

Lec 11 05/04/22 (Tue)

①

- $W = K$ -cylindrical Wiener process  $\rightsquigarrow$  Filtration  $\mathcal{F} = \{\mathcal{F}_t : t \geq 0\}$
- For  $\phi: \mathbb{R}_+ \times \Omega \rightarrow K \otimes B$   $\mathcal{F}$ -adapted and simple,

“usual conditions”  
 $\left( \begin{array}{l} \mathcal{F}_t = \mathcal{F}_{t+} \quad \forall t \\ \mathcal{F}_0 \text{ contains all } \mathbb{P}\text{-null sets} \end{array} \right) \rightarrow \bigcap_{\varepsilon > 0} \mathcal{F}_{t+\varepsilon}$

$$\phi = \text{lin. Comb. of } \mathbb{I}_{(s,t] \times A} \otimes \begin{pmatrix} h \otimes b \\ \downarrow \quad \downarrow \\ K \quad B \end{pmatrix}$$

$s < t, \frac{0}{s}$

$$\Rightarrow \int_0^\infty \phi dW := \text{lin. Comb. of } \mathbb{I}_A W(\mathbb{I}_{(s,t]} \otimes h) b$$

P9, Lec 10

Lemma 2:  $B$  has  $MT_2$  &  $\phi$  simple adapted <sup>martingale type 2</sup>

Then  $\mathbb{E} \left[ \left\| \int_0^\infty \phi dW \right\|_B^2 \right] \lesssim \mathbb{E} \int_0^\infty \|\phi(t)\|_{\mathcal{H}(K,B)}^2 dt$  #3

Today: Thm (Doob's maximal inequality in  $MT_2$  setting)

(i) for  $\phi$  adapted, simple:  $\mathbb{E} \sup_{t \geq 0} \left\| \int_0^t \phi dW \right\|_B^2 \lesssim \text{RHS of } \textcircled{\#3}$

(ii) Extension for progressively measurable process with  $\mathbb{E} \int_0^\infty \|\phi(t)\|_{\delta(K, B)}^2 dt < \infty$  (#4)

$$M_t := \int_0^t \mathbb{I}_{[0, t]} \phi dW \in L^2(\Omega; C(\mathbb{R}_+; B))$$

def: we say  $\{\phi(s, \omega)\}_{s \in \mathbb{R}_+, \omega \in \Omega}$  is progressively measurable if  $\forall T \in (0, \infty)$ ,

$(s, \omega) \in [0, T] \times \Omega \mapsto \phi(s, \omega) \in \delta(K, B)$  is  $\mathcal{B}([0, T]) \otimes \mathcal{F}_T$ -measurable

In particular, simple, adapted processes are progressively measurable

generate the so-called predictable  $\sigma$ -algebra that is equivalent to the one generated by the adapted and (left) continuous processes.

dense in

$$L^2(\text{prog.}) = \left\{ \text{progressively measurable processes with } \textcircled{\#4} \right\}$$

(A predictable process is actually  $\{\mathcal{F}_{t-} : t \in \mathbb{R}_+\}$ -adapted,  $\mathcal{F}_t = \bigvee_{s < t} \mathcal{F}_s$ )

smallest  $\sigma$ -algebra that contains  $\bigcup_{s < t} \mathcal{F}_s$

(ii) follows by standard density argument. ch3 Ondreját '04 thesis.

proof of (i) From page 10, Lec 10, if  $\phi$  is simple, adapted process,

then  $\int_0^\infty \phi dW =$  finite-sum of martingale differences also with respect to the filtration generated by themselves.

$$= \sum_{k=1}^N d_k$$

i.e.  $\mathbb{E}[d_{k+1} | \sigma\{d_1, \dots, d_k\}] = 0.$

Then, by definition of  $\int_0^\infty \phi dW$  for simple, adapted process,

we can see that

$$\int_0^t \phi dW = \int_0^\infty \mathbb{1}_{[0,t]}(s) \phi(s) dW(s) = \sum_{k=1}^{N_t} d_k.$$

$N_t$  just some integer nondecreasing in  $t$ .

of the same form as in Lec 10, p 10

Then the inequality in (i) reduces to

$$\mathbb{E} \sup_{t \geq 0} \left\| \sum_{k=1}^{N_t} d_k \right\|_B^2 \approx \mathbb{E} \int_0^\infty \|\phi(t')\|_{\gamma(k, B)}^2 dt'$$

Lec 10 p 10

$$= \lim_{N \uparrow \infty} \sum_{k=1}^N \mathbb{E} \|d_k\|_B^2$$

Constant after some finite  $N$

discrete sum

important!

monotone cvg.

