

# Mild Solutions of Stochastic Navier-Stokes Equations-II

## Stochastic Integrals in Infinite Dimensions

Meng Xu

Department of Mathematics  
University of Wyoming

# Outline

- 1 Nuclear and Hilbert-Schmidt Operators
  - Nuclear Operators
  - Hilbert-Schmidt Operators
- 2 Stochastic Integrals in Hilbert Spaces
  - Infinite Dimensional Wiener Processes
  - Martingales in Banach Spaces
  - Construction of Stochastic Integrals
  - Properties of Stochastic Integrals
  - Stochastic Integrals for Cylindrical Wiener Processes

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## Space setting

Let us first introduce some notations of spaces and operators that we will work on.

Let  $(U, \langle \cdot, \cdot \rangle_U)$  and  $(H, \langle \cdot, \cdot \rangle)$  be two separable Hilbert spaces.

$L(U, H)$  denotes the space of bounded linear operators from  $U$  to  $H$ .  $L^*$  is its adjoint.  $L(U) = L(U, U)$ .

We say  $L \in L(U)$  is **symmetric** if

$$\langle Lu, v \rangle_U = \langle u, Lv \rangle_U \quad \text{for all } u, v \in U$$

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## Definition

An element  $T \in L(U, H)$  is said to be a **nuclear operator** if there exists a sequence  $(a_j)_{j \in \mathbb{N}}$  in  $H$  and a sequence  $(b_j)_{j \in \mathbb{N}}$  in  $U$  such that

$$Tx = \sum_{j=1}^{\infty} a_j \langle b_j, x \rangle_U \quad \text{for all } x \in U$$

and

$$\sum_{j \in \mathbb{N}} \|a_j\| \cdot \|b_j\| < \infty$$

Denote by  $L_1(U, H)$  the space of all nuclear operators from  $U$  to  $H$ . If  $U = H$ ,  $T \in L_1(U, H)$  is nonnegative and symmetric, then  $T$  is called trace class.



## Lemma

The space  $L_1(U, H)$  endowed with the norm

$$\|T\|_{L_1(U, H)} = \inf \left\{ \sum_{j \in \mathbb{N}} \|a_j\| \cdot \|b_j\|_U \mid Tx = \sum_{j=1}^{\infty} a_j \langle b_j, x \rangle_U, x \in U \right\}$$

is a Banach space.

## Definition

Let  $T \in L(U)$ ,  $e_k, k \in \mathbb{N}$  be an orthonormal basis of  $U$ . Then we define

$$\text{tr}T := \sum_{k \in \mathbb{N}} \langle Te_k, e_k \rangle_U$$

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## Trace and $L_1(U)$

The relation between a trace operator and nuclear operator is given by the following lemma.

### Lemma

If  $T \in L_1(U)$ , then  $\text{tr}T$  is well-defined independently of the choice of orthonormal basis  $e_k, k \in \mathbb{N}$ . Moreover we have

$$|\text{tr}T| \leq \|T\|_{L_1(U)}$$

**Proof:** Let  $(a_j)_{j \in \mathbb{N}}$  and  $(b_j)_{j \in \mathbb{N}}$  be sequences in  $U$  such that

$$Tx = \sum_{j \in \mathbb{N}} a_j \langle b_j, x \rangle_U \quad \text{for all } x \in U$$

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## Proof continues...

Then we get for any orthonormal basis  $e_k, k \in \mathbb{N}$  of  $U$  that

$$\langle Te_k, e_k \rangle_U = \sum_{j \in \mathbb{N}} \langle e_k, a_j \rangle_U \cdot \langle e_k, b_j \rangle_U$$

Therefore

$$\begin{aligned} & \sum_{k \in \mathbb{N}} |\langle Te_k, e_k \rangle_U| \\ & \leq \sum_{j \in \mathbb{N}} \sum_{k \in \mathbb{N}} |\langle e_k, a_j \rangle_U \cdot \langle e_k, b_j \rangle_U| \\ & \leq \sum_{j \in \mathbb{N}} \left( \sum_{k \in \mathbb{N}} |\langle e_k, a_j \rangle_U|^2 \right)^{1/2} \cdot \left( \sum_{k \in \mathbb{N}} |\langle e_k, b_j \rangle_U|^2 \right)^{1/2} \\ & = \sum_{j \in \mathbb{N}} \|a_j\|_U \cdot \|b_j\|_U < \infty \end{aligned}$$



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$|\operatorname{tr} T| \leq \|T\|_{L_1(U)}$  follows and we can thus exchange the summation to get

$$\begin{aligned}\sum_{k \in \mathbb{N}} \langle T e_k, e_k \rangle_U &= \sum_{j \in \mathbb{N}} \sum_{k \in \mathbb{N}} \langle e_k, a_j \rangle_U \cdot \langle e_k, b_j \rangle_U \\ &= \sum_{j \in \mathbb{N}} \langle a_j, b_j \rangle_U\end{aligned}$$

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## Definition

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A bounded linear operator  $T : U \rightarrow H$  is called **Hilbert-Schmidt** if

$$\sum_{k \in \mathbb{N}} \|Te_k\|^2 < \infty$$

where  $e_k, k \in \mathbb{N}$  is an orthonormal basis of  $U$ .

The space of all Hilbert-Schmidt operators from  $U$  to  $H$  is denoted by  $L_2(U, H)$ .

