Hilbert spaces *

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1. Inner-product spaces

- **1.1 Definition** INNER PRODUCT SPACE. Let \mathcal{H} be a vector space on a set scalars S, where $S = \mathbb{R}$ or $S = \mathbb{C}$. An inner product on H is an application which associates to each pair of elements x and y in \mathcal{H} a complex number $\langle x, y \rangle$ such that, for all $x, y, z \in \mathcal{H}$,
- (a) $\langle x, y \rangle = \overline{\langle y, x \rangle}$;
- (b) $\langle x + y, z \rangle = \langle x, z \rangle + \langle y, z \rangle$;
- (c) $\langle \alpha x, y \rangle = \alpha \langle x, y \rangle$, for all $\alpha \in S$;
- (d) $\langle x, x \rangle \geq 0$;
- (e) $\langle x, x \rangle = 0$ if and only if x = 0.

If $\langle \cdot, \cdot \rangle$ is an inner product on \mathcal{H} , the pair $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ is called an inner product space, and the elements of \mathcal{H} are also called elements of the inner-product space $(\mathcal{H}, \langle \cdot, \cdot \rangle)$. When there is no ambiguity on the definition of the inner product, the inner-product space $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ may simply be denoted \mathcal{H} .

1.2 Remark If $S = \mathbb{R}$, condition (a) reduces to

(a)'
$$\langle x, y \rangle = \langle y, x \rangle$$
.

1.3 Definition NORM ASSOCIATED WITH AN INNER PRODUCT. The norm of an element x of an inner-product space $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ is defined by

$$||x|| = \sqrt{\langle x, x \rangle} \ . \tag{1.1}$$

1.4 Example EUCLIDEAN SPACE. \mathbb{R}^n with the usual scalar product

$$\langle x, y \rangle = \sum_{i=1}^{n} x_i y_i \tag{1.2}$$

where $x=(x_1,\ldots,x_x)'$ $y=(y_1,\ldots,y_x)'$, where $x_i\in\mathbb{R}$ and $y_j\in\mathbb{R}$ for all i and j, is an inner-product space whose norm is the usual Euclidean norm

$$||x|| = \left(\sum_{i=1}^{n} x_i^2\right)^{1/2} . \tag{1.3}$$

1.5 Example COMPLEX EUCLIDEAN SPACE. \mathbb{C}^n with the scalar product

$$\langle x, y \rangle = \sum_{i=1}^{n} x_i \overline{y}_i \tag{1.4}$$

where $x=(x_1,\ldots,x_n)'$ $y=(y_1,\ldots,y_n)'$, where $x_i\in\mathbb{C}$ and $y_j\in\mathbb{C}$ for all i and j, is an inner-product space. The associated norm is:

$$||x|| = \left(\sum_{i=1}^{n} |x_i|^2\right)^{1/2}.$$
 (1.5)

1.6 Example REAL-VALUED SQUARE INTEGRABLE RANDOM VARIABLES. The set $L^2 = L^2(\Omega, \mathcal{A}, P)$ of all the random variables $X : \Omega \to \mathbb{R}$ such that $E(X^2) < \infty$ with, for any $X, Y \in L^2$,

$$\langle X, Y \rangle = E(XY) \tag{1.6}$$

is an inner-product space. The associated norm is:

$$||X|| = [E(X^2)]^{1/2}$$
. (1.7)

1.7 Example COMPLEX-VALUED SQUARE INTEGRABLE RANDOM VARIABLES. The set $L^2_{\mathbb{C}}(\Omega, \mathcal{A}, P)$ of all the random variables $X: \Omega \to \mathbb{C}$ such that $E(|X|^2) < \infty$ with

$$\langle X, Y \rangle = E(X\overline{Y}) \tag{1.8}$$

is an inner-product space. The associated norm is:

$$||X|| = [E(|X|^2)]^{1/2}$$
. (1.9)

1.8 Proposition Cauchy-Schwarz inequality. Let $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ be an inner-product space. Then, for all $x, y \in \mathcal{H}$,

$$|\langle x, y \rangle| \le ||x|| \, ||y|| \tag{1.10}$$

and

$$|\langle x, y \rangle| = ||x|| \, ||y|| \iff x = \left(\frac{\langle x, y \rangle}{\langle y, y \rangle}\right) y .$$
 (1.11)

