

## Orthonormal Bases in Hilbert Space.

### Linear (Vector) Spaces.

**Definition 0.1** A linear space is a nonempty set  $L$  together with a mapping from  $L \times L$  into  $L$  called addition, denoted  $(x, y) \mapsto x + y$  and a mapping from the Cartesian product of either  $\mathbf{R}$  or  $\mathbf{C}$  with  $L$  into  $L$  called scalar multiplication, denoted  $(\alpha, x) \mapsto \alpha x$ , which satisfy the following properties.

(1) *Axioms of addition.*

(a)  $x + y = y + x$  (commutativity).

(b)  $(x + y) + z = x + (y + z)$  (associativity).

(c) There exists an element  $0 \in L$ , called the zero element, such that for all  $x \in L$ ,  $0 + x = x + 0 = x$ .

(d) For every  $x \in L$ , there is an element  $-x \in L$  such that  $x + (-x) = (-x) + x = 0$ .

(2) *Axioms of scalar multiplication.*

(a)  $\alpha(\beta x) = (\alpha\beta)x$ , for all scalars  $\alpha, \beta$  and all  $x \in L$ .

(b)  $1(x) = x$ .

(3) *Distributive Laws.*

(a)  $(\alpha + \beta)x = \alpha x + \beta x$ , for all scalars  $\alpha, \beta$  and all  $x \in L$ .

(b)  $\alpha(x + y) = \alpha x + \alpha y$ , for all scalars  $\alpha$  and all  $x, y \in L$ .

**Examples.** (a)  $L = \mathbf{R}^n$  or  $\mathbf{C}^n$  with addition and scalar multiplication defined component-wise in the usual way.

(b)  $L = C[0, 1]$ , the space of functions continuous on  $[0, 1]$ , with addition of functions defined by  $(f + g)(x) = f(x) + g(x)$ , and scalar multiplication defined by  $(\alpha f)(x) = \alpha f(x)$ .

(c) The following spaces are examples of *sequence spaces* and are in some sense the natural generalizations of  $\mathbf{R}^n$  as  $n \rightarrow \infty$ .

$$\begin{aligned}\ell^\infty &= \{\{x_n\}_{n=1}^\infty : \sup |x_n| < \infty\} \\ c_0 &= \{\{x_n\}_{n=1}^\infty : \lim x_n = 0\} \\ \ell^2 &= \{\{x_n\}_{n=1}^\infty : \sum |x_n|^2 < \infty\} \\ \ell^1 &= \{\{x_n\}_{n=1}^\infty : \sum |x_n| < \infty\} \\ f &= \{\{x_n\}_{n=1}^\infty : \exists N = N(x), x_n = 0 \forall n \geq N\},\end{aligned}$$

where  $\{x_n\} + \{y_n\} = \{x_n + y_n\}$  and  $\alpha\{x_n\} = \{\alpha x_n\}$ .

**Exercise 1.** Show that  $f \subset \ell^1 \subset \ell^2 \subset c_0 \subset \ell^\infty$  as sets.

**Definition 0.2** A subset  $L'$  of a linear space  $L$  is a subspace of  $L$  provided that for every  $x, y \in L'$  and scalars  $\alpha, \beta$ ,

$$\alpha x + \beta y \in L'.$$

In other words,  $L'$  is a subspace provided that it is a linear space in its own right with addition and scalar multiplication inherited from  $L$ .

Given a nonempty subset  $S$  of a linear space  $L$ , the linear span or span of  $S$ , denoted  $\text{span}(S)$ , is the set of all finite linear combinations of elements in  $S$ , that is

$$\text{span}(S) = \left\{ \sum_{i=1}^n \alpha_i x_i : \alpha_i \text{ scalars, } x_i \in S \right\}.$$

**Proposition 0.1** Given a nonempty subset  $S$  of a linear space  $L$ ,  $\text{span}(S)$  is a subspace of  $L$  and moreover,  $\text{span}(S)$  is the smallest subspace of  $L$  containing  $S$ , that is, if  $L'$  is any subspace containing  $S$ , then  $\text{span}(S) \subseteq L'$ .

**Proof: Exercise 2.**

### Normed Linear Spaces.

**Definition 0.3** A normed linear space is a pair  $(V, \|\cdot\|)$  where  $V$  is a linear space (over  $\mathbf{R}$  or  $\mathbf{C}$ ), and  $\|\cdot\|$  is a function  $\|\cdot\|: V \rightarrow \mathbf{R}$  called a **norm** which satisfies for all  $v, w \in V$  the following properties.

