taking square roots,  $\rho(x + y) \leq \rho(x) + \rho(y)$  and so the Triangle Inequality holds. Therefore  $\rho$  is a metric on  $V$ .  $\square$

()

## Theorem 2.1.10

**Theorem 2.1.10.** Every norm on (finite dimensional vector space)  $V$  is equivalent to the basis norm  $\rho$  for any given basis  $\{v_1, v_2, \dots, v_k\}$ . Therefore, any two norms on  $V$  are equivalent.

**Proof.** Let  $\|\cdot\|_a$  be any norm on vector space  $V$  and let  $\{v_1, v_2, \dots, v_k\}$  be a basis for the space. Then for some unique scalars  $c_1, c_2, \dots, c_k \in \mathbb{R}$  we have  $x = c_1 v_1 + c_2 v_2 + \dots + c_k v_k$ . Then, by the Triangle Inequality and “Relation of Scalar Multiplication to Real Multiplication,”

$$\|x\|_a = \left\| \sum_{i=1}^k c_i v_i \right\|_a \leq \sum_{i=1}^k |c_i| \|v_i\|_a.$$

Now with  $[|c_1|, |c_2|, \dots, |c_k|], [\|v_1\|_a, \|v_2\|_a, \dots, \|v_k\|_a] \in \mathbb{R}^k$  we have by the Schwarz Inequality (Theorem 2.1.7) that

$$\sum_{i=1}^k |c_i| \|v_i\|_a \leq \left\{ \sum_{i=1}^k |c_i|^2 \right\}^{1/2} \left\{ \sum_{i=1}^k \|v_i\|_a^2 \right\}^{1/2}.$$

()

## Theorem 2.1.10 (continued 1)

**Proof (continued).** Hence

$$\|x\|_a \leq \left\{ \sum_{i=1}^k \|v_i\|_a^2 \right\}^{1/2} \rho(x) = \tilde{s} \rho(x) \text{ for } \tilde{s} = \left\{ \sum_{i=1}^k \|v_i\|_a^2 \right\}^{1/2}.$$

Next, let  $C = \left\{ x = \sum_{i=1}^k u_i v_i \in V \mid \sum_{i=1}^k |u_i|^2 = 1 \right\}$ . Gentle states that set  $C$  is “obviously [topologically] closed” (page 20). Set  $C$  is the surface of the unit sphere in  $V$  under  $\rho$ ,  $C = \{x \in V \mid \rho(x) = 1\}$ . We give a proof that  $C$  is a topologically closed set by showing that its complement,  $V \setminus C$ , is open. Let  $x \in V \setminus C$  and let  $\varepsilon = |1 - \rho(x)| > 0$ . Then the open ball  $\{v \in V \mid \rho(v - x) < \varepsilon\}$  contains no elements of  $C$ : for  $y \in C$ ,

$$\rho(y - x) \geq \begin{cases} \rho(y) - \rho(x) \\ \rho(x) - \rho(y) \end{cases} = \begin{cases} 1 - \rho(x) \\ \rho(x) - 1 \end{cases} = \begin{cases} \varepsilon & \text{if } \rho(x) < 1 \\ \varepsilon & \text{if } \rho(x) > 1. \end{cases}$$

(Notice that the Triangle Inequality for norms implies for any  $x, y \in V$  that  $\|x\| = \|x - y + y\| \leq \|x - y\| + \|y\|$  or  $\|x - y\| \geq \|x\| - \|y\|$ .)

()

## Theorem 2.1.10 (continued 2)

**Proof (continued).** Define  $f : C \rightarrow \mathbb{R}$  as  $f(u) = \left\| \sum_{i=1}^k u_i v_i \right\|_a$ . Gentle claims that  $f$  is continuous (page 20). Let's prove this. Let  $y = \sum_{i=1}^k u_i v_i \in C$  and let  $\varepsilon > 0$ . Set  $\delta = \varepsilon$ . For any  $x = \sum_{i=1}^k u'_i v_i \in C$  with  $\|y - x\|_a < \delta$  we have

$$\begin{aligned} \varepsilon = \delta > \begin{cases} \|y\|_a - \|x\|_a \\ \|x\|_a - \|y\|_a \end{cases} &= \begin{cases} \left\| \sum_{i=1}^k u_i v_i \right\|_a - \left\| \sum_{i=1}^k u'_i v_i \right\|_a \\ \left\| \sum_{i=1}^k u'_i v_i \right\|_a - \left\| \sum_{i=1}^k u_i v_i \right\|_a \end{cases} \\ &= \begin{cases} f(y) - f(x) \\ f(x) - f(y) \end{cases} = \begin{cases} |f(y) - f(x)| & \text{if } f(y) \geq f(x) \\ |f(x) - f(y)| & \text{if } f(y) < f(x). \end{cases} \end{aligned}$$

That is,  $|f(y) - f(x)| < \varepsilon$ . So  $f : C \rightarrow \mathbb{R}$  is continuous.