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$$\|T\|_{L_2(U,H)}^2 := \sum_{k \in \mathbb{N}} \|Te_k\|^2$$

*does not depend on the choice of the orthonormal basis  $e_k, k \in \mathbb{N}$ , we have that  $\|T\|_{L_2(U,H)} = \|T^*\|_{L_2(H,U)}$ . For simplicity we write  $\|T\|_{L_2}$  or  $\|T\|_{L_2(U,H)}$ .*

- $\|T\|_{L(U,H)} \leq \|T\|_{L_2(U,H)}$
- *Let  $G$  be another Hilbert space,  $S_1 \in L(U, G)$ ,  $S_2 \in L(G, U)$  and  $T \in L_2(U, H)$ . Then  $S_1 T \in L_2(U, G)$ ,  $TS_2 \in L_2(G, H)$  and*

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# Proof of the theorem

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$$\|S_1 T\|_{L_2(U,G)} \leq \|S_1\|_{L(H,G)} \|T\|_{L_2(U,H)}$$

$$\|TS_2\|_{L_2(G,H)} \leq \|T\|_{L_2(U,H)} \|S\|_{L(G,U)}$$

**Proof:** If  $e_k, k \in \mathbb{N}$  is an orthonormal basis of  $U$  and  $f_k, k \in \mathbb{N}$  is an orthonormal basis of  $H$ . We obtain by Parseval identity that

$$\begin{aligned} \sum_{k \in \mathbb{N}} \|Te_k\|^2 &= \sum_{k \in \mathbb{N}} \sum_{j \in \mathbb{N}} |\langle Te_k, f_j \rangle|^2 \\ &= \sum_{j \in \mathbb{N}} \|T^* f_j\|_U^2 \end{aligned}$$

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Let  $x \in U$  and  $f_k, k \in \mathbb{N}$  be an orthonormal basis of  $H$ , then we get

$$\begin{aligned}\|Tx\|^2 &= \sum_{k \in \mathbb{N}} \langle Tx, f_k \rangle^2 \\ &\leq \|x\|_U^2 \sum_{k \in \mathbb{N}} \|T^* f_k\|_U^2 \\ &= \|T\|_{L_2(U, H)}^2 \|x\|_U^2\end{aligned}$$

where we used Cauchy-Schwarz inequality and property (1).

Therefore, we showed that

$$\|T\|_{L(U, H)} \leq \|T\|_{L_2(U, H)}$$

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Furthermore, since  $(TS_2)^* = S_2^* T^*$ . From above and (1), we have  $TS_2 \in L_2(G, H)$  and

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# A lemma

## Lemma

Let  $S, T \in L_2(U, H)$  and  $e_k, k \in \mathbb{N}$  be an orthonormal basis of  $U$ . If we define

$$\langle T, S \rangle_{L_2} := \sum_{k \in \mathbb{N}} \langle Se_k, Te_k \rangle$$

we obtain that  $(L_2(U, H), \langle \cdot, \cdot \rangle_{L_2})$  is a separable Hilbert space.

If  $f_k, k \in \mathbb{N}$  is an orthonormal basis of  $H$ , we get that  $f_j \otimes e_k := f_j \langle e_k, \cdot \rangle_U, j, k \in \mathbb{N}$  is an orthonormal basis of  $L_2(U, H)$ .

**Proof:** Let us first show  $L_2(U, H)$  is complete.

Let  $T_n, n \in \mathbb{N}$  be a Cauchy sequence in  $L_2(U, H)$ . Then it is clear that it is also a Cauchy sequence in  $L(U, H)$ . Because of the completeness of  $L(U, H)$ , there exists an element  $T \in L(U, H)$ , such that  $\|T_n - T\|_{L(U, H)} \rightarrow 0$  as  $n \rightarrow \infty$ .

But by Fatou's lemma, we also have for any orthonormal basis  $e_k, k \in \mathbb{N}$  of  $U$  that

$$\begin{aligned} \|T_n - T\|_{L_2}^2 &= \sum_{k \in \mathbb{N}} \langle (T_n - T)e_k, (T_n - T)e_k \rangle \\ &= \sum_{k \in \mathbb{N}} \liminf_{m \rightarrow \infty} \|(T_n - T_m)e_k\|^2 \leq \liminf_{m \rightarrow \infty} \sum_{k \in \mathbb{N}} \|(T_n - T_m)e_k\|^2 \\ &= \liminf_{m \rightarrow \infty} \|T_n - T_m\|_{L_2}^2 < \epsilon \end{aligned}$$

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Now we will prove that  $L_2(U, H)$  is separable.

If we define  $f_j \otimes e_k := f_j \langle e_k, \cdot \rangle_U$ ,  $j, k \in \mathbb{N}$ , then it is clear that  $f_j \otimes e_k \in L_2(U, H)$  for all  $j, k \in \mathbb{N}$  and for arbitrary  $T \in L_2(U, H)$  we get

$$\langle f_j \otimes e_k, T \rangle_{L_2} = \sum_{n \in \mathbb{N}} \langle e_k, e_n \rangle_U \cdot \langle f_j, T e_n \rangle = \langle f_j, T e_k \rangle$$

Thus  $f_j \otimes e_k$  is an orthonormal system.

Since  $T = 0$  if  $\langle f_j \otimes e_k, T \rangle_{L_2} = 0$  for all  $j, k \in \mathbb{N}$ ,

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## Another lemma

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Let  $(G, \langle \cdot, \cdot \rangle_G)$  be a separable Hilbert space. If  $T \in L_2(U, H)$  and  $S \in L_2(H, G)$ , then  $ST \in L_2(U, G)$  and

$$\|ST\|_{L_2(U, G)} \leq \|S\|_{L_2} \|T\|_{L_2}$$

**Proof:** Let  $f_k$  be an orthonormal basis in  $H$ , then

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The proof is complete.

