

Abstract Wiener Space¹

Starting point:²

- A sufficiently rich probability space $\mathfrak{F} = (\Omega, \mathcal{F}, \mathbb{P})$;
- A locally convex linear (over \mathbb{R}) topological space \mathbb{X} with (topological) dual \mathbb{X}^* ;
- A random element \mathbf{X} on \mathfrak{F} with values in \mathbb{X} such that, for every $f \in \mathbb{X}^*$, the random variable $f(\mathbf{X})$ is Gaussian with mean zero;
- The measure μ on $(\mathbb{X}, \mathcal{B}(\mathbb{X}))$ defined by $\mu(A) = \mathbb{P}(\mathbf{X} \in A)$ and the corresponding Hilbert space $L_2(\mathbb{X}, \mu)$.

Basic constructions.

- (1) **Covariance operator** \mathbf{K} of \mathbf{X} is a continuous linear mapping from \mathbb{X}^* to \mathbb{X} : for $f, g \in \mathbb{X}^*$,

$$\mathbb{E}(f(\mathbf{X})g(\mathbf{X})) = f(\mathbf{K}(g)) = g(\mathbf{K}(f)).$$

- (2) The **canonical embedding operator** \mathbf{i}^* of \mathbb{X}^* into $L_2(\mathbb{X}, \mu)$: if $f \in \mathbb{X}^*$, then $f \in L_2(\mathbb{X}, \mu)$, because

$$\int_{\mathbb{X}} |f(x)|^2 \mu(dx) = \mathbb{E}|f(\mathbf{X})|^2 < \infty.$$

- (3) The Hilbert space \mathbb{X}_μ^* of **measurable linear functionals** is the closure of the image $\mathbf{i}^*(\mathbb{X}^*)$ in $L_2(\mathbb{X}, \mu)$; this space is identified with its (topological) dual by the Riesz representation theorem. If $g \in \mathbb{X}_\mu^*$, then $g(\mathbf{X})$ is defined and is a zero-mean Gaussian random variable (being a mean-square limit of zero-mean Gaussian random variables).

- (4) The **dual operator** $\mathbf{i} : \mathbb{X}_\mu^* \rightarrow \mathbb{X}$ for the operator $\mathbf{i}^* : \mathbb{X}^* \mapsto \mathbb{X}_\mu^*$:

$$f(\mathbf{i}(g)) = \mathbb{E}(f(\mathbf{X})g(\mathbf{X})) \equiv (\mathbf{i}^* f, g)_{\mathbb{X}_\mu^*}.$$

- (5) The **Cameron-Martin space** $H_\mu = \mathbf{i}(\mathbb{X}_\mu^*)$, which is a separable Hilbert space with inner product

$$(x, y)_{H_\mu} = (\mathbf{i}^{-1}(x), \mathbf{i}^{-1}(y))_{L_2(\mathbb{X}, \mu)} \equiv \mathbb{E}((\mathbf{i}^{-1}(x))(\mathbf{X})(\mathbf{i}^{-1}(y))(\mathbf{X})), \quad (1.1)$$

and is compactly embedded into \mathbb{X} .

As a result,

$$\mathbf{K} = \mathbf{i}^*, \quad \mathbf{K}(\mathbb{X}^*) \subset H_\mu \subset \mathbb{X},$$

and H_μ is a *reproducing kernel Hilbert space* with kernel

$$K(f, g) = \mathbb{E}(f(\mathbf{X})g(\mathbf{X})), \quad f, g \in \mathbb{X}^*.$$

In the above setting, the **abstract Wiener space** is the triple (\mathbb{X}, H_μ, μ) . In the original construction of LEONARD GROSS³ (around 1965), the *starting point* is the triple, in which \mathbb{X} is separable Banach space and H_μ is a separable Hilbert space that is densely and continuously embedded into \mathbb{X} ; the key point is *existence* of the corresponding measure μ .

