

WS 4:

Ex 1

X r.v. s.t. $E(|X|) < +\infty$. (\mathcal{F}_t) filtration
 $(Y_t = E(X | \mathcal{F}_t))_{t \geq 0}$ is a \mathcal{F}_t -martingale

$$\forall t \geq 0, E(|Y_t|) = E(|E(X | \mathcal{F}_t)|) \\ \leq E(E(|X| | \mathcal{F}_t)) = E(|X|) < +\infty.$$

$\forall t$, Y_t is \mathcal{F}_t -measurable by definition
 $E(X | \mathcal{F}_t)$ is \mathcal{F}_t -measurable

$$\forall t, \Delta > 0, E(Y_{t+\Delta} | \mathcal{F}_t) = Y_t.$$

$E(Y_{t+\Delta} | \mathcal{F}_t) = E(E(X | \mathcal{F}_{t+\Delta}) | \mathcal{F}_t)$
as $\mathcal{F}_t \subset \mathcal{F}_{t+\Delta}$ (because $(\mathcal{F}_t)_{t \geq 0}$ is a
filtration)

$$E(Y_{t+\Delta} | \mathcal{F}_t) = E(X | \mathcal{F}_t) = Y_t$$

\Rightarrow
 $(Y_t)_t$ is \mathcal{F}_t -martingale.

Ex 2: ①

(X_t) $\forall t > s$ $X_t - X_s \perp\!\!\!\perp \mathcal{G}(X_u, u \leq s)$
and $E(X_t) = E(X_0)$, $\forall t$

1) (X_t) is a martingale with respect to
 $\mathcal{F}_t^X = \mathcal{G}(X_s, s \leq t)$

• $\forall t$. $E(|X_t|) < \infty$ (integrable process)

• by def. X_t is \mathcal{F}_t^X measurable.

$$\begin{aligned} \forall s \geq 0 \quad E(X_{t+s} | \mathcal{F}_t) &= E(X_{t+s} - X_t + X_t | \mathcal{F}_t) \\ &= E(X_{t+s} - X_t | \mathcal{F}_t) + E(X_t | \mathcal{F}_t) \\ &= E(X_{t+s} - X_t) + X_t \\ &= \underbrace{E(X_{t+s})}_{E(X_0)} - \underbrace{E(X_t)}_{E(X_0)} + X_t = X_t \end{aligned}$$

$(X_t)_{t \geq 0}$ is a martingale with respect to \mathcal{F}_t^X .

② $t \geq 0$, $E(X_t^2) < \infty$ then $(X_t^2 - E(X_t^2))_{t \geq 0}$ is a martingale.

$$\begin{aligned} E(|X_t^2 - E(X_t^2)|) &\leq E(X_t^2) + E(X_t^2) \\ &= 2E(X_t^2) \end{aligned}$$

• $Y_t = f(X_t)$ so \mathcal{F}_t -measurable
 X_t is \mathcal{F}_t -measurable by def

Y_t is \mathcal{F}_t -measurable as it is a function of X_t .

$$\begin{aligned} \bullet E(Y_{t+s} | \mathcal{F}_t) &= E(X_{t+s}^2 - E(X_{t+s}^2) | \mathcal{F}_t) \\ &= E(X_{t+s}^2 | \mathcal{F}_t) - E(E(X_{t+s}^2) | \mathcal{F}_t) \\ &= E((X_{t+s} - X_t + X_t)^2 | \mathcal{F}_t) - E(X_{t+s}^2) \\ &= E((X_{t+s} - X_t)^2 + X_t^2 + 2X_t(X_{t+s} - X_t) | \mathcal{F}_t) \\ &\quad - E(X_{t+s}^2) \end{aligned}$$

one idea:
martingale property

$$= E(X_{t+s} - X_t)^2 + X_t^2 + 2 \underbrace{E(X_{t+s} - X_t) X_t | \mathcal{F}_t} - E(X_{t+s}^2)$$

$$X_t E(X_{t+s} - X_t)$$

$$= E(X_{t+s}) - E(X_t) = 0$$

$$= E(X_{t+s}^2 + X_t^2 - 2X_{t+s}X_t) + X_t^2 - E(X_{t+s}^2)$$

$$= E(X_t^2) - 2E(X_t X_{t+s}) + X_t^2$$

$$E(X_{t+s} X_t) = E(((X_{t+s} - X_t) + X_t) X_t)$$

$$= E((X_{t+s} - X_t) X_t + X_t^2)$$

$$= E((X_{t+s} - X_t) X_t) + E(X_t^2)$$

$$= E(X_{t+s} - X_t) E(X_t) + E(X_t^2)$$

$$E(X_{t+s} - X_t) = E(X_{t+s}) - E(X_t) = E(X_0) - E(X_0)$$

$$\text{So } E(Y_{t+s} | \mathcal{F}_t) = -2E(X_t^2) + X_t^2 + E(X_t^2)$$

$$= X_t^2 - E(X_t^2) = Y_t$$

Ex 4:

$$\mathcal{F}_t^{(B_1, B_2)} = \sigma \{ \underbrace{B_1(s)}_{r.v.}, \underbrace{B_2(s)}_{r.v.}, \Delta \leq t \}$$

$$X(t) = B_1(t) \cdot B_2(t)$$

$$E|X_t| = E|B_1(t) B_2(t)| = E|B_1(t)| \cdot E|B_2(t)|$$

indep.

$X(t)$ is $\mathcal{F}_t^{(B_1, B_2)}$ measurable as a function of elements $\mathcal{F}_t^{(B_1, B_2)}$ - measurable.

$$\forall s < t \quad E(X_t | \mathcal{F}_s) = X_s$$

$$E(X_t | \mathcal{F}_s) = E(B_1(t) \cdot B_2(t) | \mathcal{F}_s)$$

$$= E\left(\underbrace{[B_1(t) - B_1(s)]}_{\text{ind}} + B_1(s) \right) \underbrace{[B_2(t) - B_2(s)]}_{\text{ind}} + B_2(s) \Big| \mathcal{F}_s$$

$$= E\left((B_1(t) - B_1(s)) (B_2(t) - B_2(s)) \Big| \mathcal{F}_s \right)$$

$$+ E\left((B_1(t) - B_1(s)) B_2(s) \Big| \mathcal{F}_s \right)$$

$$\begin{aligned}
& + E(B_1(s) (B_2(t) - B_2(s)) | \mathcal{F}_s) \\
& + E(B_1(s) B_2(s) | \mathcal{F}_s) \\
& = E(B_1(t) - B_1(s)) E(B_2(t) - B_2(s)) \\
& + B_2(s) E(B_1(t) - B_1(s)) \\
& + B_1(s) E(B_2(t) - B_2(s)) \\
& + B_1(s) B_2(s) \\
& = B_1(s) B_2(s)
\end{aligned}$$

Ex 5:

B BM started in $x \in \mathbb{R} \Leftrightarrow B_0^x = x$.

$T_a = \inf \{ t \geq 0, B_t = a \}$

hitting time

$B_t^x \sim N(x, t)$ $B_t^x = x + B_t$ B_t is standard BM

1) B is a martingale (cf ex 2 (1))

A martingale $(M_t)_{t \geq 0}$ is uniformly integrable iff =

$$\lim_{A \rightarrow +\infty} \sup_{t \geq 0} E(M_t | \mathbb{1}_{\{|M_t| \leq A\}}) = 0$$

$$\begin{aligned}
& \forall t > 0, E\left(|B_t^n| \mathbb{1}_{\{|B_t^n| \geq A\}}\right) = \\
& = \int_{\mathbb{R}} |y| \mathbb{1}_{\{|y| \geq A\}} \frac{1}{\sqrt{2\pi t}} e^{-\frac{(y-n)^2}{2t}} dy \\
& = \frac{1}{\sqrt{2\pi t}} \left(\int_A^{+\infty} y e^{-\frac{(y-n)^2}{2t}} dy - \int_{-\infty}^{-A} y e^{-\frac{(y-n)^2}{2t}} dy \right) \\
& \quad z = y - n \\
& = \frac{1}{\sqrt{2\pi t}} \left[\int_{A-n}^{+\infty} (z+n) e^{-\frac{z^2}{2t}} dz - \int_{-\infty}^{-A-n} (z+n) e^{-\frac{z^2}{2t}} dz \right] \\
& = \frac{1}{\sqrt{2\pi t}} \left(\int_{A-n}^{+\infty} z e^{-\frac{z^2}{2t}} dz + \int_{-\infty}^{-A-n} -z e^{-\frac{z^2}{2t}} dz \right) \\
& \quad + \mathbb{X} \left[\underbrace{P(B_t^n \geq A-n)}_{\geq -A} - \underbrace{P(B_t^n \leq -A-n)}_{\geq -A} \right] \\
& \quad \cdot \frac{1}{\sqrt{2\pi t}} \int_{A-n}^{+\infty} z e^{-\frac{z^2}{2t}} dz = \frac{1}{\sqrt{2\pi t}} \left[-t e^{-\frac{z^2}{2t}} \right]_{A-n}^{+\infty} \\
& = \sqrt{\frac{t}{2\pi}} e^{-\frac{(A-n)^2}{2t}}
\end{aligned}$$

$$\forall A > 0, \sup_{t > 0} E\left(|B_t^n| \mathbb{1}_{\{|B_t^n| \geq A\}}\right) \geq$$

$$\sup_{t \geq 0} \left(\sqrt{\frac{t}{2\pi}} e^{-\frac{(A-x)^2}{2t}} \right) - 1 = +\infty$$