**Remark:** Let  $e_k, k \in \mathbb{N}$  be an orthonormal basis of  $U$ . If  $T \in L(U)$  is symmetric, nonnegative with  $\sum_{k \in \mathbb{N}} \langle T e_k, e_k \rangle < \infty$ , then  $T \in L_1(U)$ .

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## Third lemma

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Let  $L \in L(H)$  and  $B \in L_2(U, H)$ , then  $LBB^* \in L_1(H)$ ,  
 $B^*LB \in L_1(U)$  and we have

$$\operatorname{tr}LBB^* = \operatorname{tr}B^*LB$$

**Proof:** By previous theorem and lemma,  $LBB^* \in L_1(H)$  and  
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## Proof continues...

$$\begin{aligned} & \sum_{k \in \mathbb{N}} \sum_{n \in \mathbb{N}} \| \langle f_k, B e_n \rangle \cdot \langle f_k, L B e_n \rangle \| \\ & \leq \sum_{n \in \mathbb{N}} \left( \sum_{k \in \mathbb{N}} | \langle f_k, B e_n \rangle |^2 \right)^{1/2} \cdot \left( \sum_{k \in \mathbb{N}} | \langle f_k, L B e_n \rangle |^2 \right)^{1/2} \\ & = \sum_{n \in \mathbb{N}} \| B e_n \| \cdot \| L B e_n \| \\ & \leq \| L \|_{L(H)} \cdot \| B \|_{L_2}^2 \end{aligned}$$

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$$\begin{aligned}
 \operatorname{tr}LBB^* &= \sum_{k \in \mathbb{N}} \langle LBB^* f_k, f_k \rangle = \sum_{k \in \mathbb{N}} \langle B^* f_k, B^* L^* f_k \rangle_U \\
 &= \sum_{k \in \mathbb{N}} \sum_{n \in \mathbb{N}} \langle B^* f_k, e_n \rangle_U \cdot \langle B^* L^* f_k, e_n \rangle_U \\
 &= \sum_{n \in \mathbb{N}} \sum_{k \in \mathbb{N}} \langle f_k, Be_n \rangle \cdot \langle f_k, LBe_n \rangle \\
 &= \sum_{n \in \mathbb{N}} \langle Be_n, LBe_n \rangle \\
 &= \sum_{n \in \mathbb{N}} \langle e_n, B^* LBe_n \rangle_U \\
 &= \operatorname{tr}B^*LB
 \end{aligned}$$

The proof is complete.

## Gaussian measure

Consider two separable Hilbert spaces  $(U, \langle \cdot, \cdot \rangle_U)$  and  $(H, \langle \cdot, \cdot \rangle)$ . Denote  $\mathcal{B}(X)$  as the Borel  $\sigma$ -algebra of  $X$ .

### Definition

A probability measure  $\mu$  on  $(U, \mathcal{B}(U))$  is called **Gaussian** if for all  $v \in U$ , the bounded linear mapping  $V' : U \rightarrow \mathbb{R}$  defined by  $u \mapsto \langle u, v \rangle_U$ ,  $u \in U$  has a Gaussian law, i.e. for all  $v \in U$ , there exists  $m := m(v) \in \mathbb{R}$  and  $\sigma := \sigma(v) \in [0, \infty[$  such that if  $\sigma(v) > 0$

$$(\mu \circ (V')^{-1})(A) = \mu(V' \in A) = \frac{1}{\sqrt{2\pi\sigma^2}} \int_A e^{-\frac{(x-m)^2}{2\sigma^2}} dx$$

for all  $A \in \mathcal{B}(\mathbb{R})$ . If  $\sigma(v) = 0$ ,  $\mu \circ (V')^{-1} = \delta_{m(v)}$ .

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## Mean and Covariance

### Theorem

A measure  $\mu$  on  $(U, \mathcal{B}U)$  is Gaussian if and only if

$$\hat{m}u(u) := \int_U e^{i\langle u, v \rangle} \mu(dv) = e^{i\langle m, u \rangle - \frac{1}{2} \langle Qu, u \rangle}, \quad u \in U,$$

where  $m \in U$  and  $Q \in L(U)$  is nonnegative, symmetric with finite trace (trace class).

In this case  $\mu$  is denoted by  $N(m, Q)$  and  $m$  is called **mean** and  $Q$  is called **covariance**. The measure  $\mu$  is uniquely determined by  $m$  and  $Q$ .



## Mean and Covariance

### Theorem

Furthermore, for all  $h, g \in U$

$$\int \langle x, h \rangle_U \mu(dx) = \langle m, h \rangle_U,$$

$$\int (\langle x, h \rangle_U - \langle m, h \rangle_U)(\langle x, g \rangle_U - \langle m, g \rangle_U) \mu(dx) \\ = \langle Qh, g \rangle_U,$$

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# Representation of Gaussian random variables

## Lemma

*If  $Q \in L(U)$  is of trace class, then there exists an orthonormal basis  $e_k, k \in \mathbb{N}$  of  $U$  such that*

$$Qe_k = \lambda_k e_k, \quad \lambda_k \geq 0, k \in \mathbb{N}$$

*0 is the only accumulation point of  $(\lambda_k)_{k \in \mathbb{N}}$ .*

## Theorem

*Let  $m \in U$ ,  $Q \in L(U)$  of trace class. In addition, assume  $\{e_k\}$  is an orthonormal basis of  $U$  with eigenvectors of  $Q$  and corresponding eigenvalues  $\lambda_k$ .*

