## **TOPICS IN STATISTICAL THEORY – EXAMPLES 1/2**

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**Ex. 1** In what follows  $L^2([0,1])$  denotes the space of functions  $f:[0,1] \to \mathbb{R}$  such that

$$\|f\|_{L^2([0,1])}^2 \equiv \int_0^1 f^2(x) dx < \infty.$$

where dx is Lebesgue measure on [0, 1]. Moreover  $L^2(P)$  denotes the space of random variables X defined on the probability space  $(\Omega, \mathcal{A}, P)$  such that  $EX^2 \equiv ||X||_{L^2(P)}^2 < \infty$ .

a) [Existence of the Wiener and the Gaussian white noise process]. Show that the mappings  $\Phi: T \times T \to \mathbb{R}$  given by

$$\Phi(s,t) = \min(s,t), \ T = [0,1],$$

and by

$$\Phi(s,t) = \int_0^1 s(x)t(x)dx, \quad T = L^2([0,1]),$$

define *covariances* of a Gaussian process  $(G(t) : t \in T)$ .

b) [Construction of Gaussian white noise as a stochastic integral.] Let  $(W(t) : t \in [0, 1])$  be a Wiener process defined on some probability space  $(\Omega, \mathcal{A}, P)$ . For  $h \in L^2([0, 1])$  take approximations  $h_N = \sum_{i < N} a_i \mathbf{1}_{I_i}$  where the  $I_i = (x_i, y_i]$ 's are disjoint intervals, and where the  $a_i$ 's are real coefficients, such that  $||h_N - h||_{L^2([0,1])} \to 0$  as  $N \to \infty$ . Define new random variables

$$\mathbb{W}(1_{(x,y]}) = W(y) - W(x), x < y, \text{ and } \mathbb{W}(h_N) = \sum_{i \le N} a_i \mathbb{W}(1_{I_i}).$$

Show that the  $L^2(P)$ -limit  $\mathbb{W}(h)$  of  $\mathbb{W}(h_N)$  exists, and that  $(\mathbb{W}(h) : h \in L^2([0,1])$  is a Gaussian process with covariance  $E\mathbb{W}(g)\mathbb{W}(h) = \int_0^1 g(x)h(x)dx$  for all  $g, h \in L^2([0,1])$ .

c) Consider an infinite sequence of random variables  $(g_k : k \in \mathbb{N})$ , all i.i.d. N(0, 1), defined on some common probability space  $(\Omega, \mathcal{A}, P)$ . For  $\{e_k : k \in \mathbb{N}\}$  any orthonormal basis of  $L^2([0, 1])$ , show that  $\mathbb{W}(h)$  from b) can also be obtained as the limit in  $L^2(P)$  of the random variables  $\tilde{\mathbb{W}}_N = \sum_{k \leq N} h_k g_k$  where  $h_k = \int_0^1 h(x) e_k(x) dx$ .

**Ex. 2** Let P, Q be two probability measures on a measure space  $(\mathcal{X}, \mathcal{A})$ . Suppose there exists a common dominating measure  $\mu$  on  $(\mathcal{X}, \mathcal{A})$ , and let  $p = dP/d\mu$ ,  $q = dQ/d\mu$ . Show that

$$\sup_{A \in \mathcal{A}} |P(A) - Q(A)| = \frac{1}{2} \int_{\mathcal{X}} |p(x) - q(x)| d\mu(x).$$

**Ex. 3** Let  $N \in \mathbb{N}$  and let  $\{x_0, x_1, \ldots, x_N\}$  be the dissection of [0, 1] given by  $x_i = i/N, i = i/N$  $0,\ldots,N$ . Denote by  $\mathcal{V}_N$  the space of functions that are piece-wise constant on the intervals  $I_{iN} = (x_{i-1}, x_i], i = 1, \dots N$ . Show that

a) the mapping  $(f,g) \mapsto \langle f,g \rangle_N = \sum_{i=1}^N f(x_i)g(x_i)$  defines an inner product on  $\mathcal{V}_N$ , b) the functions  $\{1_{I_{iN}} : i = 1, \dots N\}$  form an orthonormal basis of  $\mathcal{V}_N$  for  $\langle \cdot, \cdot \rangle_N$ ,

