

# Föreläsning 7/11-13

## Section 5.8 Martingales

$$E(Y | X_1 = x_1, \dots, X_n = x_n) = g(x_1, \dots, x_n) = \int_{-\infty}^{\infty} y f_{Y|X_1, \dots, X_n}(y|x_1, \dots, x_n) dy$$

$$E(Y | \underbrace{X_1, \dots, X_n}_{\text{info that we know } X_1, \dots, X_n \text{ denoted } F_n}) = g(X_1, \dots, X_n)$$

$$\stackrel{\text{info that we know } X_1, \dots, X_n \text{ denoted } F_n}{=} E(Y | F_n)$$

Properties of conditional expectation:

1)  $E(aY_1 + bY_2 | F_n) = aE(Y_1 | F_n) + bE(Y_2 | F_n)$

2)  $E(Y | F_n) \geq 0$  if  $Y \geq 0$

3)  $E(Y | F_n) = Y$  if  $Y$  is determined by  $F_n = (X_1, \dots, X_n)$ ,  $Y$  is  $F_n$ -measurable.

4) (7 iboken)  $E(E(Y | F_n)) = E(Y) \quad (= E(g(X_1, \dots, X_n)))$

4)  $E(YZ | F_n) = ZE(Y | F_n)$  if  $Z$  is determined by  $X_1, \dots, X_n$ , if  $Z$  is  $F_n$ -measurable.

5)  $E(E(Y | F_n) | F_m) = E(Y | F_m)$  for  $m \leq n$

5)  $E(Y | F_n) = E(Y)$  if  $Y$  is independent of  $F_n$

8)  $E(g(Y) | F_n) \geq g(E(Y | F_n))$  for  $g$  convex function 

$$E(g(Y)) \geq g(E(Y))$$

example:  $g(x) = x^2$ ,  $E(X^2) \geq (E(X))^2$ ,  $\text{Var}(X) = E(X^2) - (E(X))^2 \geq 0$ .

Definition:

A discrete time random process  $(M_n, n \geq 0)$  is a martingale <sup>with respect to</sup> wrt  $F_n$  (where  $F_n$  is knowledge about  $M_1, \dots, M_n$  if nothing else is said) if

$$E(M_{n+1} | F_n) = M_n \quad \forall n$$

$$E(|M_n|) < \infty \quad \forall n$$

$$\Rightarrow E(M_{m+n} | F_n) = M_n \quad \text{for } m \geq 1, \text{ for } m=1 \text{ we are done, assume } m \geq 2$$

Proof: 6)  $\Rightarrow E(\underbrace{E(M_{m+n} | F_{m+n-1})}_{M_{m+n-1}} | F_n) = \dots = E(M_{n+1} | F_n) = M_n$

example

$X_n = Y_1 + \dots + Y_n$ ,  $n \geq 0$  where  $Y_1, Y_2, \dots$  independent r.v.'s

$F_n = \text{info } Y_1, \dots, Y_n$  is  $X_n$  martingale wrt  $F_n$ ?

$$E(X_{n+1} | F_n) = E(Y_1 + \dots + Y_n + Y_{n+1} | F_n) \stackrel{\text{IID}}{=} E(Y_1 + \dots + Y_n | F_n) + E(Y_{n+1} | F_n) =$$

$$= \{3\} + \{5\} = Y_1 + \dots + Y_n + E(Y_{n+1}) = X_n + E(Y_{n+1})$$

Answer:  $X_n$  martingale  $\iff E(Y_1) = E(Y_2) = \dots = 0$ .

example

$$X_n = e^{\sum_{i=1}^n Y_i} = E(e^{Y_i})^{-n} \quad Y_1, Y_2, \dots \text{ IID}, n \geq 0$$

$F_n$  is knowledge of  $Y_1, \dots, Y_n$  martingale?

$$E(X_{n+1} | F_n) = E\left(e^{\sum_{i=1}^{n+1} Y_i} \mid F_n\right) = E\left(\underbrace{e^{Y_1 + \dots + Y_n}}_{X_n} E(Y_{n+1})^{-1} e^{Y_{n+1}} \mid F_n\right)$$

$$= \{4\} = X_n E(e^{Y_{n+1}} E(e^{Y_1})^{-1} | F_n) = \{5\} = E(e^{Y_{n+1}}) E(e^{Y_1})^{-1} X_n =$$

$$= \{IID\} = \frac{E(e^{Y_1})}{E(e^{Y_1})} \cdot X_n = X_n$$

$= 1$

Answer: Yes it is martingale!

**Thm**  $E(M_n) = E(M_0)$  for martingale

**Proof**  $E(M_n) = \{7\} = E(E(M_n | F_{n-1})) = E(M_{n-1}) = \dots = E(M_0)$  ▀

**Thm** If  $M_n$  is martingale wrt  $F_n$  and  $T$  is a certain  $N = \{0, 1, \dots\}$ -valued random variable called stopping time s.t.

$E(T) < \infty$   
 $E(M_{T+1}) < \infty$   
 $\lim_{n \rightarrow \infty} E(|M_n| I_{\{T > n\}}) = 0$

} then  $E(M_T) = E(M_0)$

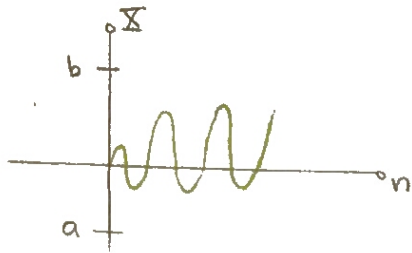
**Definition:** An  $N$ -valued r.v  $T$  is stopping time if  $\{T = n\}$  is  $F_n$ -measurable. That is we can decide if this event happens using info  $F_n$ .

example

$X_n = Y_1 + \dots + Y_n$  where  $Y_1, Y_2, \dots$  are IID with  $\begin{cases} P(Y_i = -1) = 1/2 \\ P(Y_i = 1) = 1/2 \end{cases}$

Take integers  $a < 0 < b$ ,  $T = \min\{n : X_n = a \text{ or } X_n = b\}$ .

$\xrightarrow{\text{forts.}}$



$$P(X_n \text{ reaches } b \text{ before } a) = ?$$

**Solution:**  $0 = E(X_0) = E(X_T) = P(X_n \text{ reaches } b \text{ before } a) \cdot b + (1 - P(X_n \text{ reaches } b \text{ before } a)) \cdot a$

$$\Rightarrow P(X_n \text{ reaches } b \text{ before } a) = \frac{-a}{b-a}$$

### Versions of martingales

submartingales:  $E(X_{n+1} | F_n) \geq X_n$

Supmartingales:  $E(X_{n+1} | F_n) \leq X_n$

example:  $X_n$  martingale, prove  $|X_n|$  is submartingale.  
 $E(|X_{n+1}| | F_n) \geq \{8\} \geq \underbrace{|E(X_{n+1} | F_n)|}_{X_n} = |X_n|$

$(M(t), t \geq 0)$  cont. time process,  $F_t = \text{know. about } (M_s)_{0 \leq s \leq t}$

$E(M_t | F_s) = M_s$  for  $s < t$ , cont. time mart.

example:  $M(t) = X(t) - \lambda t$  where  $X(t)$  is Poisson process with intensity  $\lambda$ .  
 $E(M(t) | F_s) = \underbrace{E(M(t) - M(s) | F_s)}_{\lambda t - \lambda t - (\lambda t - \lambda s)} + E(M(s) | F_s) = M(s) \quad (M_s = M(s))$