$\lim_{n \leq N} \mathbb{E} \sup_{k=1}^n \| \sum_{k=1}^n d_k \|_B^2$

It remains to show for each  $N \geq 1$ ,

$$\mathbb{E} \sup_{n \leq N} \left\| \sum_{k=1}^n d_k \right\|_B^2 \stackrel{\sim}{\leq} \sum_{k=1}^N \mathbb{E} \left[ \|d_k\|_B^2 \right]$$

indep. of  $N$

In fact,  $\left\{ X_n = \left\| \sum_{k=1}^n d_k \right\|_B : n \geq 1 \right\}$  is a real valued **sub-martingale**

$$\begin{aligned} \mathbb{E} \left[ X_{n+1} \mid d_1, \dots, d_n \right] &= \mathbb{E} \left[ \left\| \sum_{k=1}^{n+1} d_k \right\|_B \mid d_1, \dots, d_n \right] \\ &\stackrel{\geq}{\geq} \left\| \mathbb{E} \left[ \sum_{k=1}^{n+1} d_k \mid d_1, \dots, d_n \right] \right\|_B = X_n. \end{aligned}$$

Jensen  
p8, Pisier '86 book

Jacod protter "Probability Essentials"  
Thm 26.3

Then by Doob's  $L^2$ -inequality

$$\mathbb{E} \left( \max_{n \leq N} X_n^2 \right) \leq \left( \frac{2}{2-1} \right)^2 \mathbb{E} \left[ X_N^2 \right] \Rightarrow \mathbb{E} \sup_{n \leq N} \left\| \sum_{k=1}^n d_k \right\|_B^2 \leq 4 \mathbb{E} \left[ \left\| \sum_{k=1}^N d_k \right\|_B^2 \right] \stackrel{MT_2}{\leq} \sum_{k=1}^N \mathbb{E} \left[ \|d_k\|_B^2 \right]$$



"Localization": For  $\phi$  progressively measurable with  $\int_0^\infty \|\phi(t)\|_{\mathcal{H}(K,B)}^2 dt < \infty$  almost surely. (5)

(1) Define  $\tau_n = \inf \left\{ t \geq 0 : \int_0^t \left( 1 + \|\phi(s)\|_{\mathcal{H}(K,B)}^2 \right) ds \geq n \right\}$   $[\tau_n \uparrow + \infty]$

which is a stopping time.  $\Rightarrow \mathbb{I}_{[0, \tau_n]}$  is progressively-measurable.

(2) Define  $\int_0^T \phi dW := \int_0^\infty \underbrace{\mathbb{I}_{[0, \tau_n]} \phi}_{\text{progressively measurable}} dW$  on  $\underbrace{\{T \leq \tau_n\}}_{\text{an event with probability } \uparrow 1 \text{ as } n \rightarrow +\infty}$

Stochastic integral well defined.

so  $\int_0^T \phi dW$  is an almost surely well-defined object

$$(3) \quad M_t = \int_0^t \phi dW \Rightarrow M_{t \wedge \bar{\tau}_n} = \int_0^{t \wedge \bar{\tau}_n} \phi dW = \int_0^\infty \mathbb{I}_{[0, t \wedge \bar{\tau}_n]} \phi dW \in L^2(\Omega; \mathcal{B}) \quad (6)$$

Moreover,  $(M_{t \wedge \bar{\tau}_n} : t \geq 0)$  is a martingale with respect to the filtration  $(\mathcal{F}_{t \wedge \bar{\tau}_n} : t \geq 0)$ , for each  $n$ .

- $\bar{\tau}_n$  stopping time  $\Rightarrow$  so is  $\bar{\tau}_n \wedge t$
- For a IF-stopping time  $\bar{\tau}$ ,  $\mathcal{F}_{\bar{\tau}} = \{A : A \cap \{\bar{\tau} \leq t\} \in \mathcal{F}_t, \forall t \geq 0\}$  is a  $\sigma$ -algebra  
= all information up to the random time  $\bar{\tau}$ .

Lemma 3.6 in Ondreját '04 thesis

let  $\bar{\tau}$  be a IF-stopping time,  $\phi$  progressively-meas in  $\mathcal{X}(K, \mathcal{B})$

s.t.  $\mathbb{E} \int_0^\infty \|\phi(s)\|_{\mathcal{X}(K, \mathcal{B})}^2 ds < \infty,$

Then  $M_t = \int_0^t \phi(s) dW, t \geq 0$  defines a martingale and  $M_{t \wedge \bar{\tau}} = \int_0^t \mathbb{I}_{[0, \bar{\tau}]} \phi dW$  almost surely.

