

## The Markov Property

Let  $(\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{t \in T}, P)$  be a filtered probability space,  
(with  $T = \mathbb{N}$  or  $T = \mathbb{R}$ ).

An adapted stochastic process  $X_t: (\Omega, \mathcal{F}_t) \rightarrow (S, \mathcal{B})$   
satisfies the **Markov property** if, for  $f \in \mathcal{B}(S, \mathcal{B})$ ,

$$\mathbb{E}[f(X_t) | \mathcal{F}_s] = \mathbb{E}[f(X_t) | X_s] \text{ a.s. for all } s < t \text{ in } T.$$

This is equivalent to

$$\mathbb{E}[f(X_t) | \mathcal{F}_s] = F(X_s) \text{ a.s. for some } F \in \mathcal{B}(S, \mathcal{B})$$

$\Rightarrow$  Doob-Dynkin ✓

$$\begin{aligned} \Leftarrow & \text{ If } \mathbb{E}_{\mathcal{F}_s}[f(X_t)] = F(X_s) = \mathbb{E}_{\sigma(X_s)}[F(X_s)] = \mathbb{E}_{\sigma(X_s)}[\mathbb{E}_{\mathcal{F}_s}[f(X_t)]] \\ & = \mathbb{E}_{\sigma(X_s)}[f(X_t)] = \mathbb{E}[f(X_t) | X_s]. \end{aligned}$$



Let  $\mathcal{F}_t^X := \sigma(X_s : s \leq t)$ . (X. is adapted iff  $\mathcal{F}_t^X \subseteq \mathcal{F}_t \quad \forall t$ )

Earlier, we derived a Markov property

$$\rightarrow \mathbb{E}[f(X_t) | \mathcal{F}_s^X] = \mathbb{E}[f(X_t) | X_s] \quad \forall f \in \mathcal{B}(S, \mathcal{B}), s < t.$$

$$\mathbb{E}[f(X_t) | \mathcal{F}_s] = \mathbb{E}[f(X_t) | X_s] \quad " \quad s < t.$$

This is implied by the  $(\mathcal{F}_t)_{t \in T}$ -defined Markov property.

Pf.  $\mathbb{E}[f(X_t) | \mathcal{F}_s^X] = \mathbb{E}_{\mathcal{F}_s^X}[\mathbb{E}_{\mathcal{F}_s}[f(X_t)]] = \mathbb{E}_{\mathcal{F}_s^X}[\mathbb{E}_{\sigma(X)}[f(X_t)]] = \mathbb{E}_{\sigma(X)}[f(X_t)].$

///

Note: if  $T$  is countable, it suffices to state the Markov property

in the form  $\mathbb{E}[f(X_{n+1}) | \mathcal{F}_n] = \mathbb{E}[f(X_{n+1}) | X_n]$  base case  $k=n$

Induction: Suppose  $\mathbb{E}[f(X_n) | \mathcal{F}_k] = \mathbb{E}[f(X_n) | X_k]$  for some  $n \geq k$ .

$$\mathbb{E}_{\mathcal{F}_n}[f(X_{n+1})] = F(X_n) \quad \therefore \mathbb{E}_{\mathcal{F}_k}[f(X_{n+1})] = \mathbb{E}_{\mathcal{F}_k}[\mathbb{E}_{\mathcal{F}_n}[f(X_{n+1})]] = \mathbb{E}_{\mathcal{F}_k}[F(X_n)]$$

$$\therefore \mathbb{E}[f(X_{n+1}) | X_k] = \mathbb{E}_{\sigma(X_k)}[\mathbb{E}_{\mathcal{F}_n}[f(X_{n+1})]] = \mathbb{E}[f(X_{n+1}) | X_k].$$

///

The Markov property is about the present vs. the past.

But it also tells us about the future.

**Def:** Given a stochastic process  $(X_t)_{t \in T}$ , the future  $\sigma$ -field  $\mathcal{F}_{\geq s}^X$  is defined as

$$\mathcal{F}_{\geq s}^X := \sigma(X_t : t \geq s)$$

(It is a reverse filtration: if  $s_1 \leq s_2$ ,  $\mathcal{F}_{s_1}^X \supseteq \mathcal{F}_{s_2}^X$ .)

