

Theory of Matrices

Chapter 2. Vectors and Vector Spaces

2.1. Operations on Vectors—Proofs of Theorems

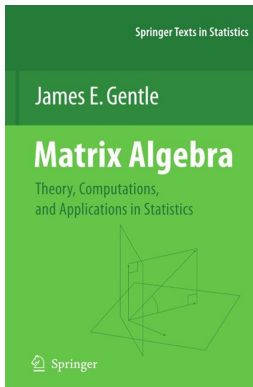


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Theorem 2.1.1(A1)

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}
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Theorem 2.1.1(A1) (continued)

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Proof (continued).

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 (x + y) + z &= [x_1 + (y_1 + z_1), x_2 + (y_2 + z_2), \dots, x_n + (y_n + z_n)] \\
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 &= [x_1, x_2, \dots, x_n] + [y_1 + z_1, y_2 + z_2, \dots, y_n + z_n] \\
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 &= [x_1, x_2, \dots, x_n] + ([y_1, y_2, \dots, y_n] + [z_1, z_2, \dots, z_n]) \\
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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.

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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 $x_2, y_2 \in V_2$.

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

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

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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 $v_1 - v'_1 = v'_2 - v_2$.

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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

$x = b_1v_1 + b_2v_2 + \dots + b_kv_k = c_1v_1 + c_2v_2 + \dots + c_kv_k$. Then

$$\begin{aligned} 0 &= x - x = (b_1v_1 + b_2v_2 + \dots + b_kv_k) - (c_1v_1 + c_2v_2 + \dots + c_kv_k) \\ &= (b_1 - c_1)v_1 + (b_2 - c_2)v_2 + \dots + (b_k - c_k)v_k. \end{aligned}$$

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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. □

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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_1y_1 + x_2y_2 + \dots + x_ny_n) \text{ by the definition of } \langle x, y \rangle \\
 &= a(x_1y_1) + a(x_2y_2) + \dots + a(x_ny_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} \\
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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 $c = \langle y, y \rangle$.

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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

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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.

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Next

$$\begin{aligned} \rho(ax) &= \rho(a(c_1 v_1 + c_2 v_2 + \dots + c_k v_k)) = \rho((ac_1)v_1 + (ac_2)v_2 + \dots + (ac_k)v_k) \\ &= \left\{ \sum_{j=1}^k (ac_j)^2 \right\}^{1/2} = |a| \left\{ \sum_{j=1}^k c_j^2 \right\}^{1/2} = |a|\rho(x) \end{aligned}$$

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

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

Theorem 2.1.8 (continued 1)

Proof (continued). Finally,

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

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 ρ 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 ρ 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,”

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

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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

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

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Theorem 2.1.10 (continued 1)

Proof (continued). Hence

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

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 ρ , $C = \{x \in V \mid \rho(x) = 1\}$.

Theorem 2.1.10 (continued 1)

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

Theorem 2.1.10 (continued 1)

Proof (continued). Hence

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(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 1)

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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.

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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

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

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Theorem 2.1.10 (continued 4)

Theorem 2.1.10. Every norm on (finite dimensional vector space) V is equivalent to the basis norm ρ 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 $a_i \neq 0$.

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

But then

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

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Theorem 2.1.11 (continued)

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Proof.

$$\begin{aligned} \langle v_i, v_i \rangle &= (-a_1/a_i)\langle v_1, v_i \rangle + (-a_2/a_i)\langle v_2, v_i \rangle + \cdots + (-a_{i-1}/a_i)\langle v_{i-1}, v_i \rangle \\ &\quad + (-a_{i+1}/a_i)\langle v_{i+1}, v_i \rangle + \cdots + (-a_k/a_i)\langle v_k, v_i \rangle = 0, \end{aligned}$$

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. □

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.

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$$\begin{aligned} \langle v_i, v_i \rangle &= (-a_1/a_i)\langle v_1, v_i \rangle + (-a_2/a_i)\langle v_2, v_i \rangle + \cdots + (-a_{i-1}/a_i)\langle v_{i-1}, v_i \rangle \\ &\quad + (-a_{i+1}/a_i)\langle v_{i+1}, v_i \rangle + \cdots + (-a_k/a_i)\langle v_k, v_i \rangle = 0, \end{aligned}$$

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