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*Then a  $U$ -valued random variable  $X$  on  $(\Omega, \mathcal{F}, P)$  is Gaussian with  $P \circ X^{-1} = N(m, Q)$  if and only if  $X = \sum_{k \in \mathbb{N}} \sqrt{\lambda_k} \beta_k e_k + m$ . Here  $\beta_k$  are independent real-valued random variables with  $P \circ \beta_k^{-1} = N(0, 1)$  for all  $k \in \mathbb{N}$  with  $\lambda_k > 0$ . The series converges in  $L^2(\Omega, \mathcal{F}, P; U)$ .*

**Proof:** Let  $X$  be a Gaussian random variable with mean  $m$  and covariance  $Q$ . Set  $\langle, \rangle = \langle, \rangle_U$ , then  $X = \sum_{k \in \mathbb{N}} \langle X, e_k \rangle e_k$  in  $U$ ,  $\langle X, e_k \rangle$  is normally distributed with mean  $\langle m, e_k \rangle$  and variance  $\lambda_k, k \in \mathbb{N}$  by lemma.



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## Proof continues...

Define

$$\beta_k := \begin{cases} \frac{\langle X, e_k \rangle - \langle m, e_k \rangle}{\sqrt{\lambda_k}} & \text{if } k \in \mathbb{N}, \lambda_k > 0 \\ 0 & \text{otherwise} \end{cases}$$

then  $P \circ \beta_k^{-1} = N(0, 1)$  and  $X = \sum_{k \in \mathbb{N}} \sqrt{\lambda_k} \beta_k e_k + m$ .

To prove the independence of  $\beta_k$ , take an arbitrary  $n \in \mathbb{N}$ ,  $a_k \in \mathbb{R}$ ,  $1 \leq k \leq n$  for

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$$\langle X, \sum_{k=1, \lambda_k \neq 0}^n \frac{a_k}{\sqrt{\lambda_k}} e_k \rangle + c$$

which is normally distributed since  $X$  is a Gaussian random variable. Thus  $\beta_k$  form a Gaussian family.

$$E(\beta_i \beta_j) = \frac{1}{\sqrt{\lambda_i \lambda_j}} E(\langle X - m, e_i \rangle \langle X - m, e_j \rangle)$$

$$\frac{1}{\sqrt{\lambda_i \lambda_j}} \langle Q e_i, e_j \rangle = \frac{\lambda_j}{\sqrt{\lambda_i \lambda_j}} \langle e_i, e_j \rangle$$

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## Proof continues...

$$\begin{aligned}\sum_{k=1}^n a_k \beta_k &= \sum_{k=1, \lambda_k \neq 0}^n \frac{a_k}{\sqrt{\lambda_k}} \langle X, e_k \rangle + c \\ &\langle X, \sum_{k=1, \lambda_k \neq 0}^n \frac{a_k}{\sqrt{\lambda_k}} e_k \rangle + c\end{aligned}$$

which is normally distributed since  $X$  is a Gaussian random variable. Thus  $\beta_k$  form a Gaussian family.

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Besides,  $\sum_{k=1}^n \sqrt{\lambda_k} \beta_k \mathbf{e}_k$ ,  $n \in \mathbb{N}$  converges in  $L^2(\Omega, \mathcal{F}, P; U)$  since the space is complete and

$$\begin{aligned} E(\|\sum_{k=m}^n \sqrt{\lambda_k} \beta_k \mathbf{e}_k\|^2) &= \sum_{k=m}^n \lambda_k E(\|\beta_k\|^2) \\ &= \sum_{k=m}^n \lambda_k \rightarrow 0 \quad \text{as } n, m \rightarrow \infty \end{aligned}$$

because  $\sum_{k \in \mathbb{N}} \lambda_k = \text{tr} Q < \infty$ . This completes one direction of the proof.

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# Proof continues...

Let  $e_k$  be an orthonormal basis of  $U$  such that  $Qe_k = \lambda_k e_k, k \in \mathbb{N}$ . Let  $\beta_k$  be a family of independent real-valued Gaussian random variables with mean 0 and variance 1. Then

$$\sum_{k=1}^n \sqrt{\lambda_k} \beta_k e_k + m, n \in \mathbb{N} \rightarrow x := \sum_{k \in \mathbb{N}} \sqrt{\lambda_k} \beta_k e_k + m$$

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## Proof continues...

Fix  $u \in U$ , we get

$$\left\langle \sum_{k=1}^n \sqrt{\lambda_k} \beta_k \mathbf{e}_k + m, u \right\rangle = \sum_{k=1}^n \sqrt{\lambda_k} \beta_k \langle \mathbf{e}_k, u \rangle + \langle m, u \rangle$$

is normally distributed for all  $n \in \mathbb{N}$  and the sequence converges in  $L^2(\Omega, \mathcal{F}, P)$ . This implies that the limit  $\langle X, u \rangle$  is also normally distributed.

$$\begin{aligned} E(\langle X, u \rangle) &= E\left(\sum_{k \in \mathbb{N}} \sqrt{\lambda_k} \beta_k \langle \mathbf{e}_k, u \rangle + \langle m, u \rangle\right) \\ &= \lim_{n \rightarrow \infty} E\left(\sum_{k=1}^n \sqrt{\lambda_k} \beta_k \langle \mathbf{e}_k, u \rangle\right) + \langle m, u \rangle = \langle m, u \rangle \end{aligned}$$

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## Proof continues...

and

$$\begin{aligned} & E((\langle X, u \rangle - \langle m, u \rangle)(\langle X, v \rangle - \langle m, v \rangle)) \\ &= \lim_{n \rightarrow \infty} E\left(\sum_{k=1}^n \sqrt{\lambda_k} \beta_k \langle e_k, u \rangle \sum_{k=1}^n \sqrt{\lambda_k} \beta_k \langle e_k, v \rangle\right) \\ &= \sum_{k \in \mathbb{N}} \lambda_k \langle e_k, u \rangle \langle e_k, v \rangle = \sum_{k \in \mathbb{N}} \langle Q e_k, u \rangle \langle e_k, v \rangle \\ &= \sum_{k \in \mathbb{N}} \langle e_k, Qu \rangle \langle e_k, v \rangle = \langle Qu, v \rangle \end{aligned}$$

for all  $u, v \in U$ . The proof is complete.

