

# Hilbert spaces\*

Jean-Marie Dufour<sup>†</sup>  
McGill University

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<sup>†</sup> William Dow Professor of Economics, McGill University, Centre interuniversitaire de recherche en analyse des organisations (CIRANO), and Centre interuniversitaire de recherche en économie quantitative (CIREQ). Mailing address: Department of Economics, McGill University, Leacock Building, Room 519, 855 Sherbrooke Street West, Montréal, Québec H3A 2T7, Canada. TEL: (1) 514 398 8879; FAX: (1) 514 398 4938; e-mail: jean-marie.dufour@mcgill.ca . Web page: <http://www.jeanmariedufour.com>

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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  $x, y, z \in \mathcal{H}$ ,

$$(a) \quad \langle x, y \rangle = \overline{\langle y, x \rangle};$$

$$(b) \quad \langle x + y, z \rangle = \langle x, z \rangle + \langle y, z \rangle;$$

$$(c) \quad \langle \alpha x, y \rangle = \alpha \langle x, y \rangle, \text{ for all } \alpha \in S;$$

$$(d) \quad \langle x, x \rangle \geq 0;$$

$$(e) \quad \langle x, x \rangle = 0 \text{ if and only if } x = 0.$$

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

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

**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

$$\|x\| = \sqrt{\langle x, x \rangle}. \quad (1.1)$$

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

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

where  $x = (x_1, \dots, x_n)'$   $y = (y_1, \dots, y_n)'$ , 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

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

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

$$\langle x, y \rangle = \sum_{i=1}^n x_i \bar{y}_i \quad (1.4)$$

where  $x = (x_1, \dots, x_n)'$   $y = (y_1, \dots, y_n)'$ , 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:

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

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

$$\langle X, Y \rangle = E(XY) \quad (1.6)$$

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

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

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

$$\langle X, Y \rangle = E(X\bar{Y}) \quad (1.8)$$

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

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

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

$$|\langle x, y \rangle| \leq \|x\| \|y\| \quad (1.10)$$

and

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

**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*

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

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

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

**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  $x, y \in \mathcal{H}$  :*

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

$$(b) \quad \|x + y\| \leq \|x\| + \|y\| ; \quad (\text{Triangle inequality})$$

$$(c) \quad \|\alpha x\| \leq |\alpha| \|x\| , \text{ for all } \alpha \in \mathbb{C} ;$$

$$(d) \quad \|x\| \geq 0 ;$$

$$(e) \quad \|x\| = 0 \text{ if and only if } x = 0 ;$$

$$(f) \quad \|x + y\|^2 + \|x - y\|^2 = 2\|x\|^2 + 2\|y\|^2 . \quad (\text{Parallelogram law})$$

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

$$\|x_n - x\| \xrightarrow{n \rightarrow \infty} 0. \quad (1.14)$$

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

$$\|x_n\| \xrightarrow{n \rightarrow \infty} \|x\| \quad (1.15)$$

and

$$\langle x_n, y_n \rangle \xrightarrow{n \rightarrow \infty} \langle x, y \rangle. \quad (1.16)$$

**1.13 Example** MEAN SQUARE CONVERGENCE. *Let  $\{X_n : n = 1, 2, \dots\}$  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 convergence:*

$$\|X_n - X\| \xrightarrow{n \rightarrow \infty} 0 \Leftrightarrow E[(X_n - X)^2] \xrightarrow{n \rightarrow \infty} 0. \quad (1.17)$$

## 2. Hilbert spaces

**2.1 Definition** CAUCHY SEQUENCE. *Let  $\{x_n : n = 1, 2, \dots\}$  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*

$$\|x_n - x_m\| \xrightarrow{m, n \rightarrow \infty} 0. \quad (2.1)$$

**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  $\mathcal{H}$ ,*

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



**2.3 Proposition** NORM CONVERGENCE AND CAUCHY CRITERION. Let  $\{x_n : n = 1, 2, \dots\}$  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.

$$x_n \in \mathcal{M}, \forall n, \text{ and } \|x_n - x\| \xrightarrow{n \rightarrow \infty} 0 \Rightarrow x \in \mathcal{M}. \quad (3.1)$$

**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.

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

**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

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

and

$$\begin{aligned}
(b) \quad & \hat{x} \in \mathcal{M} \text{ and } \|x - \hat{x}\| = \inf_{y \in \mathcal{M}} \|x - y\| \\
& \Leftrightarrow \hat{x} \in \mathcal{M} \text{ and } x - \hat{x} \in \mathcal{M}^\perp.
\end{aligned} \tag{3.4}$$

