

CECC122: Linear Algebra and Matrix Theory Lecture Notes 5: Vector Spaces: Part A



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4.1 Vectors in ${old R}^n$

• Vectors in the plane:

a vector x in the plane is represented by a directed line segment with its initial point at the origin and its terminal point at (x_1, x_2) .



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Vector Spaces



• *n*-space: R^n

 $R^{1} = 1$ -space = set of all real number (x_{1}, x_{2}) $R^{2} = 2$ -space = set of all ordered pair of real numbers (x_{1}, x_{2}, x_{3}) $R^{3} = 3$ -space = set of all ordered triple of real numbers $(x_{1}, x_{2}, \dots, x_{n})$ \vdots $R^{n} = n$ -space = set of all ordered *n*-tuple of real numbers

Notes: An *n*-tuple (x₁, x₂, ..., x_n) can be viewed as
(1) <u>a point</u> in Rⁿ with the x_i's as its coordinates.
(2) <u>a vector</u> x in Rⁿ with the x_i's as its components.

<u>a vector</u> \boldsymbol{x} in \mathbb{R}^n will be represented also as $\boldsymbol{x} = (x_1, x_2, \dots, x_n)$

 x_1

 \mathcal{X}_n

 $\boldsymbol{x} = \begin{vmatrix} x_2 \\ \vdots \end{vmatrix}$



Equal:

$$\boldsymbol{u} = \boldsymbol{v}$$
 if and only if $u_1 = v_1, u_2 = v_2, \dots, u_n = v_n$

• Vector addition (the sum of *u* and *v*):

 $u + v = (u_1 + v_1, u_2 + v_2, \dots, u_n + v_n)$

- Scalar multiplication (the scalar multiple of *u* by *c*):

$$c\boldsymbol{u} = (cu_1, cu_2, \cdots, cu_n)$$

• Notes:

The sum of two vectors and the scalar multiple of a vector in \mathbb{R}^n are called the standard operations in \mathbb{R}^n .



• Negative:

$$-\boldsymbol{u} = (-u_1, -u_2, \cdots, -u_n)$$

Difference:

$$\boldsymbol{u} - \boldsymbol{v} = (u_1 - v_1, u_2 - v_2, \dots, u_n - v_n)$$

- Zero vector:
 - $\mathbf{0}=(0,\,0,\,\cdots,\,0)$
- Notes:

(1) The zero vector **0** in Rⁿ is called the additive identity in Rⁿ.
(2) The vector -v is called the additive inverse of v.



• Ex 3:

Let u = (-1, 0, 1) and v = (2, -1, 5) in \mathbb{R}^3 . Perform each vector operation:

(a) u + v (b) 2u (c) v - 2u

Sol:

(a)
$$u + v = (-1, 0, 1) + (2, -1, 5) = (1, -1, 6)$$

(b) $2u = 2(-1, 0, 1) = (-2, 0, 2)$
(c) $v - 2u = (2, -1, 5) - (-2, 0, 2) = (4, -1, 3)$





• Theorem 4.1: (Properties of vector addition and scalar multiplication)

Let u, v, and w be vectors in \mathbb{R}^n , and let c and d be scalars

(1) $\boldsymbol{u} + \boldsymbol{v}$ is a vector in \mathbb{R}^n (2) $\boldsymbol{u} + \boldsymbol{v} = \boldsymbol{v} + \boldsymbol{u}$ (3) (u + v) + w = u + (v + w)(4) u + 0 = u(5) u + (-u) = 0(6) cu is a vector in \mathbb{R}^n (7) $c(\boldsymbol{u} + \boldsymbol{v}) = c\boldsymbol{u} + c\boldsymbol{v}$ (8) (c+d)u = cu + du(9) $c(d\boldsymbol{u}) = (cd)\boldsymbol{u}$ (10) 1(u) = u

Closure under addition Commutative property of addition Associative property of addition Additive identity property Additive inverse property Closure under scalar multiplication Distributive property Distributive property Associative property of multiplication Multiplicative identity property



• Ex 4: (Vector operations in R^4)

Let u = (2, -1, 5, 0), v = (4, 3, 1, -1) and w = (-6, 2, 0, 3) be vectors in \mathbb{R}^4 . Solve x for each of the following:

(a)
$$x = 2u - (v + 3w)$$

(b) $3(x + w) = 2u - v + x$

Sol: (a)
$$x = 2u - (v + 3w)$$

= $2u - v - 3w$
= $(4, -2, 10, 0) - (4, 3, 1, -1) - (-18, 6, 0, 9)$
= $(4 - 4 + 18, -2 - 3 - 6, 10 - 1 - 0, 0 + 1 - 9)$
= $(18, -11, 9, -8)$



(b)
$$3(x + w) = 2u - v + x$$

 $3x + 3w = 2u - v + x$
 $3x - x = 2u - v - 3w$
 $2x = 2u - v - 3w$
 $x = u - \frac{1}{2}v - \frac{3}{2}w$
 $= (2, 1, 5, 0) + (-2, \frac{-3}{2}, \frac{-1}{2}, \frac{1}{2}) + (9, -3, 0, \frac{-9}{2})$
 $= (9, \frac{-11}{2}, \frac{9}{2}, -4)$



- Theorem 4.2: (Properties of additive identity and additive inverse)
- Let v be a vector in \mathbb{R}^n , and c be a scalars. Then the properties below are true: (1) The additive identity is unique. That is, if u + v = v, then u = 0(2) The additive inverse of v is unique. That is, if v + u = 0, then u = -v(3) 0v = 0(4) $c \mathbf{0} = \mathbf{0}$ (5) If cv = 0, then c = 0 or v = 0(6) - (-v) = v



• Linear combination:

The vector x is called a linear combination of $v_1, v_2, ..., v_n$ if it can be expressed in the form $x = c_1v_1 + c_2v_2 + \cdots + c_nv_n$ $c_1, c_2, ..., c_n$: scalars

• Ex 5: Given x = (-1, -2, -2), u = (0, 1, 4), v = (-1, 1, 2), and w = (3, 1, 2) in \mathbb{R}^3 , find a, b, and c such that x = au + bv + cw.

Sol:

$$-b + 3c = -1$$

$$a + b + c = -2$$

$$4a + 2b + 2c = -2$$

$$\Rightarrow a = 1, b = -2, c = -1$$
Thus $x = u - 2v - w$



4.2 Vector Spaces

• Vector spaces:

Let V be a set on which two operations (vector addition and scalar multiplication) are defined. If the following axioms are satisfied for every u, v, and w in V and every scalar c and d, then V is called a vector space.

Addition:

(1) $u + v$ is in V	Closure under addition
(2) $u + v = v + u$	Commutative property
(3) $u + (v + w) = (u + v) + w$	Associative property
(4) V has a zero vector 0 : for every \boldsymbol{u} in V,	u + 0 = u Additive identity
(5) For every \boldsymbol{u} in V , there is a vector in V	denoted by $-u$: $u + (-u) = 0$ Scalar identity



Scalar multiplication:

(6) cu is a vector in V (7) c(u + v) = cu + cv(8) (c + d)u = cu + du(9) c(du) = (cd)u(10) 1(u) = u Closure under scalar multiplication Distributive property Distributive property Associative property Scalar identity

• Notes:

(1) A vector space (V, +, .) consists of <u>four entities</u>:
a nonempty set V of vectors, a set of scalars, and two operations (+, .)
(2) V = {0} zero vector space



- Examples of vector spaces:
 - (1) *n*-tuple space: $V = R^n$

 $(u_1, u_2, \dots, u_n) + (v_1, v_2, \dots, v_n) = (u_1 + v_1, u_2 + v_2, \dots, u_n + v_n)$ vector addition $k(u_1, u_2, \dots, u_n) = (ku_1, ku_2, \dots, ku_n)$ scalar multiplication

(2) Matrix space: $V = M_{mxn}$ (the set of all $m \times n$ matrices with real values) Ex: (m = n = 2)

$$\begin{bmatrix} u_{11} & u_{12} \\ u_{21} & u_{22} \end{bmatrix} + \begin{bmatrix} v_{11} & v_{12} \\ v_{21} & v_{22} \end{bmatrix} = \begin{bmatrix} u_{11} + v_{11} & u_{12} + v_{12} \\ u_{21} + v_{21} & u_{22} + v_{22} \end{bmatrix}$$
 vector addition
$$k \begin{bmatrix} u_{11} & u_{12} \\ u_{21} & u_{22} \end{bmatrix} = \begin{bmatrix} ku_{11} & ku_{12} \\ ku_{21} & ku_{22} \end{bmatrix}$$
 scalar multiplication