- (1)  $\|v\| \geq 0$ .
- (2)  $\|v\| = 0$  if and only if  $v = 0$ .
- (3)  $\|\alpha v\| = |\alpha| \|v\|$  for every scalar  $\alpha$ .
- (4)  $\|v + w\| \leq \|v\| + \|w\|$  (triangle inequality).

**Remark 0.1** Given a normed linear space  $V$ , we can define a metric on  $V$  by  $d(x, y) = \|x - y\|$ , making  $(V, d)$  a metric space.

### EXAMPLES.

1.  $(C[0, 1], \|\cdot\|_\infty)$ , where  $\|f\|_\infty = \sup_{x \in [0, 1]} |f(x)|$ .
2.  $(C[0, 1], \|\cdot\|_1)$ , where  $\|f\|_1 = \int_0^1 |f(x)| dx$ .
3.  $(C[0, 1], \|\cdot\|_2)$ , where  $\|f\|_2 = \left( \int_0^1 |f(x)|^2 dx \right)^{1/2}$ .
4.  $(\mathbf{R}^n, \|\cdot\|_2)$ , where  $\|x\|_2 = \left( \sum_{i=1}^n |x_i|^2 \right)^{1/2}$ .
5.  $(\mathbf{R}^n, \|\cdot\|_1)$ , where  $\|x\|_1 = \sum_{i=1}^n |x_i|$ .

6.  $(\mathbf{R}^n, \|\cdot\|_\infty)$ , where  $\|x\|_\infty = \max_{i=1, \dots, n} |x_i|$ .
7.  $\ell^1$  where  $\|\{x_n\}\|_1 = \sum_n |x_n|$ .
8.  $\ell^2$  where  $\|\{x_n\}\|_2 = (\sum_n |x_n|^2)^{1/2}$ .
9.  $\ell^\infty$  where  $\|\{x_n\}\|_\infty = \sup_n |x_n|$ .

**Exercise 3.** Show that for all  $f \in C[0, 1]$ ,

- (a)  $\|f\|_1 \leq \|f\|_\infty$
- (b)  $\|f\|_2 \leq \|f\|_\infty$
- (c)  $\|f\|_1 \leq \|f\|_2$  (Use the Cauchy-Schwarz inequality proved later.)

Show that in each case an opposite inequality does not hold in the sense that (for example) there is no number  $C > 0$  such that for all  $f \in C[0, 1]$ ,  $\|f\|_\infty \leq C \|f\|_1$ . In other words, for each number  $C$ , find a function  $f \in C[0, 1]$  (which will depend on  $C$ ) such that  $\|f\|_\infty > C \|f\|_1$ . Find corresponding examples for (b) and (c).

**Exercise 4.** Show that for all  $x \in \mathbf{C}^n$ ,

- (a)  $\|x\|_\infty \leq \|x\|_1$
- (b)  $\|x\|_\infty \leq \|x\|_2$
- (c)  $\|x\|_2 \leq \|x\|_1$
- (d)  $\|x\|_2 \leq \sqrt{n} \|x\|_\infty$
- (e)  $\|x\|_1 \leq n \|x\|_\infty$ .

Use these inequalities to show that the three norms are equivalent.

**Exercise 5.** Show that for all complex-valued sequences  $x = \{x_n\}_{n=1}^\infty$

- (a)  $\|x\|_\infty \leq \|x\|_1$
- (b)  $\|x\|_\infty \leq \|x\|_2$
- (c)  $\|x\|_2 \leq \|x\|_1$ .

Show that in each case an opposite inequality does not hold in the same sense as in Exercise 3. That is, show that (for example) given  $C > 0$  there is a sequence  $x = \{x_n\}$  (which will depend on  $C$ ) such that  $\|x\|_1 > C \|x\|_\infty$ .

## Inner Product and Hilbert Spaces.

**Definition 0.4** An inner product space is a pair  $(V, \langle \cdot, \cdot \rangle)$  where  $V$  is a vector space over  $\mathbf{C}$  or  $\mathbf{R}$  and where  $\langle \cdot, \cdot \rangle$  is a complex valued function

$$\langle \cdot, \cdot \rangle: V \times V \longrightarrow \mathbf{C}$$

called the inner product on  $V$  satisfying the following properties. For all  $x, y, z \in V$  and  $\alpha \in \mathbf{C}$ ,

(1)  $\langle x, x \rangle \geq 0$  and  $\langle x, x \rangle = 0$  if and only if  $x = 0$ .

