# Gaussian Measures in Hilbert Spaces [Da Prato, 2006, Chap. 1]

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# **Motivation**

"There is no infinite dimensional Lebesgue measure"

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#### **Theorem**

Let  $(E, \|\cdot\|)$  be a normed space with  $\dim E = \infty$ . Then there is no non-trivial, translation-invariant,  $\sigma$ -additive Borel measure  $\mu$  on  $(E, \|\cdot\|)$  s.t.  $\mu[B_{\varepsilon}(0)] < \infty$  for all  $\varepsilon > 0$ .

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Alternative: Gaussian measures

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Unless otherwise specified,  ${\cal H}$  denotes a real, separable Hilbert space.

$$\begin{split} \mathcal{B}(H) & \text{... Borel } \sigma - \text{algebra on } H \\ L(H) & := \{T \in L(H) | \text{ linear, bounded} \} \\ L^+(H) & := \{T \in L(H) | \text{ symmetric, pos. semi-definite} \} \\ L_1^+(H) & := \left\{ T \in L^+(H) \middle| \sum_{k=1}^\infty \langle e_k, Te_k \rangle < \infty, (e_k)_{k \in \mathbb{N}} \text{ ONB of } H \right\}. \end{split}$$

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For 
$$H=\mathbb{R}^d$$
 we have 
$$L_1^+(\mathbb{R}^d)=L^+(\mathbb{R}^d)\subset L(\mathbb{R}^d).$$

# Spectral Theorem for $Q \in L_1^+(H)$

#### **Theorem**

Let  $Q \in L_1^+(H)$ . Then there exists an ONB  $(e_k)_{k \in \mathbb{N}}$  of H and a sequence  $(\lambda_k)_{k \in \mathbb{N}} \in \ell^1$  s.t.

$$\forall k \in \mathbb{N} : Qe_k = \lambda_k e_k, \quad \lambda_k \ge 0$$

in particular  $\lambda_k \to 0$  as  $k \to \infty$ .

## **Product** measures

Define

$$\mathcal{F} := \sigma \left( \underbrace{\{x \in \mathbb{R}^{\infty} : (x_{k_1}, \dots, x_{k_n}) \in A\}}_{C_{k_1, \dots, k_n, A}} : A \in \mathcal{B}(\mathbb{R}^n), n \in \mathbb{N} \right)$$

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 Proposition

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#### **Theorem**

Let  $(\mathbb{P}_k)_{k\in\mathbb{N}}$  be a sequence of probability measures on  $(\mathbb{R},\mathcal{B}(\mathbb{R}))$ . Then there exists a unique probability measure on  $(\mathbb{R}^{\infty},\mathcal{F})$  s.t. for every  $C_{k_1,\ldots,k_n,A}$  we have

$$\mathbb{P}(C_{k_1,\ldots,k_n,A}) = (\mathbb{P}_{k_1} \times \ldots \times \mathbb{P}_{k_n})(A).$$

In particular, for every  $i \in \mathbb{N}$  the projection onto the k-th coordinate  $\pi_k : x \mapsto x_k$  has distribution  $\mathbb{P}_k$  and  $\{\pi_k\}_{k=1}^\infty$  is a set of independent real valued random variables w.r.t.  $\mathbb{P}$ .

# One-dimensional Hilbert spaces

# Definition (1-dim.)

Let  $a \in \mathbb{R}, \lambda \geq 0$ . Then define the measure  $N_{a,\lambda}$  on  $\mathcal{B}(\mathbb{R})$  by

$$(\lambda=0) \quad N_{a,\lambda}(B)=\delta_a(B)=\begin{cases} 1 & a\in B,\\ 0 & a\not\in B \end{cases}, \quad \forall B\in\mathcal{B}(\mathbb{R}),$$
 and

$$(\lambda \neq 0)$$
  $dN_{a,\lambda}(x) = \frac{1}{\sqrt{2\pi\lambda}} \exp\left\{-\frac{(x-a)^2}{2\lambda}\right\} dx.$ 

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# Moments and characteristic function (1-dim.)

