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1 Vector Spaces

A vector space V is a set equipped with 2 operations:

- Addition: given $v, w \in V, v + w \in V$
- Scalar Multiplication: given $v \in V$ and $c \in \mathbb{R}$, $c \cdot v \in V$

These 2 operations are required to satisfy the following axioms $\forall u, v, w \in V$ and all scalars $c, d \in \mathbb{R}$:

- Commutativity of Addition: v + w = w + v
- Associativity of Addition: (u+v)+w=u+(v+w)
- Additive Identity: \exists a zero element $\mathbf{0} \in V \ni v + \mathbf{0} = v = \mathbf{0} + v$
- Additive Inverse: $\forall v \in V, \exists (-v) \in V \ni v + (-v) = \mathbf{0} = (-v) + v$
- Distribuvity: $(c+d) \cdot v = cv + dv$ and $c \cdot (v+w) = cv + cw$
- Associativity of Scalar Multiplication: $c \cdot (dv) = (cd) \cdot v$
- Unit for Scalar Multiplication: the scalar $1 \in \mathbb{R}$ satisfies 1v = v

1.1 Subspace

A subspace W of a vector space V is a subset $W \subset V$ and is a vector space on its own.

2 Inner Products

The most basic form of inner products is the inner product:

$$\langle v; w \rangle = v \cdot w = v_1 w_1 + v_2 w_2 + \dots = \sum_{i=1}^{n} v_i w_i$$

The Euclidean norm or length of a vector v is given by:

$$||v|| = \sqrt{v \cdot v} = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2}$$

Definition: An inner product on the vector space V is a pairing that takes two vectors $v, w \in V$ and produces $\langle v; w \rangle \in \mathbb{R}$ which is required to satisfy the following axioms, given $u, v, w \in V$ and $c, d, k \in \mathbb{R}$:

• Additivity:

$$\langle u + v; w \rangle = \langle u + v \rangle + \langle u + w \rangle$$

• Homoneneity:

$$\langle ku + v; w \rangle = k \langle u + v; w \rangle$$

• Bilinearity:

$$\langle c \cdot u + d \cdot v; w \rangle = c \langle u; v \rangle + d \langle v; w \rangle$$

• Symmetry:

$$\langle v; w \rangle = \langle w; v \rangle$$

• Positivity:

$$\langle v; v \rangle > 0$$
 whenever $v \neq 0$ while $\langle 0; 0 \rangle = 0$

3 The Cauchy-Schwarz Inequality

Every inner product satisfies the Cauchy-Schwarz inequality:

$$v \cdot w = ||v|| \ ||w|| \cos \theta$$
$$|\cos \theta| \le 1$$
$$v \cdot w \le ||v|| \ ||w||$$

4 Orthogonal Vectors

Definition: Two elements $v, w \in V$ of an inner product space V are called orthogonal if $\langle v; w \rangle = 0$. Note that the property of orthogonality depends on which type of inner product is used.

4.1 Orthogonal Set

Definition: A non-empty set in \mathbb{R}^n is called an orthogonal set if all pairs of distinct vectors in the set are orthogonal.

5 The Triangle Inequality

Given $u, v \in V$ and any scalar $k \in \mathbb{R}$, then $||u + v|| \le ||u|| + ||v||$.

6 Parallelogram Equation for Vectors

Given
$$u, v \in \mathbb{R}^n$$
, then $||u + v||^2 + ||u - v||^2 = 2(||u||^2 + ||v||^2)$.

7 Norms

Definition: A norm on the vector space V assigns a real number ||v|| to each vector $v \in V$ subject to the axioms of Positivity, Homogeneity and Triangle Inequality.

The 1st norm of a vector $v = (v_1, v_2, ..., v_n)^T$ is given by the sum of the absolute values of its entries:

$$||v||_1 = |v_1| + |v_2| + \dots + |v_n|$$

The max or ∞ norm is equal to its maximal entry in absolute values:

$$||v||_{\infty} = Sup(|v_1|, |v_2|, ..., |v_n|)$$

In general, the p-norm is given by:

$$||v||_p = \sqrt[p]{\sum_{i=1}^n |v_i|^p}$$

The triangle inequality for norms:

$$\sqrt[p]{\sum_{i=1}^{n} |v_i + w_i|^p} = \sqrt[p]{\sum_{i=1}^{n} |v_i|^p} + \sqrt[p]{\sum_{i=1}^{n} |w_i|^p}$$

8 Unit Vectors

If V is a fixed normed vector space, the elements $u \in V$ with unit norm ||u|| = 1 are known as unit vectors. If v is any non-zero vector, to obtain a unit vector u parallel to v:

$$u = \frac{v}{||v||}$$

9 Orthogonal Bases

Definition: A basis $u_1, u_2, ..., u_n$ of V is called orthogonal if $\langle u_i; u_j \rangle = 0 \ \forall \ i \neq j$. The basis is called orthogonormal if each vector has unit length that is $||u_i|| = 1 \ \forall \ i = 1, 2, ..., n$.

If $v_1, v_2, ..., v_n$ is an orthogonal basis of V, then the normalised vectors $u_i = v_i/||v_i|| \forall i = 1, 2, ..., n$ form an orthogonormal basis of V.

If $v_1, v_2, ..., v_n \in V$ are non-zero and mutually orthogonal $(\langle v_i; v_j \rangle = 0 \ \forall \ i \neq j)$, then they are linearly independent.

10 Vector Norms

Definition: Let $v:\mathbb{C}^n \to \mathbb{R}$. Then v is a vector norm if $\ \forall \ x,y \in \mathbb{C}$:

- Positive definite: $x \neq 0 \implies v(x) > 0$
- Homogeneity: $v(\alpha x) = |\alpha|v(x)$
- Obeys triangle inequality: $v(x+y) \le v(x) + v(y)$