1.9 Definition ANGLE AND ORTHOGONALITY. Let x and y be two elements of an inner product space $(\mathcal{H}, \langle \cdot, \cdot \rangle)$. The angle between the vectors x and y is defined by

$$\theta = \cos^{-1}[\langle x, y \rangle / (\|x\| \|y\|)]$$
 (1.12)

x and y are said to be orthogonal (denoted $x \perp y$) if and only if

$$\langle x, y \rangle = 0. (1.13)$$

1.10 Proposition PROPERTIES OF THE NORM. Let $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ be an inner-product space with the norm $||x|| = \sqrt{\langle x, x \rangle}$. Then the following properties hold for all $x, y \in \mathcal{H}$:

(a)
$$||x + y||^2 = ||x||^2 + ||y||^2 + \langle x, y \rangle + \langle y, x \rangle$$
;

- (b) $||x + y|| \le ||x|| + ||y||$; (Triangle inequality)
- (c) $\|\alpha x\| \leq |\alpha| \|x\|$, for all $\alpha \in \mathbb{C}$;
- $(d) ||x|| \ge 0;$
- (e) ||x|| = 0 if and only if x = 0;

$$(f) ||x+y||^2 + ||x-y||^2 = 2||x||^2 + 2||y||^2.$$
 (Parallelogram law)

1.11 Proposition Convergence in Norm. Let $\{x_n : n = 1, 2, ...\}$ be a sequence of elements of an iner product space $(\mathcal{H}, \langle \cdot, \cdot \rangle)$. We say that x_n converges to $x \in \mathcal{H}$ if and only if

$$||x_n - x|| \underset{n \to \infty}{\longrightarrow} 0. \tag{1.14}$$

1.12 Proposition CONTINUITY OF INNER PRODUCT. Let $\{x_n: n=1,2,...\}$ and $\{y_n: n=1,2,...\}$ be two sequences of elements of an inner-product space $(\mathcal{H},\langle\cdot,\cdot\rangle)$ such that $\|x_n-x\|\underset{n\to\infty}{\longrightarrow} 0$ and $\|y_n-y\|\underset{n\to\infty}{\longrightarrow} 0$, where $x,y\in\mathcal{H}$. Then

$$||x_n|| \underset{n \to \infty}{\longrightarrow} ||x|| \tag{1.15}$$

and

$$\langle x_n, y_n \rangle \xrightarrow[n \to \infty]{} \langle x, y \rangle$$
 (1.16)

1.13 Example MEAN SQUARE CONVERGENCE. Let $\{X_n : n = 1, 2, ...\}$ a sequence of real random variables in $L^2 \equiv L^2(\Omega, \mathcal{A}, P)$. If we define $\langle X, Y \rangle = E(XY)$, then the convergence in norm represents mean-square convrgence:

$$||X_n - X|| \underset{n \to \infty}{\longrightarrow} 0 \Leftrightarrow E[(X_n - X)^2] \underset{n \to \infty}{\longrightarrow} 0.$$
 (1.17)

2. Hilbert spaces

2.1 Definition CAUCHY SEQUENCE. Let $\{x_n : n = 1, 2, ...\}$ be a sequence of elements of an inner-product space $(\mathcal{H}, \langle \cdot, \cdot \rangle)$. Then x_n is a Cauchy sequence of $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ if and only if

$$||x_n - x_m|| \underset{n \to \infty}{\longrightarrow} 0. \tag{2.1}$$

2.2 Definition HILBERT SPACE. Let $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ an inner-product space. Then $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ is a Hilbert space if and only if every Cauchy sequence of $(\mathcal{H}, \langle \cdot, \cdot \rangle)$ converges in norm to an element of \mathcal{H} ,

$$||x_n - x_m|| \underset{m, n \to \infty}{\longrightarrow} 0 \Rightarrow \exists x \in \mathcal{H} \text{ such that } ||x_n - x|| \underset{n \to \infty}{\longrightarrow} 0.$$
 (2.2)