$t \rightarrow \infty \rightarrow 1$

(B^x) is not uniformly integrable.

(2) We can fix $t_0 > 0$ and consider $(B_{t \wedge t_0}^x)_{t \geq 0}$

For any martingale $(M_t)_{t \geq 0}$, the stopped martingale $(M_{t \wedge t_0})_{t \geq 0}$ is uniformly integrable.

Thus, we can use Doob's optional stopping time theorem with $(B_{t \wedge t_0}^x)_{t \geq 0}$

For any stopping time T ,
 $E(B_{T \wedge t_0}^x) = E(B_{0 \wedge t_0}^x) = E(B_0^x) = x$

$T = T_a \wedge T_b = \inf \{ t \geq 0, B_t = a \text{ or } B_t = b \}$

is a stopping time \Rightarrow

$$E(B_{T \wedge t_0}^x) = x, \forall t_0$$

$$\begin{aligned}
 E(B_{T_a \wedge T_b}^n) &= a \mathbb{P}(T_a < T_b) + b \mathbb{P}(T_a > T_b) \\
 &= a(1 - \mathbb{P}(T_b < T_a)) + b \mathbb{P}(T_a > T_b) \\
 &= (b-a) \cdot \mathbb{P}(T_b < T_a) + a.
 \end{aligned}$$

x • We want to have =

$$\lim_{t_0 \rightarrow +\infty} E(B_{T_a \wedge T_b \wedge t_0}^n) = E(B_{T_a \wedge T_b}^n)$$

the process
is always
bounded
above

we use dominated convergence theorem

As $T_a < +\infty$ a.s (course)

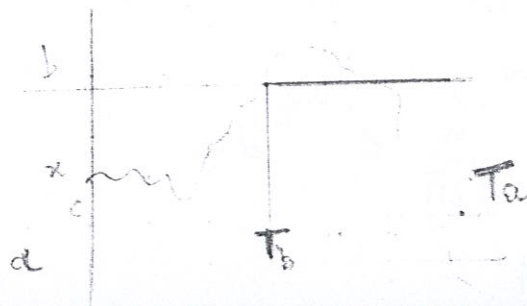
$T_a \wedge T_b < +\infty$ a.s and a.s $\lim_{t_0 \rightarrow \infty} B_{T_a \wedge T_b \wedge t_0}^n = B_{T_a \wedge T_b}^n$

• $\forall t_0, |B_{T_a \wedge T_b \wedge t_0}| \leq |a| + |b|$ deterministic

$$E[|a| + |b|] = |a| + |b| < +\infty.$$

⇒ According to the theorem

$$\lim_{t_0 \rightarrow +\infty} E(B_{T_a \wedge T_b \wedge t_0}^n) = E(B_{T_a \wedge T_b}^n)$$



$$x = E(B_{T_a \wedge T_b}^x) = a + (b-a) P(T_b < T_a)$$

$$\Rightarrow P(T_b < T_a) = \frac{n-a}{b-a}$$

3.a) part (2) of ex 2.

$$b) M_t = B_t^2 - t$$

$$\text{Doob's} = \Pi_{t \wedge T_a \wedge T_b} \quad (\text{TCM})$$

opt random = stopping time

$$E(M_{\tau} | F_0) = M_0$$

$$E(E(M_{\tau} | F_0)) = E(M_0)$$

$$E(M_{\tau}) = E(M_0)$$

03.11.14

3a) $t \leq T$

$$\begin{aligned} E(B_T^2 - t | F_s^B) &= E((B_s + (B_T - B_s))^2 - t | F_s^B) \\ &= B_s^2 + E((B_T - B_s)^2 | F_s^B) + 2B_s E(B_T - B_s | F_s^B) \\ &\quad - t = B_s^2 + t - \Delta - t = B_s^2 - \Delta \end{aligned}$$

$$\Delta = t - \Delta$$

knowing $B_s, (B_T - B_s)_{E \sim A}$ is a BM.