Linear Algebra: Vectors

1. Vector Spaces

• Scalars / Numbers:

In defining vector spaces, we consider the *field* to be the set of reals: \Re . The elements of \Re are called *scalars* or *numbers*.

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- An *n*-vector x is an ordered set of n numbers $(x_1, x_2, ..., x_n)$.
 - 'Ordered' \Rightarrow the ordering of the numbers matters.
 - The set of all *n*-vectors is called *n*-space, denoted by \Re^n .

• Some Special Vectors:

- Sum Vector: The vector all of whose coordinates are 1.
- Null Vector: The vector all of whose coordinates are 0.
- **Unit Vectors:** The *i*-th unit vector is the vector whose *i*-th coordinate is 1, and whose other coordinates are 0.
 - We denote the *i*-th unit vector by e^i .

• Vector Operations:

- Two *n*-vectors, $x = (x_1, x_2, ..., x_n)$ and $y = (y_1, y_2, ..., y_n)$, are said to be *equal* (written x = y) if $x_i = y_i$, for i = 1, 2, ..., n.
- Addition: If x and y are *n*-vectors, their sum x + y is the *n*-vector whose *i*-th element is $x_i + y_i$, i = 1, 2, ..., n.
- Scalar Multiplication: If x is an *n*-vector and λ is a number, the product λx is the vector whose *i*-th element is λx_i , i = 1, 2, ..., n.
- Given these two definitions, a number of properties follow immediately.

• Properties of Vector Addition:

- Commutative Law: x + y = y + x.
- Associative Law: (x + y) + z = x + (y + z).
- Law of Subtraction: For every x and y, there is z such that x + z = y.

• Properties of Scalar Multiplication:

- Vector Distributive Law: $\lambda (x + y) = \lambda x + \lambda y$.
- Scalar Distributive Law: $(\lambda + \mu) x = \lambda x + \mu x$.
- Scalar Associative Law: $\lambda(\mu x) = (\lambda \mu)x$.
- Identity Law: 1x = x (here 1 is the scalar 'one').

• Vector Spaces:

The properties listed above may be taken as axioms for an abstract algebraic system. Such systems are called *vector spaces*.

- The vector space that we will study consists of
 - the field $\Re,$
 - the *n*-space \Re^n ,
 - the operations of addition and scalar multiplication.
- We will, as a shorthand, refer to this vector space by \Re^n .

2. Linear Dependence of Vectors

• A set of vectors $x^1, x^2, ..., x^m$ (here $x^i = (x_1^i, x_2^i, ..., x_n^i) \in \Re^n$ is a vector) is *linearly* dependent if there exist numbers $\lambda_1, \lambda_2, ..., \lambda_m$, not all zero, such that

 $\lambda_1 x^1 + \lambda_2 x^2 + \ldots + \lambda_m x^m = 0$ (the zero vector in \Re^n).

- A set of vectors is called *linearly independent* if the vectors are *not* linearly dependent.
- #1. Are the following two vectors in \Re^2 , $x^1 = (1,2)$ and $x^2 = (-3,-6)$, linearly dependent?
- #2. Are the following two vectors in \Re^2 , $x^1 = (1, 2)$ and $x^2 = (0, 1)$, linearly dependent?
- #3. Are the following three vectors in \Re^2 , $x^1 = (1, 1)$, $x^2 = (2, 3)$ and $x^3 = (3, 7)$ linearly dependent?
- #4. Show that the unit vectors in \Re^2 , $e^1 = (1, 0)$ and $e^2 = (0, 1)$ are linearly independent.

• A vector $y \in \Re^n$ can be written as a *linear combination* of the set of vectors $x^1, x^2, ..., x^m$ in \Re^n if there are numbers $\lambda_1, \lambda_2, ..., \lambda_m$ such that

$$y = \lambda_1 x^1 + \lambda_2 x^2 + \dots + \lambda_m x^m$$

#5. Express y = (0,0) as a linear combination of $x^1 = (1,2)$ and $x^2 = (1,0)$.

#6. Express y = (3, 2) as a linear combination of $x^1 = (1, 0)$ and $x^2 = (0, 1)$.

#7. Show that any vector in \Re^n can be written as a linear combination of the unit vectors $e^1, e^2, ..., e^n$ in \Re^n .

• Theorem 1 (Fundamental Theorem on Vector Spaces):

If each of the (m+1) vectors $y^0, y^1, ..., y^m$ in \Re^n can be expressed as a linear combination of the *m* vectors $x^1, x^2, ..., x^m$ in \Re^n , then the vectors $y^0, y^1, ..., y^m$ are linearly dependent.

- Proof: See Gale (1960).