()

## Theorem 2.1.10 (continued 3)

**Proof (continued).** By the Heine-Borel Theorem (since  $C$  is closed and bounded and  $V$  is finite dimensional),  $C$  is compact and so continuous function  $f$  attains a minimum value on  $C$ , say  $f(u_*) \leq f(u)$  for all  $u \in C$ . Let  $\tilde{r} = f(u_*) > 0$ . If  $x = \sum_{i=1}^k c_i v_i \neq 0$  then

$$\|x\|_a = \left\| \sum_{i=1}^k c_i v_i \right\|_a = \left\{ \sum_{j=1}^k c_j^2 \right\}^{1/2} \left\| \sum_{i=1}^k \left( \frac{c_i}{\left\{ \sum_{j=1}^k c_j^2 \right\}^{1/2}} \right) v_i \right\|_a = \rho(x) f(\tilde{c})$$

where  $\tilde{c} = \sum_{i=1}^k \left( c_i / \left\{ \sum_{j=1}^k c_j^2 \right\}^{1/2} \right) v_i$ , so  $\tilde{c} \in C$  since

$$\rho(\tilde{c}) = \sum_{i=1}^k \left| \frac{c_i}{\left\{ \sum_{j=1}^k c_j^2 \right\}^{1/2}} \right|^2 = \frac{1}{\sum_{j=1}^k c_j^2} \sum_{i=1}^k c_i^2 = 1.$$

()

## Theorem 2.1.10 (continued 4)

**Theorem 2.1.10.** Every norm on (finite dimensional vector space)  $V$  is equivalent to the basis norm  $\rho$  for any given basis  $\{v_1, v_2, \dots, v_k\}$ . Therefore, any two norms on  $V$  are equivalent.

**Proof (continued).** Since  $\tilde{r} \in C$  then  $f(\tilde{c}) \geq \tilde{r}$ , and so  $\|x\|_a \geq \tilde{r}\rho(x)$  for all  $x \in V$ ,  $x \neq 0$ . Of course  $\|x\|_a \geq \tilde{r}\rho(x)$  for  $x = 0$ , so for all  $x \in V$  we have  $\tilde{r}\rho(x) \leq \|x\|_a \leq \tilde{s}\rho(x)$ . That is,  $\|\cdot\|_a \cong \rho(\cdot)$ . Since  $\cong$  is an equivalence relation for Theorem 2.1.9, we have that any two norms on  $V$  are equivalent.  $\square$

## Theorem 2.1.11

**Theorem 2.1.11.** A set of nonzero vectors  $\{v_1, v_2, \dots, v_k\}$  in a vector space with an inner product for which  $\langle v_i, v_j \rangle = 0$  for  $i \neq j$  (the vectors are said to be *mutually orthogonal*) is a linearly independent set.

**Proof.** Let  $\{v_1, v_2, \dots, v_k\}$  be a set of mutually orthogonal nonzero vectors. ASSUME the set is not linearly independent. Then  $a_1 v_1 + a_2 v_2 + \dots + a_i v_i + \dots + a_k v_k = 0$  is satisfied where some coefficient is nonzero, say  $a_i \neq 0$ . So

$$v_i = (-a_1/a_i)v_1 + (-a_2/a_i)v_2 + \dots + (-a_{i-1}/a_i)v_{i-1} \\ + (-a_{i+1}/a_i)v_{i+1} + \dots + (-a_k/a_i)v_k.$$

But then

$$\langle v_i, v_i \rangle = \langle (-a_1/a_i)v_1 + (-a_2/a_i)v_2 + \dots + (-a_{i-1}/a_i)v_{i-1} \\ + (-a_{i+1}/a_i)v_{i+1} + \dots + (-a_k/a_i)v_k, v_i \rangle$$

## Theorem 2.1.11 (continued)

**Theorem 2.1.11.** A set of nonzero vectors  $\{v_1, v_2, \dots, v_k\}$  in a vector space with an inner product for which  $\langle v_i, v_j \rangle = 0$  for  $i \neq j$  (the vectors are said to be *mutually orthogonal*) is a linearly independent set.

**Proof.**

$$\langle v_i, v_i \rangle = (-a_1/a_i)\langle v_1, v_i \rangle + (-a_2/a_i)\langle v_2, v_i \rangle + \dots + (-a_{i-1}/a_i)\langle v_{i-1}, v_i \rangle \\ + (-a_{i+1}/a_i)\langle v_{i+1}, v_i \rangle + \dots + (-a_k/a_i)\langle v_k, v_i \rangle = 0,$$

a CONTRADICTION to the fact that  $v_i \neq 0$ . So the assumption that the set is not linearly independent is false; that is, the set is linearly independent, as claimed.  $\square$