**Prop:** Let  $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in T}, P)$  be a filtered probability space, and let  $(X_t)_{t \in T}$  be an adapted stochastic process satisfying the Markov property. Then for  $s \in T$ ,

$$E[Y | \mathcal{F}_s] = E[Y | X_s] \quad \forall Y \in \mathcal{B}(\Omega, \mathcal{F}_{\geq s}^X).$$

Eg. If  $g_0, g_1, \dots, g_n \in \mathcal{B}(S, \mathcal{M})$ ,  $s = t_0 < t_1 < \dots < t_n$ ,

$$Y = g_0(X_{t_0}) g_1(X_{t_1}) \cdots g_n(X_{t_n}) \in \mathcal{B}(\Omega, \mathcal{F}_{\geq s}^X)$$

Lemma: Let  $\mathbb{M} = \{ g_0(X_{t_0}) g_1(X_{t_1}) \dots g_n(X_{t_n}) : n \in \mathbb{N}, s = t_0 < t_1 < \dots < t_n, g_j \in \mathcal{B}(S, \mathcal{B}) \}$ .

Then  $\mathbb{M}$  is a multiplicative system, and  $\sigma(\mathbb{M}) = \mathcal{F}_{\geq s}^X$ .

Pf. Multiplicative system ✓

$$\text{If } A \in \mathcal{F}_{\geq s}^X = \sigma(X_t : t \geq s) \quad A = X_t^{-1}(B), B \in \mathcal{B}, \\ t \geq s. \\ = (\cup_{t \geq s} X_t)^{-1}(1) \in \sigma(\mathbb{M}).$$

If  $A \in \sigma(\mathbb{M}) \therefore A = Y^{-1}(E), Y \in \mathbb{M}, E \in \mathcal{B}(\mathbb{R})$

$\uparrow$   
function of  $X_{t_0}, X_{t_1}, \dots, X_{t_n} \quad t_0, \dots, t_n \geq s$

///

Lemma: Let  $\mathbb{H} = \{ Y \in \mathcal{B}(\Omega, \mathcal{F}_{\geq s}^X) : \mathbb{E}[Y | \mathcal{F}_s] = \mathbb{E}[Y | X_s] \}$ .

Then  $\mathbb{H}$  is a subspace, contains 1, and is closed under bounded convergence.

Pf. Subspace: Linearity of  $\mathbb{E}_Y$   $\mathbb{H} = \mathcal{F}_s$  or  $\mathbb{H} = \sigma(X_s)$

Contains 1:  $b \in \mathbb{H} \quad \mathbb{E}_Y[1] = 1$ . ✓

///

Closed under bounded convergence: cDCT. ✓

Prop: Let  $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in T}, P)$  be a filtered probability space, and let  $(X_t)_{t \in T}$  be an adapted stochastic process satisfying the Markov property. Then for  $s \in T$ ,

$$E[Y | \mathcal{F}_s] = E[Y | X_s] \quad \forall Y \in \mathcal{B}(\Omega, \mathcal{F}_{\geq s}^X).$$

Pf. WTS  $H = \mathcal{B}(\Omega, \mathcal{F}_{\geq s}^X)$ . By the lemmas, suffices to show  $M \subseteq H$ , where

$$M = \{g_0(X_{t_0}) g_1(X_{t_1}) \dots g_n(X_{t_n}) : n \in \mathbb{N}, s = t_0 < t_1 < \dots < t_n, g_j \in \mathcal{B}(S, \mathcal{B})\}.$$

$$\begin{aligned} E_{\mathcal{F}_s}[Y] &= E_{\mathcal{F}_s}\left[E_{\mathcal{F}_{t_{n-1}}}[Y]\right] \\ &= E_{\mathcal{F}_s}\left[g_0(X_{t_0}) \dots g_{n-1}(X_{t_{n-1}}) \underbrace{h(X_{t_n})}_{\tilde{g}_{n-1}(X_{t_{n-1}})}\right] \end{aligned}$$

repeat ...

$$= E_{\mathcal{F}_s}[F(X_{t_0})] = F(X_s)$$

$$\therefore E_{\mathcal{F}_s}[Y] = F(X_s) = E[F(X_s) | X_s] = E[Y | X_s].$$

## Conditional Independence

Let  $(\Omega, \mathcal{F}, P)$  be a probability space;  $\mathcal{A}, \mathcal{B}, \mathcal{G} \subseteq \mathcal{F}$  sub- $\sigma$ -fields.