For  $a \in \mathbb{R}, \lambda \geq 0$  we have

mean 
$$a = \int_{\mathbb{R}} x \; \mathrm{d}N_{a,\lambda}(x)$$
 variance 
$$\lambda = \int_{\mathbb{R}} (x-a)^2 \; \mathrm{d}N_{a,\lambda}(x)$$
 char. function 
$$\widehat{N_{a,\lambda}}(h) = \exp\left\{iah - \frac{1}{2}\lambda h^2\right\}, \;\; h \in \mathbb{R}$$

# Finite-dimensional Hilbert spaces

# Definition (fin. dim.)

#### **Definition**

A measure  $\mu$  on H is called Gaussian, if for every  $h \in H$  the functional  $x \mapsto \langle h, x \rangle$  has law  $N_{a,\lambda}$  for some  $a \in \mathbb{R}, \lambda \geq 0$ .

# Construction (fin. dim.)

1. Let H be a real Hilbert space with  $\dim(H)=d$ ,  $a\in H$ ,  $Q\in L^+(H)$ . Then let  $\{e_1,\ldots,e_d\}\subseteq H$  be an ONB of H s.t.

$$\forall 1 \le k \le d : Qe_k = \lambda_k e_k, \quad \lambda_k \ge 0 .$$

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$$\forall 1 \le k \le d : Qe_k = \lambda_k e_k, \quad \lambda_k \ge 0$$
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- 2. Identify H with  $\mathbb{R}^d$  via  $x \mapsto (\langle x, e_1 \rangle, \dots, \langle x, e_d \rangle)$ .
- 3. Then define the measure  $N_{a,Q}$  on  $\mathcal{B}(H)$  by

$$N_{a,Q} = \underset{k=1}{\overset{d}{\times}} N_{a_k,\lambda_k}$$

## Is this construction Gaussian?

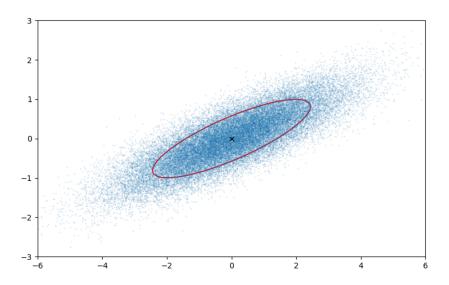
#### **Theorem**

 $N_{a,Q}$  is a Gaussian measure.

Proof.

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# Figure in $H = \mathbb{R}^2$



# Moments and characteristic function (fin. dim.)

For  $a, \in H, Q \in L^+(H)$  we have

mean 
$$a = \int_H x \, \mathrm{d}N_{a,Q}(x)$$
 covariance 
$$\langle y,Qz\rangle = \int_H \langle y,(x-a)\rangle \langle z,(x-a)\rangle \, \mathrm{d}N_{a,Q}(x)$$
 char. functional 
$$\widehat{N_{a,Q}}(h) = \exp\left\{i\langle h,a\rangle - \frac{1}{2}\langle h,Qh\rangle\right\}, \ \ h\in H$$

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 $\langle y, Qz \rangle = \int_{H} \langle y, (x-a) \rangle \langle z, (x-a) \rangle dN_{a,Q}(x)$ 

char. functional 
$$\widehat{N_{a,Q}}(h) = \exp\{i\langle h,a\rangle - \frac{1}{2}\langle h,Qh\rangle\}, h \in H$$

# **Proposition**

covariance

If 
$$\det(Q) > 0$$
 i.e.  $\lambda_k > 0$  for every  $k \in \{1, \dots, d\}$ , then

$$dN_{a,Q}(x) = \frac{1}{\sqrt{(2\pi)^d \det Q}} \exp\left\{-\frac{1}{2}\left\langle (x-a), Q^{-1}(x-a)\right\rangle\right\} dx.$$

# Separable Hilbert spaces

## **Definition of mean**

Let  $\mu$  be a measure on  $(H, \mathcal{B}(H))$  s.t.  $\int_H \|x\| \ \mathrm{d}\mu(x) < \infty$ .