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$$\begin{aligned} & E((\langle X, u \rangle - \langle m, u \rangle)(\langle X, v \rangle - \langle m, v \rangle)) \\ &= \lim_{n \rightarrow \infty} E\left(\sum_{k=1}^n \sqrt{\lambda_k} \beta_k \langle e_k, u \rangle \sum_{k=1}^n \sqrt{\lambda_k} \beta_k \langle e_k, v \rangle\right) \\ &= \sum_{k \in \mathbb{N}} \lambda_k \langle e_k, u \rangle \langle e_k, v \rangle = \sum_{k \in \mathbb{N}} \langle Qe_k, u \rangle \langle e_k, v \rangle \\ &= \sum_{k \in \mathbb{N}} \langle e_k, Qu \rangle \langle e_k, v \rangle = \langle Qu, v \rangle \end{aligned}$$

for all  $u, v \in U$ . The proof is complete.



## Existence result

From the proof above, we have the following existence result for Gaussian measure.

### Corollary

*Let  $Q \in L(U)$  be trace class and  $m \in U$ . Then there exists a Gaussian measure  $\mu = N(m, Q)$  on  $(U, \mathcal{B}U)$ .*

# Q-Wiener Processes

## Definition

A  $U$ -valued stochastic process  $W(t)$ ,  $t \in [0, T]$  on probability space  $(\Omega, \mathcal{F}, P)$  is called a (standard) **Q-Wiener process** if

- $W(0) = 0$
- $W$  has  $P$ -a.s. continuous trajectories
- $W(t_1), W(t_2) - W(t_1), \dots, W(t_n) - W(t_{n-1})$  are independent for all  $0 \leq t_1 < \dots < t_n \leq T$ ,  $n \in \mathbb{N}$
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$$P_{\circ}(W(t) - W(s))^{-1} = N(0, (t-s)Q) \quad \text{for all } 0 \leq s \leq t \leq T$$

## Representation of $Q$ -Wiener processes

### Lemma

*Let  $e_k$  be an orthonormal basis of  $U$  consisting of eigenvectors of  $Q$  with corresponding eigenvalues  $\lambda_k, k \in \mathbb{N}$ . Then a  $U$ -valued stochastic process  $W(t), t \in [0, T]$  is a  $Q$ -Wiener process if and only if*

$$W(t) = \sum_{k \in \mathbb{N}} \sqrt{\lambda_k} \beta_k(t) e_k, \quad t \in [0, T]$$

*where  $\beta_k(t), k \in \{n \in \mathbb{N} | \lambda_n > 0\}$  are independent real-valued Brownian motion on  $(\Omega, \mathcal{F}, P)$ . The series converges in  $L^2(\Omega, \mathcal{F}, P, C([0, T], U))$ , thus has a  $P$ -a.s. continuous modification. In particular, for any  $Q$  as above there exists a  $Q$ -Wiener process on  $U$ .*

# Normal filtration

## Definition

A filtration  $\mathcal{F}_t$ ,  $t \in [0, T]$  on  $(\Omega, \mathcal{F}, P)$  is called **normal** if

- $\mathcal{F}_0$  contains all elements  $A \in \mathcal{F}$  with  $P(A) = 0$
- $\mathcal{F}_t = \mathcal{F}_{t+} = \bigcap_{s>t} \mathcal{F}_s$  for all  $t \in [0, T]$

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- $W(t)$ ,  $t \in [0, T]$  is adapted to  $\mathcal{F}_t, t \in [0, T]$
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# Existence of normal filtration $Q$ -Wiener processes

Define

$$\mathcal{N} := \{A \in \mathcal{F} \mid P(A) = 0\}, \quad \tilde{\mathcal{F}}_t := \sigma(W(s) \mid s \leq t)$$

$$\tilde{\mathcal{F}}_t^0 := \sigma(\tilde{\mathcal{F}}_t \cup \mathcal{N})$$

Then we get

$$\mathcal{F}_t := \bigcap_{s>t} \tilde{\mathcal{F}}_s^0, \quad t \in [0, T]$$

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# Existence of normal filtration $Q$ -Wiener processes

From above we have

## Lemma

*Let  $W(t)$ ,  $t \in [0, T]$  be an arbitrary  $U$ -valued  $Q$ -Wiener process on  $(\Omega, \mathcal{F}, P)$ . Then it is a  $Q$ -Wiener process w.r.t. the normal filtration  $\mathcal{F}_t$ ,  $t \in [0, T]$  defined as above.*

# Bochner integrable random variables

## Lemma

*Assume that  $E$  is a separable real Banach space. Let  $X$  be a Bochner integrable  $E$ -valued random variable defined on  $(\Omega, \mathcal{F}, P)$ . Let  $\mathcal{G}$  be a  $\sigma$ -field contained in  $\mathcal{F}$ .*