- **2.3 Proposition** NORM CONVERGENCE AND CAUCHY CRITERION. Let $\{x_n : n = 1, 2, ...\}$ be a sequence of elements of a Hilbert space $(\mathcal{H}, \langle \cdot, \cdot \rangle)$. Then x_n converges in norm if and only if x_n is a Cauchy sequence.
- **2.4 Theorem** COMPLETENESS OF $L^2(\Omega, \mathcal{A}, P)$. The inner-product space $L^2(\Omega, \mathcal{A}, P)$ with $\langle X, Y \rangle = E(XY)$ is complete and thus constitutes a Hilbert space.

PROOF. See Brockwell and Davis (1991, Section 2.10). ■

3. Projection theorems

3.1 Definition CLOSED SUBSPACE. A linear subspace \mathcal{M} of a Hilbert space \mathcal{H} is a closed subspace of \mathcal{H} if and only if \mathcal{M} contains all its limit points, i.e.

$$x_n \in \mathcal{M}, \ \forall n, \ \text{and} \ \|x_n - x\| \underset{n \to \infty}{\longrightarrow} 0 \ \Rightarrow \ x \in \mathcal{M}.$$
 (3.1)

3.2 Definition ORTHOGONAL COMPLEMENT. The orthogonal complement \mathcal{M}^{\perp} of a subset \mathcal{M} of \mathcal{H} is the set of all $x \in \mathcal{H}$ which are orthogonal to all the elements of \mathcal{M} , i.e.

$$x \in \mathcal{M}^{\perp} \Leftrightarrow \langle x, y \rangle = 0 , \forall y \in \mathcal{M} .$$
 (3.2)

- **3.3 Proposition** CLOSURE OF ORTHOGONAL COMPLEMENT. If \mathcal{M} is any subset of a Hilbert space \mathcal{H} , then \mathcal{M}^{\perp} is a closed subspace of \mathcal{H} .
- **3.4 Theorem** PROJECTION THEOREM. Let \mathcal{M} be a closed subspace of a Hilbert space \mathcal{H} . Then, for any $x \in \mathcal{H}$,
- (a) there is a unique element $\hat{x} \in \mathcal{M}$ such that

$$||x - \widehat{x}|| = \inf_{y \in \mathcal{M}} ||x - y||$$
 (3.3)

and

(b)
$$\widehat{x} \in \mathcal{M} \text{ and } ||x - \widehat{x}|| = \inf_{y \in \mathcal{M}} ||x - y||$$
$$\Leftrightarrow \widehat{x} \in \mathcal{M} \text{ and } x - \widehat{x} \in \mathcal{M}^{\perp}. \tag{3.4}$$

3.5 Corollary EXISTENCE OF PROJECTION MAPPING OF \mathcal{H} ONTO \mathcal{M} . Let \mathcal{M} be a closed subspace of a Hilbert space \mathcal{H} , and let I be the identity mapping on \mathcal{H} . Then there is a unique mapping $P_{\mathcal{M}}$ of \mathcal{H} onto \mathcal{M} such that $I-P_{\mathcal{M}}$ maps \mathcal{H} onto \mathcal{M}^{\perp} . $P_{\mathcal{M}}$ is called the projection mapping of \mathcal{H} onto \mathcal{M} .

- **3.6 Theorem** PROPERTIES OF PROJECTION MAPPINGS. Let \mathcal{H} be a Hilbert space and $P_{\mathcal{M}}$ the projection mapping onto the closed subspace \mathcal{M} . Then the following properties hold for all $x, y \in \mathcal{H}$:
- (a) each element $x \in \mathcal{H}$ has a unique representation as the sum of an element of \mathcal{M} and an element of \mathcal{M}^{\perp} ,