(2)  $\langle \alpha x, y \rangle = \alpha \langle x, y \rangle$ .

(3)  $\langle x + z, y \rangle = \langle x, y \rangle + \langle z, y \rangle$ .

(4)  $\langle x, y \rangle = \overline{\langle y, x \rangle}$ .

EXAMPLES.

1.  $\mathbf{R}^n$  with  $\langle x, y \rangle = \sum_{i=1}^n x_i y_i$ .

2.  $V = C[0, 1]$  with  $\langle f, g \rangle = \int_0^1 f(x) \overline{g(x)} dx$ .

3.  $V = \ell^2$  with  $\langle x, y \rangle = \sum_{n=1}^{\infty} x_n \overline{y_n}$ .

**Theorem 0.1** Let  $V$  be an inner product space. Define the function

$$\| \cdot \|: V \longrightarrow \mathbf{R}$$

by  $\|x\| = \langle x, x \rangle^{1/2}$ . Then  $\| \cdot \|$  is a norm on  $V$  making  $V$  a normed linear space.

**Corollary 0.1** (Cauchy-Schwarz Inequality). For  $x, y \in V$ , an inner product space,

$$|\langle x, y \rangle| \leq \|x\| \|y\|.$$

**Exercise 6.** Prove the Cauchy-Schwarz inequality for a real inner product space. (Hint: For  $t \in \mathbf{R}$ , and  $x, y \in V$ , note that  $0 \leq \|x + ty\|^2 = \langle x + ty, x + ty \rangle$  defines a real-valued, non-negative quadratic function of  $t$ . What must such a quadratic satisfy?)

**Exercise 7.** Prove the triangle inequality for the norm defined through the inner product as in the Theorem. (Hint: Expand  $\|x + y\|^2$  as an inner product and use the C-S inequality).

**Definition 0.5** A Hilbert space is a complete inner product space. (Complete here means that every Cauchy sequence converges.)

EXAMPLES.

1.  $L^2[0, 1]$  with the inner product given by

$$\langle f, g \rangle = \int_0^1 f(x) \overline{g(x)} dx$$

is a Hilbert space.

2.  $\ell^2$  with the inner product given by

$$\langle \{x_n\}, \{y_n\} \rangle = \sum_n x_n \overline{y_n}$$

is a Hilbert space.

3.  $L^2(\mathbf{R})$  with the inner product given by

$$\langle f, g \rangle = \int_{-\infty}^{\infty} f(x) \overline{g(x)} dx$$

is a Hilbert space.

**Exercise 8.** Show that the sequence space  $f$  of finite sequences is *not* complete by finding a sequence in  $f$  which is Cauchy but which does not converge to anything in  $f$ .

### Orthonormal systems in Hilbert spaces.

**Definition 0.6** Let  $V$  be an inner product space. Two vectors  $x, y \in V$  are said to be **orthogonal** if  $\langle x, y \rangle = 0$ . We also write in this case  $x \perp y$ .

A collection of vectors  $\{x_\alpha\}_{\alpha \in A} \subseteq V$  is said to be an **orthonormal system** if  $\langle x_\alpha, x_\beta \rangle = 0$  for  $\alpha \neq \beta$  and if  $\langle x_\alpha, x_\alpha \rangle = 1$  for all  $\alpha \in A$ .

**Lemma 0.1** (Best Approximation Lemma). Let  $\{x_n\}_{n=1}^N$  be an orthonormal system in an inner product space  $V$  and let  $\{a_n\}_{n=1}^N$  be a finite sequence of scalars. Then for all  $x \in V$ ,

$$\left\| x - \sum_{n=1}^N a_n x_n \right\|^2 = \|x\|^2 - \sum_{n=1}^N |\langle x, x_n \rangle|^2 + \sum_{n=1}^N |a_n - \langle x, x_n \rangle|^2.$$

**Corollary 0.2** (1)  $\left\| \sum_{n=1}^N a_n x_n \right\|^2 = \sum_{n=1}^N |a_n|^2$ .

$$(2) \left\| x - \sum_{n=1}^N \langle x, x_n \rangle x_n \right\|^2 = \|x\|^2 - \sum_{n=1}^N |\langle x, x_n \rangle|^2.$$

$$(3) \text{ (Bessel's Inequality) } \sum_{n=1}^N |\langle x, x_n \rangle|^2 \leq \|x\|^2.$$

(4) If  $\{x_n\}_{n=1}^\infty$  is an orthonormal system, then for each  $x \in V$ , the series  $\sum_{n=1}^\infty |\langle x, x_n \rangle|^2$

converges and  $\sum_{n=1}^\infty |\langle x, x_n \rangle|^2 \leq \|x\|^2$ .