*Then there exists a unique Bochner integrable  $E$ -valued random variable  $z$  a.s., measurable with respect to  $\mathcal{G}$  such that*

$$\int_A X dP = \int_A z dP \quad \text{for all } A \in \mathcal{G}$$

The random variable  $z$  is denoted by  $E(X|\mathcal{G})$  and is called the conditional expectation of  $X$  given  $\mathcal{G}$ . Furthermore

$$\|E(X|\mathcal{G})\| \leq E(\|X\| | \mathcal{G}).$$

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## Property of conditional expectation

### Lemma

*Let  $(E_1, \varepsilon_1)$  and  $(E_2, \varepsilon_2)$  be two-separable spaces and  $\psi : E_1 \times E_2 \rightarrow \mathbb{R}$  a bounded measurable function. Let  $X_1, X_2$  be two random variables on  $(\Omega, \mathcal{F}, P)$  with values in  $(E_1, \varepsilon_1)$ ,  $(E_2, \varepsilon_2)$  respectively and let  $\mathcal{G} \subset \mathcal{F}$  be a fixed  $\sigma$ -field.*

*Assume that  $X_1$  is  $\mathcal{G}$ -measurable and  $X_2$  is independent of  $\mathcal{G}$ , then*

$$E(\phi(X_1, X_2)|\mathcal{G}) = \phi\hat{X}_1$$

*where  $\phi(\hat{x}_1) = E(\phi(x_1, x_2))$ ,  $x_1 \in E_1$ .*

# $\mathcal{F}_t$ -martingale

## Definition

Let  $M(t)$ ,  $t \geq 0$  be a stochastic process on  $(\Omega, \mathcal{F}, P)$  with values in a separable Banach space  $E$ , let  $\mathcal{F}_t, t \geq 0$  be a filtration on  $(\Omega, \mathcal{F}, P)$ . The process  $M$  is called an  **$\mathcal{F}$ -martingale**, if

- $E(\|M(t)\|) < \infty$  for all  $t \geq 0$
- $M(t)$  is  $\mathcal{F}_t$ -measurable for all  $t \geq 0$
- $E(M(t)|\mathcal{F}_s) = M(s)$   $P$ -a.s. for all  $0 \leq s \leq t < \infty$ .

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## Properties of $\mathcal{F}_t$ -martingale

### Lemma

*If  $M(t), t \geq 0$  is an  $E$ -valued  $\mathcal{F}_t$ -martingale and  $p \in [1, \infty)$ , then  $\|M(t)\|^p, t \geq 0$  is a real-valued  $\mathcal{F}_t$ -martingale.*

**Proof:** Since  $E$  is separable, there exists  $l_k \in E^*$ ,  $k \in \mathbb{N}$  such that  $\|z\| = \sup l_k(z)$  for all  $z \in E$ . Then for  $s < t$ ,

$$\begin{aligned} E(\|M_t\| | \mathcal{F}_s) &\geq \sup_k E(l_k(M_t) | \mathcal{F}_s) \\ &= \sup_k l_k(E(M_t | \mathcal{F}_s)) \\ &= \sup_k l_k(M_s) = \|M_s\| \end{aligned}$$

## Properties of $\mathcal{F}_t$ -martingale

### Lemma

*If  $M(t), t \geq 0$  is an  $E$ -valued  $\mathcal{F}_t$ -martingale and  $p \in [1, \infty)$ , then  $\|M(t)\|^p, t \geq 0$  is a real-valued  $\mathcal{F}_t$ -martingale.*

**Proof:** Since  $E$  is separable, there exists  $l_k \in E^*$ ,  $k \in \mathbb{N}$  such that  $\|z\| = \sup l_k(z)$  for all  $z \in E$ . Then for  $s < t$ ,

$$\begin{aligned} E(\|M_t\| | \mathcal{F}_s) &\geq \sup_k E(l_k(M_t) | \mathcal{F}_s) \\ &= \sup_k l_k(E(M_t | \mathcal{F}_s)) \\ &= \sup_k l_k(M_s) = \|M_s\| \end{aligned}$$

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This proves the lemma for  $p = 1$ . Jensen's inequality implies that for all  $p \in [1, \infty)$ ,

$$E(\|M_t\|^p | \mathcal{F}_s) \geq (E(\|M_t\| | \mathcal{F}_s))^p$$

Thus the lemma holds for any  $p \in [1, \infty)$ .

### Theorem

Let  $p > 1$ . Let  $E$  be a separable Banach space. If  $M(t), t \in [0, T]$  is a right-continuous  $E$ -valued  $\mathcal{F}_t$ -martingale, then

$$\begin{aligned} E\left(\sup_{t \in [0, T]} \|M(t)\|^p\right)^{1/p} &\leq \frac{p}{p-1} \sup_{t \in [0, T]} (E(\|M(t)\|^p))^{1/p} \\ &= \frac{p}{p-1} (E(\|M(T)\|^p))^{1/p} \end{aligned}$$

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## Square integrable martingales

Denote by  $\mathcal{M}_T^2(E)$  the space of all  $E$ -valued **continuous square integrable martingales**  $M(t), t \in [0, T]$ .

### Lemma

*The space  $\mathcal{M}_T^2(E)$  equipped with the norm*

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**Proof:** The continuity follows from the definition of  $Q$ -Wiener processes.

