

Hsu Section 5.8: Martingales

More on conditional expectation

* In undergraduate probability one defines

$$E(Y | \mathcal{X}_0 = x_0, \dots, \mathcal{X}_n = x_n) = \begin{cases} \int_{-\infty}^{+\infty} y f_{Y | \mathcal{X}_0, \dots, \mathcal{X}_n}(y | x_0, \dots, x_n) dy & \text{for continuous r.v.'s} \\ \sum_{k=-\infty}^{+\infty} k f_{Y | \mathcal{X}_0, \dots, \mathcal{X}_n}(k | x_0, \dots, x_n) & \text{for discrete r.v.'s} \end{cases}$$

where

$$f_{Y | \mathcal{X}_0, \dots, \mathcal{X}_n}(y | x_0, \dots, x_n) = \frac{f_{Y, \mathcal{X}_0, \dots, \mathcal{X}_n}(y, x_0, \dots, x_n)}{f_{\mathcal{X}_0, \dots, \mathcal{X}_n}(x_0, \dots, x_n)}$$

and then we have

$$E(Y | \mathcal{X}_0 = x_0, \dots, \mathcal{X}_n = x_n) = g(x_0, \dots, x_n) \leftarrow \begin{matrix} \text{a function} \\ g: \mathbb{R}^M \rightarrow \mathbb{R} \text{ of } \bar{x} \end{matrix}$$

* In more advanced probability one defines

$$E(Y | \mathcal{X}_0, \dots, \mathcal{X}_n) = g(\mathcal{X}_0, \dots, \mathcal{X}_n) \leftarrow \begin{matrix} \text{a random variable} \\ \text{where } g \text{ is as above} \end{matrix}$$

This can be viewed as the information that $(\mathcal{X}_0, \dots, \mathcal{X}_n) = (x_0, \dots, x_n)$ is not inserted but one keeps the random values of $\mathcal{X}_0, \dots, \mathcal{X}_n$ themselves and that is the information conditioned on.

* We use the short hand notation F_n for the information $\mathcal{X}_0, \dots, \mathcal{X}_n$ and we also use $\sigma(\mathcal{X}_0, \dots, \mathcal{X}_n)$ to denote information $F_n = \sigma(\mathcal{X}_0, \dots, \mathcal{X}_n)$

* $E(Y | X_0, \dots, X_n) = E(Y | \sigma(X_0, \dots, X_n)) = E(Y | F_n)$
 $= g(X_0, \dots, X_n)$

where

$$g(x_0, \dots, x_n) = E(Y | X_0 = x_0, \dots, X_n = x_n)$$

* A r.v. Y is called F_n -measurable if it can be written as a function of X_0, \dots, X_n .

Thm

- ① $E(aY_1 + bY_2 | F_n) = aE(Y_1 | F_n) + bE(Y_2 | F_n)$
- ② $E(Y | F_n) \geq 0$ for $Y \geq 0$
- ③ $E(Y | F_n) = Y$ for Y F_n -measurable
- ④ $E(ZY | F_n) = ZE(Y | F_n)$ for Z F_n -measurable
- ⑤ $E(Y | F_n) = E(Y)$ for Y independent of F_n
- ⑥ $E(E(Y | F_n) | F_m) = E(Y | F_m)$ for $m \leq n$
- ⑦ $E(E(Y | F_n)) = E(Y)$
- ⑧ $E(g(Y) | F_n) \geq g(E(Y | F_n))$ for g convex

Proof

- ①-② and ⑧ proved as for ordinary expectation.
- ③-⑤ by inspection of definition.
- ⑥ Omitted.
- ⑦
$$\int_{x_0=-\infty}^{x_0=+\infty} \int_{x_n=-\infty}^{x_n=+\infty} \int_{y=-\infty}^{y=+\infty} y \frac{f_{Y, X_0, \dots, X_n}(y, x_0, \dots, x_n)}{f_{X_0, \dots, X_n}(x_0, \dots, x_n)} f_{X_0, \dots, X_n}(x_0, \dots, x_n) dy dx = E(Y)$$

* Until further notice $\{X_n\}_{n=0}^{+\infty}$ is a discrete time random process, $F_n = \sigma(X_0, \dots, X_n)$ and we call $\{F_n\}_{n=0}^{+\infty}$ a filtration. (Not important really.)