$$x = P_{\mathcal{M}}x + (I - P_{\mathcal{M}})x; \qquad (3.5)$$

- (b) $||x||^2 = ||P_{\mathcal{M}}x||^2 + ||(I P_{\mathcal{M}})x||^2$;
- (c) $P_{\mathcal{M}}(\alpha x + \beta y) = \alpha P_{\mathcal{M}} x + \beta P_{\mathcal{M}} y$, for all $\alpha, \beta \in S$;
- (d) $||x_n x|| \underset{n \to \infty}{\longrightarrow} 0 \Rightarrow P_{\mathcal{M}} x_n \underset{n \to \infty}{\longrightarrow} P_{\mathcal{M}} x;$
- (e) $x \in \mathcal{M} \Leftrightarrow P_{\mathcal{M}}x = x$;
- (f) $x \in \mathcal{M}^{\perp} \Leftrightarrow P_{\mathcal{M}}x = 0$;
- (g) if \mathcal{M}_1 and \mathcal{M}_2 are two subspaces of \mathcal{H} such that $\mathcal{M}_1 \subseteq \mathcal{M}_2$, then

$$P_{\mathcal{M}_1}(P_{\mathcal{M}_2}x) = P_{\mathcal{M}_1}x. \tag{3.6}$$

4. Orthonormal sets

- **4.1 Definition** CLOSED SPAN. Let \mathcal{M} be a subset of a Hilbert space \mathcal{H} . The closed span of \mathcal{M} , denoted $\overline{sp}(\mathcal{M})$, is the smallest closed subspace of \mathcal{H} that contains \mathcal{M} . If $\mathcal{M} = \{e_t : t \in T\}$, we can also denote $\overline{sp}(\mathcal{M})$ by $\overline{sp}\{e_t : t \in T\}$.
- **4.2 Proposition** CLOSED SPAN OF A FINITE SET. The closed span of a finite set $\mathcal{M} = \{x_1, ..., x_n\}$ is the set of all linear combinations

$$y = \alpha_1 x_1 + \dots + \alpha_n x_n , \alpha_1, \dots, \alpha_n \in S$$

$$(4.1)$$

where $S = \mathbb{R}$ or $S = \mathbb{C}$.

4.3 Definition ORTHOGONAL AND ORTHONORMAL SETS. Let $\{e_t : t \in T\}$ be a subset of an inner-product space $(\mathcal{H}, \langle \cdot, \cdot \rangle)$. Then $\{e_t : t \in T\}$ is said to be orthogonal if and only if

$$\langle e_s, e_t \rangle = 0$$
, for $s \neq t$. (4.2)

Further, $\{e_t : t \in T\}$ is said to be orthonormal if and only it is orthogonal and $\langle e_t, e_t \rangle = 1$ for all $t \in T$.

4.4 Theorem PROJECTIONS ON FINITE ORTHONORMAL SETS. If $\{e_1, ..., e_n\}$ is an orthonormal subset of a Hilbert space \mathcal{H} and $\mathcal{M} = \overline{sp}\{e_1, ..., e_n\}$, then the following properties hold for all $x \in \mathcal{H}$:

(a)
$$P_{\mathcal{M}}x = \sum_{i=1}^{n} \langle x, e_i \rangle e_i$$
;

$$(b) \|P_{\mathcal{M}}x\|^2 = \sum_{i=1}^n |\langle x, e_i \rangle|^2;$$

(c)
$$\|x - \sum_{i=1}^{n} \langle x, e_i \rangle e_i\| \le \|x - \sum_{i=1}^{n} c_i e_i\|$$
, for all $c_1, ..., c_n \in S$;

(d)
$$\|x - \sum_{i=1}^{n} \langle x, e_i \rangle e_i\| = \|x - \sum_{i=1}^{n} c_i e_i\| \iff c_i = \langle x, e_i \rangle , i = 1, ..., n;$$
 (4.3)

(e)
$$\sum_{i=1}^{n} |\langle x, e_i \rangle|^2 \le ||x||^2$$
. (Bessel's inequality)

4.5 Remark The coefficients $c_i = \langle x, e_i \rangle$, i = 1, ..., n, are called the *Fourier coefficients* relative to the set $\{e_1, ..., e_n\}$.