For each  $t \in [0, T]$ , we have

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# Proof of the lemma

Hence let  $0 \leq s \leq t \leq T$ ,  $A \in \mathcal{F}_s$ . By proposition,

$$\begin{aligned} \left\langle \int_A W(t) - W(s) dP, u \right\rangle_U &= \int_A \langle W(t) - W(s), u \rangle_U dP \\ &= P(A) \int \langle W(t) - W(s), u \rangle_U dP = 0 \end{aligned}$$

for all  $u \in U$  as  $\mathcal{F}_s$  is independent of  $W(t) - W(s)$  and

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# Elementary process

First we consider the following class of processes:

## Definition

An  $L = L(U, H)$ -valued process  $\phi(t), t \in [0, T]$  on  $(\Omega, \mathcal{F}, P)$  with normal filtration  $\mathcal{F}_t, t \in [0, T]$  is said to be **elementary** if there exists  $0 = t_0 < \dots < t_k = T, k \in \mathbb{N}$  such that

$$\phi(t) = \sum_{m=1}^{k-1} \phi_m \mathbf{1}_{]t_m, t_{m+1}]}(t), \quad t \in [0, T]$$

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Denote the space of elementary process defined as  $\varepsilon$ . Define

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# Hilbert-Schmidt

To show the mapping above is an isometry and extend the class  $\varepsilon$  to its completion, we recall the Hilbert-Schmidt operators:

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Let  $e_k, k \in \mathbb{N}$  be an orthonormal basis of  $U$ . An operator  $A \in L(U, H)$  is called **Hilbert-Schmidt** if

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## Important lemmas

### Lemma

*If  $Q \in L(U)$  is nonnegative and symmetric then there exists a unique  $Q^{1/2} \in L(U)$  nonnegative and symmetric such that  $Q^{1/2} \circ Q^{1/2} = Q$*

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## Remark

If two elementary processes  $\phi$  and  $\bar{\phi}$  belong to one equivalence class with respect to  $\|\cdot\|_{\mathcal{T}}$ , it does **not** follow that they are equal  $P_t$ -a.e. Because their values only have to correspond on  $Q^{1/2}(U)$   $P_t$ -a.e.

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## A Hilbert space

In this section we will seek an explicit representation of the completion  $\bar{\varepsilon}$ .

First, define  $U_0 := Q^{1/2}(U)$  with inner product

$$\langle u_0, v_0 \rangle_0 := \langle Q^{-1/2}u_0, Q^{-1/2}v_0 \rangle_U, \quad u_0, v_0 \in U_0$$

where  $Q^{-1/2}$  is the pseudo-inverse of  $Q^{1/2}$  in the case that  $Q$  is not one-to-one.

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The separable Hilbert space  $L_2(U_0, H)$  is called  $L_2^0$ . By proposition, we know  $Q^{1/2}g_k, k \in \mathbb{N}$  is an orthonormal basis of  $(U_0, \langle \cdot, \cdot \rangle_0)$  if  $g_k, k \in \mathbb{N}$  is an orthonormal basis of  $(\ker Q^{1/2})^\perp$ . This basis can be supplemented to a basis of  $U$  by elements of  $\ker Q^{1/2}$ .

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## Representation of $\bar{\varepsilon}$

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Define

$$\begin{aligned} \mathcal{N}_W^2(0, T; H) &:= \{ \phi : [0, T] \times \Omega \rightarrow L_2^0 \mid \phi \text{ is predictable and } \|\phi\|_t < \infty \} \\ &= L^2([0, T] \times \Omega, \mathcal{P}_T, dt \otimes P; L_2^0) \end{aligned}$$

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To prove  $\bar{\varepsilon} = \mathcal{N}_W^2$ , we need

- Since  $L(U, H)_0 \subset L_2^0$  and  $\phi \in \varepsilon$  is  $L_2^0$ -predictable, we have  $\varepsilon \subset \mathcal{N}_W^2$ .
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## Extension of stochastic integrals

Finally we will extend the definition of stochastic integrals from  $\mathcal{N}_W^2$  to

$$\mathcal{N}_W(0, T; H) :=$$

$$\left\{ \phi : \Omega_T \rightarrow L_2^0 \mid \phi \text{ is predictable with } P\left(\int_0^T \|\phi(s)\|_{L_2^0}^2 ds < \infty\right) = 1 \right\}$$

We call  $\mathcal{N}_W(0, T; H)$  the space of **stochastically integrable processes**.

## First lemma

Let  $W(t)$  be a  $Q$ -Wiener process,  $T > 0$ .

### Lemma

Let  $\phi$  be a  $L_2^0$ -valued stochastically integrable process.  
 $(\tilde{H}, \|\cdot\|_{\tilde{H}})$  is a separable Hilbert space and  $L \in L(H, \tilde{H})$ . Then

$L(\phi(t))$ ,  $t \in [0, T]$  is an element of  $\mathcal{N}_W(0, T; \tilde{H})$  and

$$L\left(\int_0^T \phi(t) dW(t)\right) = \int_0^T L(\phi(t)) dW(t), \quad P - a.s.$$



## Second lemma

### Lemma

Let  $\phi \in \mathcal{N}_W(0, T)$  and  $f$  is an  $\mathcal{F}_t$ -adapted continuous  $H$ -valued process. Set

$$\int_0^T \langle f(t), \phi(t) dW(t) \rangle := \int_0^T \tilde{\phi}_f(t) dW(t)$$

with  $\tilde{\phi}_f(t) := \langle f(t), \phi(t)u \rangle$ ,  $u \in U_0$ . Then this integral is well-defined as a continuous  $\mathbb{R}$ -valued stochastic process.