Def

- $\{X_n\}_{n=0}^{+\infty}$ is a martingale wrt. $\{F_n\}_{n=0}^{+\infty}$ if
- ① $E(|X_n|) < \infty$ for $n \geq 0$
- ② $E(X_{n+1} | F_n) = X_n$ for $n \geq 0$

Def $\{X_n\}_{n=0}^{+\infty}$ is a submartingale (supermartingale)
wrt. $\{F_n\}_{n=0}^{+\infty}$ if
 $E\{|X_n|\} < \infty$ for $n \geq 0$
 $E(X_{n+1} | F_n) \geq (\leq) X_n$ for $n \geq 0$

Thm For $\{X_n\}_{n=0}^{+\infty}$ a martingale
 ① $E(X_n) = E(X_0)$ for $n \geq 0$
 ② $E(X_{m+n} | F_n) = X_n$ for $m \geq 1$
 ③ $\{g(X_n)\}_{n=0}^{+\infty}$ is a submartingale for g convex

Proof ① $E(X_{n-1}) = E(E(X_n | F_{n-1})) = E(X_n)$ for $n \geq 1$
 ② $E(X_{m+n} | F_n) = E(E(X_{m+n} | F_{m+n-1}) | F_n) = E(X_{m+n-1} | F_n)$
 $= \dots = E(X_{n+1} | F_n) = X_n$ for $m \geq 2$
 ③ $E(g(X_{n+1}) | F_n) \geq g(E(X_{n+1} | F_n)) = g(X_n) \neq$

Def A $\mathbb{N} \cup \{+\infty\}$ -valued r.v. T is a stopping time
if $\{T=n\}$ is F_n -measurable (i.e., if one can say
whether $\{T=n\}$ or not using information $\sigma(X_0, \dots, X_n)$)

Thm (Optional Stopping Theorem) For $\{X_n\}_{n=0}^{+\infty}$ a
martingale and T a stopping time such that
 ① $E(T) < \infty$
 ② $E(|X_T|) < \infty$
 ③ $\lim_{n \rightarrow \infty} E(|X_n| 1_{\{T > n\}}) = 0$
 we have
 $E(X_T) = E(X_0)$

Proof Difficult!

Example (Computational exercise 3). Let

$$X_0 = 100 \text{ and } X_n = X_0 + \sum_{i=1}^n Y_i \text{ for } n \geq 1$$

where Y_1, \dots, Y_n are iid random variables with

$$P(Y_i = 4) = \frac{1}{5} \text{ and } P(Y_i = -1) = \frac{4}{5}$$

Then we have

$$\begin{aligned} E(X_{n+1} | F_n) &= E(Y_{n+1} + X_n | F_n) = E(Y_{n+1} | F_n) + E(X_n | F_n) \\ &= E(Y_{n+1}) + X_n = 4 \cdot \frac{1}{5} + (-1) \cdot \frac{4}{5} + X_n = X_n \end{aligned}$$

so X_n is martingale. Letting

$$T = \min \{n \geq 1 : X_n = 0 \text{ or } X_n \geq 200\}$$

it is clear that T is a stoppingtime and that conditions ①-③ of optional stopping theorem holds. (See also extra material on exercise web-page.) Hence

$$\begin{aligned} 100 &= E(X_0) = E(X_T) = E(X_T 1_{\{X_T=0\}} + X_T 1_{\{X_T \geq 200\}}) \\ &= 0 + E(X_T 1_{\{X_T \geq 200\}}) \begin{cases} \leq 203 \cdot P\{X_T \geq 100\} \\ \geq 200 \cdot P\{X_T \geq 100\} \end{cases} \end{aligned}$$

so that $P\{X_T \geq 200\} \in \left[\frac{100}{203}, \frac{100}{200} \right]$.

Continuous time martingales

[5]

Def A continuous time random process $\{X(t)\}_{t \geq 0}$ is a martingale wrt. the filtration $\{\mathcal{F}_t\}_{t \geq 0}$ where \mathcal{F}_t is the information $\sigma(\{X(s)\}_{s \in [0, t]})$ of the history of the process up to time t if

- ① $E(|X(t)|) < \infty$ for $t \geq 0$
- ② $E(X(t) | \mathcal{F}_s) = X(s)$ for $0 \leq s \leq t$

Thm $E(X(t)) = E(X(0))$ for $t \geq 0$
 $g(X(t))$ is a submartingale for g convex

Proof Same as before. #