Thm. (BDG in  $MT_2$  setting)

04 Ondreját thesis

(7)

$0 < p < \infty$ ,  $B$  martingale type 2,  
separable Banach space,  $\exists C_{p,B} \in (0, \infty)$  such that

for any  $\mathbb{F}$ -adapted stopping time  $\tau$ ,  $F$  progressively measurable process in  $\mathcal{S}(K, B)$

we have:

$$\textcircled{\#5} \quad \mathbb{E} \sup_{0 \leq t \leq \tau} \left\| \int_0^t F(t') dW(t') \right\|_B^p \leq C_{p,B} \mathbb{E} \left[ \left( \int_0^\tau \|F(t')\|_{\mathcal{S}(K,B)}^2 dt' \right)^{\frac{p}{2}} \right].$$

$K$ -cylindrical Wiener process

Proof: Define  $M(\tau) = \left\| \int_0^\tau F(t') dW(t') \right\|_B$   $M_0 = 0$

running maximum  $\rightarrow$   $M^*(\tau) = \sup_{s \leq \tau} M(s)$ ,

$$N(\tau) = \left[ \int_0^\tau \|F(t')\|_{\mathcal{S}(K,B)}^2 dt' \right]^{\frac{1}{2}}$$

choose (later)  $\beta > 1$ ,  $\delta > 0$ ,  $\lambda > 0$ ,  $t \geq 0$ . (Good  $\lambda$  inequality, due to Burkholder '73)

define  $\tau_1 = \inf \{ \tau \geq 0 : M(\tau) \geq \beta \lambda \}$  }  $\Rightarrow \tau_2 < \tau_1$

$\tau_2 = \inf \{ \tau \geq 0 : M(\tau) \geq \lambda \}$

$\sigma = \inf \{ \tau \geq 0 : \underline{N(\tau) \geq \delta \lambda} \}$

$\rho_n = \inf \{ \tau \geq 0 : M(\tau) \geq n \} \Rightarrow M^*(t \wedge \rho_n) \leq n$

define  $A_1 = \{ \underline{M^*(t \wedge \rho_n) \geq \beta \lambda \text{ and } N(t) < \delta \lambda} \}$

$A_2 = \{ \left\| \int_0^{t \wedge \tau_1 \wedge \sigma \wedge \rho_n} F(t') dW(t') - \int_0^{t \wedge \tau_2 \wedge \sigma \wedge \rho_n} F dW \right\|_{B_t} \geq \lambda(\beta-1) \}$

On  $A_1$ :  $\tau_2 \leq \tau_1 \leq t \wedge \rho_n \leq t \leq \sigma$   
 $\{ M(\tau_1) = \beta \lambda, M(\tau_2) = \lambda \}$

(8)



$$\Rightarrow \begin{cases} t \wedge \bar{\sigma}_1 \wedge \sigma \wedge P_n = \bar{\sigma}_1 \\ t \wedge \bar{\sigma}_2 \wedge \sigma \wedge P_n = \bar{\sigma}_2 \end{cases} \Rightarrow \left\| \int_0^{t \wedge \bar{\sigma}_1 \wedge \sigma \wedge P_n} F dW - \int_0^{t \wedge \bar{\sigma}_2 \wedge \sigma \wedge P_n} F dW \right\|_B = \left\| \int_0^{\bar{\sigma}_1} F dW - \int_0^{\bar{\sigma}_2} F dW \right\|_B \quad (9)$$

$$\geq M(\bar{\sigma}_1) - M(\bar{\sigma}_2) = (\beta - 1)\lambda.$$

That is, we have  $A_1 \subseteq A_2$ .

Next ( $\bar{\sigma}_2 < \bar{\sigma}_1$ )

$$\mathbb{E} \left[ \left\| \int_0^{t \wedge \bar{\sigma}_1 \wedge \sigma \wedge P_n} F dW - \int_0^{t \wedge \bar{\sigma}_2 \wedge \sigma \wedge P_n} F dW \right\|_B^2 \right] = \mathbb{E} \left[ \left\| \int_0^t \mathbb{I}_{(\bar{\sigma}_2 \wedge \sigma \wedge P_n, \bar{\sigma}_1 \wedge \sigma \wedge P_n]}(s) F(s) dW \right\|_B^2 \right]$$