Then  $h \mapsto F(h) := \int_H \langle h, x \rangle d\mu(x)$  is bounded since

$$|F(h)| \leq \int_{H} |\langle h, x \rangle| \, \mathrm{d}\mu(x) \leq \|h\| \underbrace{\int_{H} \|x\| \, \mathrm{d}\mu(x)}_{<\infty}$$

Thus by Riesz' Representation theorem  $\exists! a \in H$ :

$$\langle h, a \rangle = \int_{H} \langle h, x \rangle \, d\mu(x), \quad h \in H.$$

called the **mean of**  $\mu$ .

## Intermezzo: Bochner spaces

It is clear how to integrate  $f:H\to\mathbb{R}$  when there is a measure on H. But how about  $f:H\to H$ , e.g.  $x\mapsto x$ , as in the definition of the mean?

# Intermezzo: Bochner spaces

It is clear how to integrate  $f:H\to\mathbb{R}$  when there is a measure on H. But how about  $f:H\to H$ , e.g.  $x\mapsto x$ , as in the definition of the mean?

Let  $(\Omega,\mathcal{A},\mathbb{P})$  be a probability space. Then define the Bochner space

$$L^p(\Omega;H) := \left\{ u: \Omega \to H | u \text{ measurable}, \underbrace{\int_{\Omega} \|u(\omega)\|_H^p \ \mathrm{d}\mathbb{P}(\omega)}_{=:\|u\|_{L^p(\Omega;H)}^p} < \infty \right\}$$

where  $1 \leq p < \infty$ .

# Intermezzo: Bochner integral

# **Proposition**

The set  $\left\{\sum_{i=1}^{n} 1_{A_i} h_i : A_i \in \mathcal{F}, h_i \in H\right\}$  of simple functions lies dense in  $(L^p(\Omega; H), \|\cdot\|_{L^p(\Omega; H)})$ .

# Intermezzo: Bochner integral

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#### Definition

For  $\sum_{i=1}^n 1_{A_i} h_i \in L^1(\Omega;H)$  define

$$\int \sum_{i=1}^{n} 1_{A_i} h_i d\mathbb{P} = \sum_{i=1}^{n} \mathbb{P}(A_i) h_i \in H$$

For  $u \in L^1(\Omega; H)$ , define the Bochner integral of u as

$$\int u \, d\mathbb{P} := \lim_{k \to \infty} \int \sum_{i=1}^{n^{(k)}} 1_{A_i^{(k)}} h_i^{(k)} d\mathbb{P} \in H$$

# Intermezzo: Bochner integral

# **Proposition**

Let  $f: H \to \mathbb{R}$  be a bounded linear functional and  $u \in L^1(\Omega; H)$ . Then

$$f\left[\int u(\omega) d\mathbb{P}(\omega)\right] = \int f\left[u(\omega)\right] d\mathbb{P}(\omega)$$

# Characterization of the mean

#### **Theorem**

Indeed,

$$a = \int_H x \, \mathrm{d}\mu(x).$$

#### Proof.

Let  $a\in H$  be the mean of  $\mu$  and let  $h\in H$  be arbitrary. Then  $x\mapsto \langle h,x\rangle$  defines a bounded linear functional on H. Hence it can be pulled into the integral and we have

$$\langle h, a \rangle = \int_{H} \langle h, x \rangle \, d\mu(x) = \left\langle h, \int_{H} x \, d\mu(x) \right\rangle$$

Uniqueness of a gives the result.

#### **Definition the covariance**

Let  $\mu$  be a measure on  $(H, \mathcal{B}(H))$  s.t.  $\int_H \|x\|^2 d\mu(x) < \infty$ .