- Corollary 1: Any set of (n+1) vectors in \Re^n are linearly dependent.
 - Proof: To be discussed in class.
- Corollary 2: Any system of n homogenous linear equations in (n+1) unknowns,

$$a_{10}x_0 + a_{11}x_1 + \dots + a_{1n}x_n = 0,$$

$$a_{20}x_0 + a_{21}x_1 + \dots + a_{2n}x_n = 0,$$

$$\vdots$$

$$a_{n0}x_0 + a_{n1}x_1 + \dots + a_{nn}x_n = 0,$$

has a non-zero solution.

– Proof: To be discussed in class.

3. Rank and Basis

- **Rank:** Let *S* be a subset of the vector space \Re^n . The *rank* of *S* is the maximum number of linearly independent vectors which can be chosen from *S*.
 - Note that $rank(S) \le n$ since (n + 1) vectors in S are linearly dependent by Corollary 1.
- **Basis:** Let *S* be a subset of the vector space \Re^n . If *r* is the rank of *S*, then any set of *r* linearly independent vectors in *S* is called *a basis* of *S*.
- #8. Consider $S_1 = \{(1,0), (0,1)\}$ in \Re^2 . What is the rank of S_1 ?
- #9. Consider $S_2 = \{(0,0)\}$ in \Re^2 . What is the rank of S_2 ?
- #10. (a) Are the following three vectors in \Re^3 , x = (1, 0, 0), y = (1, 1, 0) and z = (1, 1, 1) linearly dependent?

(b) Consider $S_3 = \{(x_1, x_2, x_3) \in \Re^3 : x_1 = 1\}$. What is the rank of S_3 ? Find out a basis of S_3 .

- Corollary 3: \Re^n has rank n.
 - Proof: To be discussed in class.

• Theorem 2: (Basis Theorem):

Suppose $x^1, x^2, ..., x^m$ are linearly independent vectors in the set S (in \Re^n).

(i) If every vector y in S can be expressed as a linear combination of $x^1, x^2, ..., x^m$, then $(x^1, x^2, ..., x^m)$ is a basis of S.

(ii) If $(x^1, x^2, ..., x^m)$ is a basis of *S*, then every vector *y* in *S* can be expressed as a linear combination of $x^1, x^2, ..., x^m$.

– Proof: To be discussed in class.

4. Inner Product and Norm

• Inner Product: If x and y are two vectors in \Re^n , then their inner product is

$$xy = x_1y_1 + x_2y_2 + \dots + x_ny_n.$$

- Note that inner product is a scalar (not a vector).
- $xe^i = x_i, i = 1, 2, ..., n$, where e^i is the *i*-th unit vector.

Properties of Inner Product:

- Commutative Law: xy = yx
- Mixed Associative Law: $(\lambda x) y = \lambda (xy) [\lambda \text{ is a scalar}]$
- Distributive Law: (x + y) z = xz + yz
- $-x^2 \equiv xx = 0$ if and only if x = 0.

• Norm of a Vector: The (Euclidean) norm of a vector x in \Re^n is defined as

$$||x|| = (xx)^{\frac{1}{2}} = (x_1^2 + x_2^2 + \dots + x_n^2)^{\frac{1}{2}}.$$

• Properties of Norm:

- ||x|| = 0 if and only if x = 0
- $\left\| \lambda x \right\| = \left| \lambda \right| . \left\| x \right\|$
- $||x + y|| \le ||x|| + ||y||$ (Triangle Inequality for norms)
- If xy = 0, then $||x + y||^2 = ||x||^2 + ||y||^2$ (Pythagoras Theorem).
- Orthogonal Vectors: Two vectors x and y are called *orthogonal* if their inner product is zero, that is, if xy = 0.
- Orthonormal Vectors: Two vectors x and y are called *orthonormal* if

(a) x and y are orthogonal, and (b) ||x|| = ||y|| = 1.

References

• Must read the following sections from the textbook:

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Sections 10.1 – 10.4 (pages 199 – 214),
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Sections 11.1 – 11.3 (pages 237 – 249).
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- Much of this material is standard in texts on linear algebra, like
- 1. Hadley, G., Linear Algebra, Massachusetts: Addison-Wesley, 1964 (chapter 2),
- 2. Hohn, Franz E., *Elementary Matrix Algebra*, New Delhi: Amerind, 1971 (chapters 4, 5).
- A good exposition can also be found in
- 3. Gale, David, *The Theory of Linear Economic Models*, New York: McGraw-Hill, 1960 (chapter 2).
 - You will find a proof of the fundamental theorem on vector spaces (by using mathematical induction) in Gale's book.