## Second lemma

### Lemma

More precisely,  $\tilde{\phi}_f$  is a  $P_T/\mathcal{B}(L_2(U_0, \mathbb{R}))$ -measurable map from  $[0, T] \times \Omega$  to  $L_2(U_0, \mathbb{R})$ ,  $\|\tilde{\phi}_f(t, \omega)\|_{L_2(U_0, \mathbb{R})} = \|\phi^*(t, \omega)f(t, \omega)\|_{U_0}$  for all  $(t, \omega) \in [0, T] \times \Omega$  and

$$\int_0^T \|\tilde{\phi}_f(t)\|_{L_2(U_0, \mathbb{R})}^2 dt \leq \sup_{t \in [0, T]} \|f(t)\| \int_0^T \|\phi(t)\|_{L_2}^2 dt < \infty \quad P\text{-a.s.}$$

## Third lemma

### Lemma

Let  $\phi \in \mathcal{N}_W(0, T)$  and  $M(t) := \int_0^t \phi(s) dW(s), t \in [0, T]$ . Define

$$\langle M \rangle_t := \int_0^t \|\phi(s)\|_{L_2^0}^2 ds, \quad t \in [0, T].$$

Then  $\langle M \rangle$  is the unique continuous increasing  $\mathcal{F}_t$ -adapted process starting at zero such that  $\|M(t)\|^2 - \langle M \rangle_t, t \in [0, T]$  is a local martingale.

## Third lemma

### Lemma

If  $\phi \in \mathcal{N}_W^2(0, T)$ , then for any sequence

$$I_l := \{0 = t_0^l < t_1^l < \cdots < t_{k_l}^l = T\}, \quad l \in \mathbb{N},$$

of partitions with  $\max_i(t_j^l - t_{j-1}^l) \rightarrow 0$  as  $l \rightarrow \infty$

$$\lim_{l \rightarrow \infty} E \left( \left| \sum_{t_{j+1}^l \leq t} \|M(t_{j+1}^l) - M(t_j^l)\|^2 - \langle M \rangle_t \right| \right) = 0$$

## Cylindrical Wiener processes

In case that  $Q$  is not of finite trace, we need a Hilbert space  $(U_1, \langle \cdot, \cdot \rangle_1)$  and a Hilbert-Schmidt embedding

$$J : (U_0, \langle \cdot, \cdot \rangle_0) \rightarrow (U_1, \langle \cdot, \cdot \rangle_1)$$

### Lemma

*Let  $e_k, k \in \mathbb{N}$  be an orthonormal basis of  $U_0 = Q^{1/2}(U)$  and  $\beta_k, k \in \mathbb{N}$  a family of independent real-valued Brownian motions. Define  $Q_1 := JJ^*$ . Then  $Q_1 \in L(U_1)$ ,  $Q_1$  is nonnegative definite and symmetric with finite trace and the series*

$$W(t) = \sum_{k=1}^{\infty} \beta_k(t) J e_k, \quad t \in [0, T], \quad (1)$$

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# Cylindrical Wiener processes

## Lemma

Moreover, we have that  $Q_1^{1/2}(U_1) = J(U_0)$  and for all  $u_0 \in U_0$

$$\|u_0\|_0 = \|Q_1^{-1/2}Ju_0\|_1 = \|Ju_0\|_{Q_1^{1/2}U_1}$$

i.e.  $J; U_0 \rightarrow Q_1^{1/2}U_1$  is an isometry.

# Stochastic integrals

Fix  $Q \in L(U)$  nonnegative, symmetric but not necessarily of finite trace. We integrate with respect to the standard  $U_1$ -valued  $Q_1$ -Wiener process given by the above lemma.

First we get a process  $\phi(t), t \in [0, T]$  is integrable with respect to  $W(t), t \in [0, T]$ , if it takes values in  $L_2(Q_1^{1/2}(U_1), H)$ , is predictable and if

$$P \left( \int_0^T \|\phi(s)\|_{L_2(Q_1^{1/2}(U_1), H)}^2 ds < \infty \right) = 1$$



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By previous lemma,  $Q_1^{1/2}(U_1) = J(U_0)$  and that

$$\langle Ju_0, Jv_0 \rangle_{Q_1^{1/2}(U_1)} = \langle Q_1^{-1/2}Ju_0, Q_1^{-1/2}Jv_0 \rangle_1 = \langle u_0, v_0 \rangle_0$$

for all  $u_0, v_0 \in U_0$ .

It follows that  $Je_k, k \in \mathbb{N}$  is an orthonormal basis of  $Q_1^{1/2}(U_1)$ .  
Hence

$$\phi \in L_2^0 = L_2(Q_1^{1/2}(U), H) \leftrightarrow \phi \circ J^{-1} \in L_2(Q_1^{1/2}(U_1), H)$$

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Define

$$\int_0^t \phi(s) dW(s) := \int_0^t \phi(s) \circ J^{-1} dW(s), \quad t \in [0, T]. \quad (2)$$

Then the class of all integrable processes is given by

$$\mathcal{N}_W = \left\{ \phi : \Omega_T \rightarrow L_2^0 \mid \phi \text{ predictable and } P \left( \int_0^T \|\phi(s)\|_{L_2^0}^2 ds < \infty \right) = 1 \right\}$$

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## Remark

- 1 The stochastic integral defined in the last slide is independent of the choice of  $(U_1, \langle, \rangle_1)$  and  $J$ . This follows by construction, since by (1) for elementary processes (2) does not depend on  $J$ .
- 2 If  $Q \in L(U)$  is trace class, the standard  $Q$ -Wiener process can also be considered as a cylindrical  $Q$ -Wiener process by setting  $J = I : U_0 \rightarrow U$  where  $I$  is the identity map. In this case both definitions of the stochastic integral coincide.

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