(3) *n*-th degree polynomial space: $V = P_n(x)$ (the set of all real polynomials of degree *n* or less) $p(x) + q(x) = (a_0 + b_0) + (a_1 + b_1)x + \dots + (a_n + b_n)x^n$ $kp(x) = ka_0 + ka_1x + \dots + ka_nx^n$

(4) Function space: $V = c(-\infty, \infty)$ (the set of all real functions) (f + g)(x) = f(x) + g(x)(kf)(x) = kf(x)



• Theorem 4.3: (Properties of scalar multiplication)

Let v any element of a vector space V, and let c be any scalars. Then the following properties are true:

(1) 0v = 0

(2) $c \mathbf{0} = \mathbf{0}$

(3) If cv = 0, then c = 0 or v = 0

(4) (-1)v = -v



4.3 Subspaces of Vector Spaces

- Subspace:
 - $\begin{array}{l} (V, +, .) & : \text{a vector space} \\ \hline W \neq \emptyset \\ W \subseteq V \end{array} : \text{a nonempty subset} \end{array}$
 - (W, +, .): a vector space (under the operations of addition and scalar multiplication defined in V)
 - \Rightarrow W is a subspace of V
- Trivial subspace: Every vector space V has at least two subspaces
 - (1) Zero vector space $\{0\}$ is a subspace of V.
 - (2) V is a subspace of V.



• Theorem 4.4: (Test for a subspace)

If W is a <u>nonempty subset</u> of a vector space V, then W is a subspace of V if and only if the following conditions hold:

(1) If u and v are in W, then u + v is in W.

(2) If u is in W and c is any scalar, then cu is in W.

• Notes:

(1) If u and v are in W, c and d are any scalars, then cu + dv is in W. \Rightarrow W is a subspace of V

(2) If W is a subspace of a vector space V, then W contains the zero vector $\mathbf{0}$ of V



• Ex 1: Subspace of R^2





- (1) $\{\mathbf{0}\}$ $\mathbf{0} = (0, 0, 0)$
- (2) Lines through the origin
- (3) Planes through the origin
- (4) R^3



• Ex 4: (Determining subspaces of R^2)

Which of the following two subsets is a subspace of R^2 ?

(a) The set of points on the line given by x + 2y = 0. Yes (b) The set of points on the line given by x + 2y = 1. No

Theorem 4.5: (The intersection of two subspaces is a subspace)
If V and W are both subspaces of a vector space U, then the intersection of V and W (denoted by V ∩ W) is also a subspace of U.



4.4 Spanning Sets and Linear Independence

Linear combination:

A vector v in a vector space V is called a linear combination of the vectors u_1 , u_2, \ldots, u_k in V if v can be written in the form

$$\boldsymbol{v} = c_1 \boldsymbol{u_1} + c_2 \boldsymbol{u_2} + \dots + c_k \boldsymbol{u_k}$$
 c_1, c_2, \dots, c_k : scalars

• Ex 1: (Finding a linear combination)

$$v_1 = (1, 2, 3), \quad v_2 = (0, 1, 2), \quad v_3 = (-1, 0, 1)$$

Prove (a) w = (1, 1, 1) is a linear combination of v_1, v_2, v_3

(b) w = (1, -2, 2) is not a linear combination of v_1, v_2, v_3



Sol: (a)
$$w = c_1 v_1 + c_2 v_2 + c_3 v_3$$

(1, 1, 1) $= c_1(1, 2, 3) + c_2(0, 1, 2) + c_3(-1, 0, 1)$
 $= (c_1 - c_3, 2c_1 + c_2, 2c_2 + c_3)$
 $c_1 - c_3 = 1$
 $\Rightarrow 2c_1 + c_2 = 1$
 $3c_1 + 2c_2 + c_3 = 1$
 $\Rightarrow \begin{bmatrix} 1 & 0 & -1 & | & 1 \\ 2 & 1 & 0 & | & 1 \\ 3 & 2 & 1 & | & 1 \end{bmatrix}$ Gauss-Jordan Elimination $\begin{bmatrix} 1 & 0 & -1 & | & 1 \\ 0 & 1 & 2 & | & -1 \\ 0 & 0 & 0 & | & 0 \end{bmatrix}$
 $\Rightarrow c_1 = 1 + t, c_2 = -1 - 2t, c_3 = t$ (this system has infinitely many solutions $t = 1 \Rightarrow w = 2v_1 - 3v_2 + v_3$