$$\stackrel{\substack{\uparrow \lambda < \\ \text{BDG}(p=2)}}}{\leq} \mathbb{E} \left[ \int_0^t \underbrace{\mathbb{I}_{(\bar{\sigma}_2 \wedge \sigma \wedge P_n, \bar{\sigma}_1 \wedge \sigma \wedge P_n]}(s) \|F(s)\|_{\delta(K, B)}^2 ds}_{\text{nonzero only when}} \right]$$

$$\stackrel{\substack{\text{ok to add the} \\ \text{above indicator}}}{\leq} \mathbb{E} \left[ \int_0^t \mathbb{I}_{M^*(t \wedge P_n) \geq \lambda} \|F(s)\|_{\delta(K, B)}^2 ds \right]$$

$$\leq \mathbb{E} \left[ \underbrace{\int_0^{t \wedge \delta} \|F(s)\|_{\delta(K, B)}^2 ds}_{\text{bounded by } \delta^2 \lambda^2 \text{ by def of } \sigma} \mathbb{1}_{M^*(t \wedge P_n) \geq \lambda} \right]$$

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$$\leq C, \delta^2 \lambda^2 \mathbb{P}(M^*(t \wedge P_n) \geq \lambda).$$

Therefore  $\mathbb{P}(A_1) \leq \mathbb{P}(A_2) \stackrel{\text{Chebyshev's ineq.}}{\leq} \frac{C_2 \delta^2 \lambda^2 \mathbb{P}(M^*(t \wedge P_n) \geq \lambda)}{\lambda^2 (\beta-1)^2}$

$$\begin{aligned} \Rightarrow \mathbb{P}(M^*(t \wedge P_n) \geq \beta \lambda) &\leq \mathbb{P}(M^*(t \wedge P_n) \geq \beta \lambda, N(t) < \delta \lambda) + \mathbb{P}(N(t) \geq \delta \lambda) \\ &\leq \mathbb{P}(N(t) \delta^{-1} \geq \lambda) + \mathbb{P}(A_1) \\ &\leq \mathbb{P}(N(t) \delta^{-1} \geq \lambda) + \frac{C_2 \delta^2}{(\beta-1)^2} \mathbb{P}(M^*(t \wedge P_n) \geq \lambda) \end{aligned}$$

Fact: (Layer-Cake theorem)

$$\text{For } Z \geq 0 \text{ in } L^p(\Omega): \mathbb{E}[Z^p] = \mathbb{E} \int_0^Z p \lambda^{p-1} d\lambda = \mathbb{E} \int_0^\infty \mathbb{I}_{[0, Z]}(\lambda) p \lambda^{p-1} d\lambda$$
$$= \int_0^\infty \mathbb{P}(Z \geq \lambda) p \lambda^{p-1} d\lambda$$



$$\mathbb{E} \left[ \left( M^*(t \wedge \tau_n) \beta^{-1} \right)^p \right] \leq \mathbb{E} \left[ \left( N(t) \delta^{-1} \right)^p \right] + \frac{C_1 \delta^2}{(\beta-1)^2} \mathbb{E} \left[ M^*(t \wedge \tau_n)^p \right]$$

$$\Leftrightarrow \mathbb{E} \left[ M^*(t \wedge \tau_n)^p \right] \leq \beta^p \delta^{-p} \mathbb{E} \left[ N(t)^p \right] + \underbrace{\frac{C_1 \beta^p \delta^2}{(\beta-1)^2}}_{= \frac{1}{2} \text{ for some } \delta \ll 1} \underbrace{\mathbb{E} \left[ M^*(t \wedge \tau_n)^p \right]}_{\text{bounded by } n^p < \infty \text{ so can be moved to LHS.}}$$

$$\Rightarrow \mathbb{E} [ M^*(t \wedge P_n)^p ] \lesssim \mathbb{E} [ N(t)^p ]$$

↑  
indep n.

Recall  $P_n \uparrow +\infty$  as  $n \uparrow +\infty$

the running maximum  $M^*(t \wedge P_n) \leq M^*(t)$

AND  $M^*(t \wedge P_n) \uparrow M^*(t)$ .

By monotone convergence theorem

$$\mathbb{E} [ M^*(t)^p ] \lesssim \mathbb{E} [ N(t)^p ]$$

That is,

$$\mathbb{E} \left[ \sup_{r \leq t} \left\| \int_0^r F(t') dW(t') \right\|_B^p \right] \leq C_{p,B} \mathbb{E} \left[ \int_0^t \|F(t')\|_{\mathcal{L}(K,B)}^2 dt' \right]^{p/2}$$

we can set  $t = \infty$   
 Also we can replace  $F$  by  $F \mathbb{I}_{[0, \tau]}$  → #5 P7.