Then  $(h,k)\mapsto G(h,k):=\int_H\langle h,x-a\rangle\langle k,x-a\rangle\;\mathrm{d}\mu(x)$  is bounded since

$$|G(h,k)| \le \int_{H} |\langle h, x - a \rangle| \ |\langle k, x - a \rangle| \ d\mu(x)$$

$$\le ||h|| \ ||k|| \underbrace{\int_{H} ||x - a||^{2} \ d\mu(x)}_{<\infty}$$

Thus by Riesz' Representation theorem there exists a unique bounded linear operator  $Q:H\to H$  s.t.

$$\langle h, Qk \rangle = \int_{H} \langle h, x - a \rangle \langle k, x - a \rangle d\mu(x), \quad h, k \in H.$$

called the **covariance of**  $\mu$ .

# Properties of the covariance

#### **Theorem**

Let  $\mu$  be a measure on  $(H,\mathcal{B}(H))$  s.t. a and Q exist. Then  $Q\in L_1^+(H)$  i.e. Q is symmetric, positive semi-definite and of trace class.

#### Proof.

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## **Definition of Gaussian Measures**

#### **Definition**

A measure  $\mu$  on  $(H,\mathcal{B}(H))$  is called Gaussian if  $\exists a\in H,Q\in L_1^+(H)$  s.t.

$$\int_{H} \exp\left\{i\langle h, x\rangle\right\} \ \mathrm{d}\mu(x) = \underbrace{\exp\left\{i\langle a, h\rangle - \frac{1}{2}\langle h, Qh\rangle\right\}}_{=:\widehat{N_{a,Q}}(h)}, \ h \in H.$$

 $N_{a,Q}$  is called **non-degenerate** if  $\ker Q = \{0\}$ .

## **Definition of Gaussian Measures**

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 $N_{a,Q}$  is called **non-degenerate** if  $\ker Q = \{0\}$ .

Recall: for  $H=\mathbb{R}^n$  the Fourier inversion theorem asserts that two measures with the same characteristic functional are equal. This also is still true when  $\dim H=\infty$ . In particular, Gaussian measures are entirely characterized by their mean and covariance operator.

#### **Existence of Gaussian measures**

1. Let  $a \in H$  and  $Q \in L^+(H)$  with  $\{e_k\}_{k \in \mathbb{N}} \subseteq H$  an ONB of H associated to Q. Then

$$\forall k \in \mathbb{N} : Qe_k = \lambda_k e_k, \quad \lambda_k \ge 0 .$$

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- 2. Identify H with  $\ell^2$  via  $x \mapsto (\langle x, e_k \rangle)_{k \in \mathbb{N}}$ .
- 3. Define the measure  $N_{a,Q}$  on  $\mathcal{B}(\mathbb{R}^{\infty})$  by

$$N_{a,Q} = \underset{k \in \mathbb{N}}{\times} N_{a_k,\lambda_k}.$$

# **Definition** (separable)

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#### **Theorem**

$$\mu := N_{a,Q}$$
 is concentrated on  $\ell^2$  i.e.  $\mu(\ell^2) = 1$ .

Proof.

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#### Is this construction Gaussian?

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 $N_{a,Q}$  is a Gaussian measure.

Proof.

# **Closing remarks**

## What if H is less well-behaved?

More generally, how can one define a Gaussian measure?

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More generally, how can one define a Gaussian measure?

- 1. via a density (needs  $\dim H < \infty$ )
- via cont. linear functions (needs rich enough dual theory e.g. loc. convex TVS)
- 3. via the characteristic functional (needs Fourier theory on H)
- 4. via identification  $H\simeq \ell^2$  (needs H to be a separable Hilbert space)

# **Bibliography**



Da Prato, G. (2006).

An Introduction to Infinite-Dimensional Analysis.

Springer Science & Business Media.