(b)
$$\mathbf{w} = c_1 \mathbf{v_1} + c_2 \mathbf{v_2} + c_3 \mathbf{v_3}$$

$$\Rightarrow \begin{bmatrix} 1 & 0 & -1 & | & 1 \\ 2 & 1 & 0 & | & -2 \\ 3 & 2 & 1 & | & 2 \end{bmatrix} \xrightarrow{\text{Gauss-Jordan Elimination}} \begin{bmatrix} 1 & 0 & -1 & | & 1 \\ 0 & 1 & 2 & | & -4 \\ 0 & 0 & 0 & | & 7 \end{bmatrix}$$

 \Rightarrow this system has no solution (0 \neq 7)

 $\Rightarrow w \neq c_1 v_1 + c_2 v_2 + c_3 v_3$

• A spanning set of a vector space:

If every vector in a given vector space can be written as a linear combination of vectors in a given set *S*, then *S* is called a spanning set of the vector space.



• Ex 2: (A spanning set for R^3)

The set $S = \{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$ spans R^3 because any vector $u = (u_1, u_2, u_3)$ in R^3 can be written as

$$u = u_1(1, 0, 0) + u_2(0, 1, 0) + u_3(0, 0, 1) = (u_1, u_2, u_3)$$

• The span of a set: span (S)

If $S = \{v_1, v_2, ..., v_k\}$ is a set of vectors in a vector space V, then the span of S is the set of all linear combinations of the vectors in S,

$$\operatorname{span}(S) = \left\{ c_1 \boldsymbol{v_1} + c_2 \boldsymbol{v_2} + \dots + c_k \boldsymbol{v_k} \mid \forall c_i \in R \right\}$$

(the set of all linear combinations of the vectors in S)



- Linear Independent (L.I.) and Linear Dependent (L.D.):
 - $S = \{v_1, v_2, \dots, v_k\}$ is a set of vectors in a vector space V,
 - $c_1 \boldsymbol{v_1} + c_2 \boldsymbol{v_2} + \dots + c_k \boldsymbol{v_k} = 0$
 - (1) If the equation has only the trivial solution $(c_1 = c_2 = \dots c_k = 0)$, then S is called linearly independent.
 - (2) If the equation has a non trivial solution (i.e. not all zeros), then S is called linearly dependent.



Notes

- (1) Ø is linearly independent.
- (2) $\mathbf{0} \in S \Rightarrow S$ is linearly dependent.
- (3) $v \neq 0 \Rightarrow \{v\}$ is linearly independent.

(4) $S_1 \subseteq S_2$

- S_1 is linearly dependent \Rightarrow S_2 is linearly dependent
- S_2 is linearly independent \Rightarrow S_1 is linearly independent



• Ex 3: (Testing for linearly independent)

Determine whether the following set of vectors in R^3 is L.I. or L.D.

$$S = \{(1, 2, 3), (0, 1, 2), (-2, 0, 1)\}$$

$$v_{1} v_{2} v_{3}$$
Sol:
$$c_{1}v_{1} + c_{2}v_{2} + c_{3}v_{3} = 0 \implies 2c_{1} + c_{2} = 0$$

$$3c_{1} + 2c_{2} + c_{3} = 0$$

$$\Rightarrow \begin{bmatrix} 1 & 0 & -2 & | & 0 \\ 2 & 1 & 0 & | & 0 \\ 3 & 2 & 1 & | & 0 \end{bmatrix} \xrightarrow{\text{Gauss-Jordan Elimination}} \begin{bmatrix} 1 & 0 & 0 & | & 0 \\ 0 & 1 & 0 & | & 0 \\ 0 & 0 & 1 & | & 0 \end{bmatrix}$$

$$\Rightarrow c_{1} = c_{2} = c_{3} = 0 \text{ (only the trivial solution)} \Rightarrow S \text{ is linearly independent}$$



Independence of two vectors:

Two vectors u and v in a vector space V are linearly dependent if and only if one is a scalar multiple of the other.

• Ex 4: (Testing for linear dependent of 2 Vectors)

(1) $S = \{v_1, v_2\} = \{(1, 2, 0), (-2, 2, 1)\}$ is L.I. because v_1 and v_2 are not scalar multiples of each other.

(2) $S = \{v_1, v_2\} = \{(4, -4, -2), (-2, 2, 1)\}$ is L.D. because $v_1 = -2v_2$