## Orthonormal bases in Hilbert spaces.

**Definition 0.7** A collection of vectors  $\{x_\alpha\}_{\alpha \in A}$  in a Hilbert space  $H$  is **complete** if  $\langle y, x_\alpha \rangle = 0$  for all  $\alpha \in A$  implies that  $y = 0$ .

An equivalent definition of completeness is the following.  $\{x_\alpha\}_{\alpha \in A}$  is complete in  $V$  if  $\text{span}\{x_\alpha\}$  is dense in  $V$ , that is, given  $y \in H$  and  $\epsilon > 0$ , there exists  $y' \in \text{span}\{x_\alpha\}$  such that  $\|y - y'\| < \epsilon$ . Another way to put this is that given  $y$ , every ball around  $y$  contains an element of  $\text{span}\{x_\alpha\}$ . The proof of this equivalence relies on a fundamental decomposition property of Hilbert spaces.

An **orthonormal basis** is a complete orthonormal system.

**Theorem 0.2** Let  $\{x_n\}_{n=1}^\infty$  be an orthonormal system in a Hilbert space  $H$ . Then the following are equivalent.

(1)  $\{x_n\}$  is complete.

(2) For all  $x \in H$ ,

$$x = \sum_{n=1}^{\infty} \langle x, x_n \rangle x_n,$$

where the sum converges unconditionally, that is, regardless of order.

(3)  $\|x\|^2 = \sum_{n=1}^{\infty} |\langle x, x_n \rangle|^2$ .

(4) For all  $x, y \in H$ ,

$$\langle x, y \rangle = \sum_{n=1}^{\infty} \langle x, x_n \rangle \overline{\langle y, x_n \rangle}.$$

## EXAMPLES.

1.  $\mathbf{R}^n$  or  $\mathbf{C}^n$ . In finite dimensional vector spaces we have the notion of *linear independence* and *dimension*. Specifically if the finite dimensional vector space  $X$  has dimension  $N$  and if  $V = \{v_k\}_{k=1}^N$  is an orthonormal system, then it is an orthonormal basis. Any collection of  $N$  linearly independent vectors can be orthogonalized via the Gram-Schmidt process into an orthonormal basis.
2.  $L^2[0, 1]$  is the space of all Lebesgue measurable functions on  $[0, 1]$ , square-integrable in the sense of Lebesgue. This space can be thought of as the completion of the incomplete normed linear space  $C[0, 1]$  with respect to the norm  $\|f\|_2^2 = \int_0^1 |f(x)|^2 dx$ . In other words, each element of  $L^2[0, 1]$  can be thought of as the limit (in the sense of the  $L^2$  norm) of a Cauchy sequence of continuous functions.

The trigonometric system  $\{e^{2\pi i n x}\}_{n=-\infty}^\infty$  is an orthonormal basis for  $L^2[0, 1]$ . The expansion of a function in this basis is called the *Fourier series* of that function.

Another example of an orthonormal basis for  $L^2[0, 1]$  are the Legendre polynomials which are obtained by taking the sequence of monomials  $\{1, x, x^2, \dots\}$  and applying the Gram-Schmidt orthogonalization process to it.

3.  $L^2(\mathbf{R})$  is the space of all Lebesgue measurable functions on  $\mathbf{R}$ , square-integrable in the sense of Lebesgue. This space can be thought of as the completion of the incomplete normed linear space  $C_c(\mathbf{R})$  of functions continuous on  $\mathbf{R}$  with compact support (equipped with the  $L^2$  norm), or as the completion of the incomplete normed linear space  $C_c^\infty(\mathbf{R})$  of all compactly supported, infinitely differentiable functions on  $\mathbf{R}$  equipped again with the  $L^2$  norm. This will be our setting for much of our discussion of wavelet bases. We will show how to construct orthonormal bases of this space with wavelets.

**Exercise 9.** Show that the trigonometric system  $\{e^{2\pi inx}\}_{n=-\infty}^\infty$  is an orthonormal system in  $L^2[0, 1]$ .

**Exercise 10.** Find the first four Legendre polynomials by applying the Gram-Schmidt process to the sequence  $\{1, x, x^2, \dots\}$ .