**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  $\mathcal{M}^\perp$ ,

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

$$(b) \|x\|^2 = \|P_{\mathcal{M}}x\|^2 + \|(I - P_{\mathcal{M}})x\|^2;$$

$$(c) P_{\mathcal{M}}(\alpha x + \beta y) = \alpha P_{\mathcal{M}}x + \beta P_{\mathcal{M}}y, \text{ for all } \alpha, \beta \in S;$$

$$(d) \|x_n - x\| \xrightarrow{n \rightarrow \infty} 0 \Rightarrow P_{\mathcal{M}}x_n \xrightarrow{n \rightarrow \infty} P_{\mathcal{M}}x;$$

$$(e) x \in \mathcal{M} \Leftrightarrow P_{\mathcal{M}}x = x;$$

$$(f) x \in \mathcal{M}^\perp \Leftrightarrow P_{\mathcal{M}}x = 0;$$

(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  $\mathcal{M} = \{x_1, \dots, x_n\}$  is the set of all linear combinations

$$y = \alpha_1 x_1 + \dots + \alpha_n x_n, \quad \alpha_1, \dots, \alpha_n \in S \quad (4.1)$$

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

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

Further,  $\{e_t : t \in T\}$  is said to be orthonormal if and only if 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  $\{e_1, \dots, e_n\}$  is an orthonormal subset of a Hilbert space  $\mathcal{H}$  and  $\mathcal{M} = \overline{sp}\{e_1, \dots, e_n\}$ , then the following properties hold for all  $x \in \mathcal{H}$ :

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

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

$$(c) \quad \left\| x - \sum_{i=1}^n \langle x, e_i \rangle e_i \right\| \leq \left\| x - \sum_{i=1}^n c_i e_i \right\|, \text{ for all } c_1, \dots, c_n \in S;$$

(d)

$$\left\| x - \sum_{i=1}^n \langle x, e_i \rangle e_i \right\| = \left\| x - \sum_{i=1}^n c_i e_i \right\| \Leftrightarrow c_i = \langle x, e_i \rangle, i = 1, \dots, n; \quad (4.3)$$

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

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

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

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

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

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

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

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

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

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

**4.8 Definition** SEPARABILITY. *The Hilbert space  $\mathcal{H}$  is separable if and only if  $\mathcal{H} = \overline{\text{span}}\{e_t : t \in T\}$  where  $T$  is a finite or countable infinite set.*

**4.9 Theorem** ORTHONORMAL REPRESENTATIONS IN SEPARABLE HILBERT SPACES. *Let  $\mathcal{H} = \overline{\text{span}}\{e_1, e_2, \dots\}$  be a separable Hilbert space where  $\{e_1, e_2, \dots\}$  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  $c_1, \dots, c_n \in \mathbb{C}$  such that*

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

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

$$\left\| x - \sum_{i=1}^n \langle x, e_i \rangle e_i \right\| \xrightarrow{n \rightarrow \infty} 0 ; \quad (4.7)$$

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

(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  $\mathcal{H} = \overline{\text{span}}\{e_1, e_2, \dots\}$  be a separable Hilbert space where  $\{e_1, e_2, \dots\}$  is a set of non-zero orthogonal vectors, and let  $\beta_i(x) = \langle x, e_i \rangle / \langle e_i, e_i \rangle$  ,  $i = 1, \dots, n$ . 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  $c_1, \dots, c_n \in S$  such that

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

$$(b) \quad x = \sum_{i=1}^{\infty} \beta_i(x) e_i, \text{ i.e.}$$

$$\left\| x - \sum_{i=1}^n \beta_i e_i \right\| \xrightarrow{n \rightarrow \infty} 0 ; \quad (4.9)$$

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

$$(d) \quad \begin{aligned} \langle x, y \rangle &= \sum_{i=1}^{\infty} \langle x, e_i \rangle \langle e_i, y \rangle / \|e_i\|^2 = \sum_{i=1}^{\infty} \langle x, e_i \rangle \langle e_i, y \rangle / \langle e_i, e_i \rangle \\ &= \sum_{i=1}^{\infty} \beta_i(x) \overline{\beta_i(y)} \|e_i\|^2, \quad \forall y \in \mathcal{H} ; \quad (\text{Parseval's identity}) \end{aligned}$$

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

**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

$$\|x - P_{E_n} x\| < \varepsilon \quad (4.10)$$

where  $E_n = \overline{\text{span}}\{e_1, \dots, e_n\}$ ;

$$(b) \quad P_{E_n} x \xrightarrow{n \rightarrow \infty} x.$$

## 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  $\mathcal{M}$  :

$$E_{\mathcal{M}}Y = P_{\mathcal{M}}Y . \quad (5.1)$$

**5.2 Definition** **CONDITIONAL EXPECTATION WITH RESPECT TO A RANDOM VECTOR.** Let  $X = (X_1, \dots, X_k)'$  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  $\phi$ . Then the conditional expectation of  $Y$  given  $X$  is the projection of  $Y$  on  $L^2(X)$  :

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

## 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.

## References

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