4.6 Corollary PROJECTION ON A UNIDIMENSIONAL SUBSPACE. Suppose $\mathcal{M} = \overline{sp}\{e_1\}$ where $e_1 \in \mathcal{H}$ and $e_1 \neq 0$. Then

$$P_{\mathcal{M}}x = \frac{\langle x, e_1 \rangle}{\langle e_1, e_1 \rangle} e_1 = \frac{\langle x, e_1 \rangle}{\|e_1\|^2} e_1 . \tag{4.4}$$

4.7 Corollary PROJECTIONS ON FINITE ORTHOGONAL SETS. Let $\{e_1, \ldots, e_n\}$ be a finite set of non-zero orthogonal vectors in a Hilbert space \mathcal{H} , let $\mathcal{M} = \overline{sp}\{e_1, \ldots, e_n\}$, and let $\beta_i(x) = \langle x, e_i \rangle / \langle e_i, e_i \rangle$, $i = 1, \ldots, n$. Then the following properties hold for all $x \in \mathcal{H}$:

(a)
$$P_{\mathcal{M}}x = \sum_{i=1}^{n} \beta_i(x)e_i$$
;

(b)
$$||P_{\mathcal{M}}x||^2 = \sum_{i=1}^n |\langle x, e_i \rangle|^2 / ||e_i||^2 = \sum_{i=1}^n |\langle x, e_i \rangle|^2 / \langle e_i, e_i \rangle = \sum_{i=1}^n |\beta_i(x)|^2 ||e_i||^2$$
;

(c)
$$\|x - \sum_{i=1}^{n} \beta_i(x)e_i\| \le \|x - \sum_{i=1}^{n} c_i e_i\|$$
, for all $c_1, \dots, c_n \in S$;

(d)
$$\|x - \sum_{i=1}^{n} \beta_i(x)e_i\| = \|x - \sum_{i=1}^{n} c_i e_i\| \Leftrightarrow c_i = \beta_i, \ i = 1, \dots, n;$$
 (4.5)

(e)
$$\sum_{i=1}^{n} |\langle x, e_i \rangle|^2 / \|e_i\|^2 = \sum_{i=1}^{n} |\beta_i(x)|^2 \|e_i\|^2 \le \|x\|^2$$
. (Bessel's inequality)

- **4.8 Definition** SEPARABILITY. The Hilbert space \mathcal{H} is separable if and only if $\mathcal{H} = \overline{sp}\{e_t : t \in T\}$ where T is a finite or countable infinite set.
- **4.9 Theorem** Orthonormal representations in Separable Hilbert spaces. Let $\mathcal{H} = \overline{sp}\{e_1, e_2, ...\}$ be a separable Hilbert space where $\{e_1, e_2, ...\}$ is an orthonormal set. Then the following properties hold for all $x \in \mathcal{H}$:
- (a) the set of all finite linear combinations of $\{e_1,\,e_2,\,\dots\}$ is dense in $\mathcal H$, i.e., for any $\varepsilon>0$, there is an n>0 and constants $c_1,\,\dots,\,c_n\in S$ such that

$$\left\| x - \sum_{i=1}^{n} c_i e_i \right\| < \varepsilon ; \tag{4.6}$$

(b)
$$x = \sum_{i=1}^{\infty} \langle x, e_i \rangle e_i$$
, i.e.

$$\left\| x - \sum_{i=1}^{n} \langle x, e_i \rangle e_i \right\| \xrightarrow[n \to \infty]{} 0 ; \tag{4.7}$$

(c)
$$||x||^2 = \sum_{i=1}^n |\langle x, e_i \rangle|^2$$
;

$$(d) \langle x, y \rangle = \sum_{i=1}^{\infty} \langle x, e_i \rangle \langle e_i, y \rangle , \forall y \in \mathcal{H} ;$$
 (Parseval's identity)