Say that  $\mathcal{A}, \mathcal{B}$  are **conditionally independent given  $\mathcal{G}$**  if

$$P(A \cap B | \mathcal{G}) = P(A | \mathcal{G}) P(B | \mathcal{G}) \text{ a.s. } \forall A \in \mathcal{A}, B \in \mathcal{B}$$

$$\mathbb{E}[1_A 1_B | \mathcal{G}]$$

Equivalently:  $\mathbb{E}[XY | \mathcal{G}] = \mathbb{E}[X | \mathcal{G}] \mathbb{E}[Y | \mathcal{G}]$  a.s.  $\forall X \in \mathcal{B}(\Omega, \mathcal{A}), Y \in \mathcal{B}(\Omega, \mathcal{B})$

→ Eg. Follows that, if  $C \in \mathcal{G}$ ,  $P(C) > 0$ ,

$$P(A \cap B | C) = P(A | C) P(B | C) \quad \forall A \in \mathcal{A}, B \in \mathcal{B}. \quad [\text{Why?}]$$

Eg. Let  $X, Y$  be iid with law  $\frac{1}{2}\delta_1 + \frac{1}{2}\delta_2$ . Then  $P(X=Y) = \frac{1}{2}$ .

$$\mathbb{E}[X | X=Y] = \mathbb{E}[Y | X=Y] = \frac{\mathbb{E}[X 1_{X=Y}]}{P(X=Y)} = \frac{1 \cdot \frac{1}{4} + 2 \cdot \frac{1}{4}}{\frac{1}{2}} = \frac{3}{2}$$

$$\mathbb{E}[XY | X=Y] = \frac{\mathbb{E}[XY 1_{X=Y}]}{P(X=Y)} = \frac{1 \cdot 1 \cdot \frac{1}{4} + 2 \cdot 2 \cdot \frac{1}{4}}{\frac{1}{2}} = \frac{5}{4}$$

$$\frac{3}{4} \neq \frac{3}{2} \cdot \frac{3}{2}.$$

This gives us a "poetic" way to rephrase the Markov property.

**Theorem:** The Markov property says:

"Conditioned on the present, the past and the future are independent."

More precisely: Let  $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in T}, P)$  be a filtered probability space, and let  $X_t: (\Omega, \mathcal{F}_t) \rightarrow (S, \mathcal{B})$  be an adapted stochastic process. Then

$(X_t)_{t \in T}$  satisfies the Markov property iff, for each  $s \in T$ ,

$\mathcal{F}_s$  and  $\mathcal{F}_{s+}^X$  are conditionally independent given  $\sigma(X_s)$ .

Pf. ( $\Rightarrow$ ) Let  $Z \in \mathcal{B}(\Omega, \mathcal{F}_s)$ ,  $Y \in \mathcal{B}(\Omega, \mathcal{F}_{s+}^X)$ .

$$\mathbb{E}[ZY | \mathcal{F}_s] = Z \mathbb{E}[Y | \mathcal{F}_s] = Z \mathbb{E}[Y | X_s]$$

$$\begin{aligned}\therefore \mathbb{E}_{\sigma(X_s)}[\mathbb{E}[ZY | \mathcal{F}_s]] &= \mathbb{E}_{\sigma(X_s)}[Z \mathbb{E}_{\sigma(X_s)}[Y]] \\ &= \mathbb{E}_{\sigma(X_s)}[Z] \cdot \mathbb{E}_{\sigma(X_s)}[Y].\end{aligned}$$

$$\mathbb{E}_{\sigma(X_s)}[ZY].$$

( $\Leftarrow$ ) [HW].