The main example is $\mathbb{X} = \mathcal{C}((0, T))$, the space of continuous functions on $[0, T]$, and $\mathbf{X} = W$, a standard Brownian motion. Then $H_\mu = \{f \in H_1((0, T)) : f(0) = 0\}$, where

$$H_1((0, T)) = \left\{ f : f(t) = f(0) + \int_0^t f'(s) ds, \int_0^T |f'(s)|^2 ds < \infty. \right\}$$

The **abstract Cameron-Martin theorem** becomes as follows: given a non-random $x \in \mathbb{X}$, the distribution μ of \mathbf{X} and the distribution μ_x of $\mathbf{X} + x$ [defined by $\mu_x(A) = \mathbb{P}(\mathbf{X} + x \in A)$] are mutually absolutely continuous if and only if $x \in H_\mu$, and in that case,

$$\frac{d\mu_x}{d\mu}(y) = \exp \left((\mathbf{i}^{-1}(x))(y) - \frac{\|x\|_{H_\mu}^2}{2} \right), \quad y \in \mathbb{X}. \quad (1.2)$$

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²Some stretch of imagination might be necessary to resolve potential questions about existence.

³b. 1931, Professor Emeritus at Cornell

Otherwise, the distributions are mutually singular [supported on disjoint sets].

Two related results.

Theorem. [J. Hájek (1958), J. Feldman (1958)] Two Gaussian measures on a locally convex linear topological are either equivalent or singular.

Theorem. [S. Kakutani (1948)] Two infinite product measures are equivalent if and only if the series

$$\sum_{k \geq 1} \ln \int \sqrt{\frac{d\nu_k}{d\mu_k}} d\mu_k \quad (1.3)$$

converges.

The Gaussian product measure on a separable Hilbert space.⁴

Let \mathbb{X} be a separable Hilbert space with inner product (\cdot, \cdot) , norm $\|\cdot\|$, and an orthonormal basis $\{\varphi_k, k \geq 1\}$, and let \mathbf{X} be a Gaussian random element with values in \mathbb{X} : $\mathbf{X} = \sum_k q_k \xi_k \varphi_k$ where $\xi_k, k \geq 1$, are iid $\mathcal{N}(0, 1)$, $q_k > 0$, and $\sum_k q_k^2 < \infty$:

$$\mathbb{E}\|\mathbf{X}\|^2 = \sum_k q_k^2.$$

Then

- (1) $\mathbb{X}^* = \mathbb{X}$;
- (2) The operator \mathbf{K} is positive and symmetric, with $\mathbf{v} = \mathbf{v}^* = \mathbf{K}^{1/2}$:

$$\mathbf{K}(f) = \sum_k q_k^2 f_k \varphi_k, \quad f_k = (f, \varphi_k);$$

- (3) The distribution μ of \mathbf{X} in H is a product measure of $\mathcal{N}(0, q_k^2)$;
- (4) For $x \in \mathbb{X}$, the distribution $\nu = \mu_x$ of $\mathbf{X} + x$ is the product measure of $\mathcal{N}(x_k, q_k^2)$ so that

$$\frac{d\nu_k}{d\mu_k}(t) = \exp\left(\frac{2tx_k - x_k^2}{2q_k^2}\right), \quad t \in \mathbb{R}. \quad (1.4)$$

With $\zeta_k \sim \mathcal{N}(0, q_k^2)$,

$$\int \sqrt{\frac{d\nu_k}{d\mu_k}} d\mu_k = e^{-x_k^2/(4q_k^2)} \mathbb{E}e^{\zeta_k x_k/(2q_k^2)} = e^{-x_k^2/(8q_k^2)}, \quad (1.5)$$

and (1.3) becomes

$$\sum_k q_k^{-2} x_k^2 < \infty. \quad (1.6)$$

- (5) The abstract construction leads to the Cameron-Martin space

$$H_\mu = \mathbf{K}^{1/2}(\mathbb{X}) = \left\{f \in \mathbb{X} : \sum_k q_k^{-2} f_k^2 < \infty\right\},$$

which is consistent with (1.6).

- (6) Equality (1.4) is consistent with (1.2): just as in (1.5), we write $y_k = t/q_k$ to get

$$\prod_k \frac{d\nu_k}{d\mu_k}(y_k) = \exp\left(\left(\mathbf{K}^{-1/2}x, y\right) - \frac{1}{2}\|\mathbf{K}^{-1/2}x\|^2\right).$$

⁴Sometimes referred to as the “trivial example”.