(e)
$$x = 0 \Leftrightarrow \langle x, e_i \rangle = 0$$
, $i = 1, 2, \dots$

- **4.10 Corollary** Orthogonal representations in Separable Hilbert spaces. Let $\mathcal{H}=\overline{sp}\{e_1,\,e_2,\,\dots\}$ be a separable Hilbert space where $\{e_1,\,e_2,\,\dots\}$ is a set of non-zero orthogonal vectors, and let $\beta_i(x)=\left\langle x,e_i\right\rangle/\left\langle e_i,e_i\right\rangle$, $i=1,\,\dots,n$. Then the following properties hold for all $x\in\mathcal{H}$:
- (a) the set of all finite linear combinations of $\{e_1, e_2, \dots\}$ is dense in \mathcal{H} , i.e., for any $\varepsilon > 0$, there is an n > 0 and constants $e_1, \dots, e_n \in S$ such that

$$\left\| x - \sum_{i=1}^{n} c_i e_i \right\| < \varepsilon ; \tag{4.8}$$

(b)
$$x = \sum_{i=1}^{\infty} \beta_i(x)e_i$$
, i.e.
$$\left\|x - \sum_{i=1}^{n} \beta_i e_i\right\| \xrightarrow[n \to \infty]{} 0;$$
 (4.9)

(c)
$$||x||^2 = \sum_{i=1}^n |\langle x, e_i \rangle|^2 / ||e_i||^2 = \sum_{i=1}^n |\langle x, e_i \rangle|^2 / \langle e_i, e_i \rangle = \sum_{i=1}^n |\beta_i(x)|^2 ||e_i||^2$$
;
(d) $\langle x, y \rangle = \sum_{i=1}^\infty \langle x, e_i \rangle \langle e_i, y \rangle / ||e_i||^2 = \sum_{i=1}^\infty \langle x, e_i \rangle \langle e_i, y \rangle / \langle e_i, e_i \rangle$

$$\begin{array}{lll} \langle x,y\rangle & = & \sum\limits_{i=1}^{\infty} \langle x,e_{i}\rangle \langle e_{i},y\rangle \, / \, \|e_{i}\| \, = \, \sum\limits_{i=1}^{\infty} \langle x,e_{i}\rangle \langle e_{i},y\rangle \, / \, \langle e_{i},e_{i}\rangle \\ \\ & = & \sum\limits_{i=1}^{\infty} \beta_{i}(x) \overline{\beta_{i}(y)} \, \|e_{i}\|^{2} \, , \, \forall \, y \in \mathcal{H} \, ; \end{array} \qquad \begin{array}{ll} \text{(Parseval's identity)} \end{array}$$

(e)
$$x = 0 \Leftrightarrow \langle x, e_i \rangle = 0, i = 1, 2, ...$$

- **4.11 Corollary** PROJECTIONS IN SEPARABLE HILBERT SPACES. Under the assumptions of Theorem **4.9**, we have for each $x \in \mathcal{H}$:
- (a) for any $\varepsilon > 0$, there is an n > 0 such that

$$||x - P_{E_n}x|| < \varepsilon \tag{4.10}$$

where $E_n = \overline{sp}\{e_1, ..., e_n\};$

$$(b) P_{E_n} x \xrightarrow[n \to \infty]{} x.$$

5. Conditional expectations

5.1 Definition Conditional expectation with respect to a closed subspace. Let \mathcal{M} be a closed subspace of $L^2 \equiv L^2(\Omega, \mathcal{A}, P)$ containing the constant functions and let $Y \in L^2$. Then the conditional expectation of Y given \mathcal{M} is the projection of Y on \mathcal{M} :

$$E_{\mathcal{M}}Y = P_{\mathcal{M}}Y. \tag{5.1}$$

5.2 Definition Conditional expectation with respect to a random vector. Let $X = (X_1, ..., X_k)'$ be a vector of random variables in $L^2 \equiv L^2(\Omega, \mathcal{A}, P)$ and let $L^2(X)$ be the closed subspace of $L^2(\Omega, \mathcal{A}, P)$ consisting of all the random variables in L^2 of the form $\phi(X)$ for some function ϕ . Then the conditional expectation of Y given X is the projection of Y on $L^2(X)$:

$$E(Y|X) = E(Y|X_1, ..., X_k) = P_{L^2(X)}Y$$
 (5.2)

6. Sources and additional references

A good summary of Hilbert space theory aimed at applications in time series analysis may be found in Brockwell and Davis (1991, Chapter 2). Other good reviews appear in: Debnath and Mikusiński (1990) for general applications, Small and McLeish (1994) for applications in statistical theory, and Young (1988) for a more mathematically oriented presentation.

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