c) the function

$$f_N = \sum_{i=1}^N \langle f, 1_{I_{iN}} \rangle_N 1_{I_{in}} \in \mathcal{V}_N$$

interpolates a function f at the points  $x_i$  (that is, it satisfies  $f_N(x_i) = f(x_i)$  for all i = 1, ..., N),

d) the functions

$$\{\phi_{iN} \equiv \sqrt{n} \mathbf{1}_{I_{iN}} : i = 1, \dots N\}$$

form an orthonormal basis of  $\mathcal{V}_N$  for the inner product  $\langle \cdot, \cdot \rangle_{L^2}$  of  $L^2([0,1])$ ,

e) the  $L^2([0,1])$ -projection of  $f \in L^2$  onto  $\mathcal{V}_N$ ,

$$\pi_N(f) = \sum_{i=1}^N \langle f, \phi_{iN} \rangle_{L^2} \phi_{iN},$$

satisfies, for any f of  $\alpha$ -Hölder norm  $||f||_{C^{\alpha}} \leq M, 0 < \alpha \leq 1$ , the bound

$$||f - \pi_N(f)||_{L^2}^2 \le ||f - f_N||_{L^2}^2 \lesssim M^2 N^{-2\alpha}.$$

**Ex.** 4 Let  $\alpha \in \mathbb{N}$ . Consider a function  $f \in L^2([0,1])$  such that  $D^m f \in L^2, D^m f(0) = D^m f(1)$ , for all  $0 \leq m \leq \alpha$ . Let  $\langle \cdot, \cdot \rangle$  denote the usual inner product of  $L^2([0,1])$  and consider the trigonometric basis of  $L^2([0,1])$  given by  $\{e_k = e^{-2\pi i k \cdot} : k \in \mathbb{Z}\}$ . Show that

$$\sum_{k\in\mathbb{Z}} |\langle f, e_k \rangle|^2 (1+k^{2\alpha}) < \infty.$$

If 
$$K \in \mathbb{N}$$
 and  $f_K = \sum_{|k| \le K} \langle f, e_k \rangle e_k$ , deduce that  $\|f_K - f\|_{L^2([0,1])}^2 \le K^{-2\alpha}$ 

**Ex. 5** Suppose two probability measures P, Q have a common dominating measure  $\mu$  on a measure space  $(\mathcal{X}, \mathcal{A})$ , and are absolutely continuous with respect to each other. Write  $p = dP/d\mu$  and  $q = dQ/d\mu$ , and denote by K(P, Q) the Kullback-Leibler divergence between P and Q. Show that

$$E_P \left| \log \frac{p}{q}(X) \right| \le K(P,Q) + \sqrt{2K(P,Q)}$$

**Ex.** 6 Let  $X_1, \ldots, X_n$  be i.i.d. random variables with bounded density  $f : [0,1] \to [0,\infty)$ . Assuming that f is also  $\alpha$ -Hölder for some  $0 < \alpha \leq 1$ , show that one can construct an estimator  $\hat{f}_n$  of f such that

$$E_f \|\hat{f}_n - f\|_{L^2([0,1])}^2 \le C n^{-2\alpha/(2\alpha+1)},$$

where C > 0 is a constant independent of sample size. [Hint: Estimate the Haar wavelet approximation  $\Pi_{V_J}(f)$  of f at resolution level J unbiasedly from the sample, and choose J to 'balance bias and variance'.]

**Ex.** 7 In the setting of the previous exercise with  $\alpha = 1$ , show carefully that the rate of convergence obtained is minimax optimal over the class

$$\mathcal{F} = \left\{ f: [0,1] \to [0,B], \ \int_0^1 f(x) dx = 1, \ \sup_{x \neq y} \frac{|f(x) - f(y)|}{|x - y|} \le B \right\}, \ B > 1,$$

of probability density functions that are Lipschitz (i.e.,  $\alpha$ -Hölder with  $\alpha = 1$ ).