

## Lecture 2 . Exercise 4

changes we need to introduce to the nonlinear filtering recursion, when the Markov kernels are allowed to depend on  $j$  and on  $\mathcal{F}_{j-2}^X = \sigma(Y_0, \dots, Y_{j-2})$ .

Let  $Y_{[0,j]} := (Y_0, Y_1, \dots, Y_j)$ .

We have

$$P(X_j \in B \mid \mathcal{F}_{j-2}^X \cup \mathcal{F}_{j-2}^Y) = \int_B \Delta_j(X_{j-2}, Y_{[0,j-2]}, du)$$

and

$$P(Y_j \in B \mid \mathcal{F}_j^X \cup \mathcal{F}_{j-2}^Y) = \int_B Y_j(X_j, Y_{[0,j-2]}, u) V(du)$$

Note that now  $(X_j, Y_j)$  is not a Markov process.

The law of  $Y_{[0,j]}$  given  $\mathcal{F}_j^X$  is

$$\begin{aligned} P(Y_0 \in A_0, \dots, Y_j \in A_j \mid \mathcal{F}_j^X) &= E[1_{\{Y_0 \in A_0\}} \dots 1_{\{Y_{j-2} \in A_{j-2}\}} \int_{A_j} Y_j(X_j, Y_{[0,j-2]}, u_j) dV(du)] \\ &= E[1_{\{X \in A_0\}} \dots 1_{\{Y_{j-2} \in A_{j-2}\}} \\ &\quad \times \int_{A_{j-2}} \left( \prod_{i=0}^{j-2} Y_i(X_{j-i}, Y_{[0,j-2]}, u_{j-i}) Y_j(X_j, (Y_{[0,j-2]}, u_{j-2}), u_j) V(du_{j-2}) V(du_j) \right)] \\ &= \dots \\ &= \int_{A_0} \dots \int_{A_j} \prod_{k=0}^j \underbrace{Y_k(X_k, u_{[0,k-2]}, u_{k-2})}_{!!} V(du_0) \dots V(du_j) \end{aligned}$$

Above of notation  $Y_k(X_k, u_{[0..k-2]})$

Then,

$$E[\epsilon(x_j) | \mathcal{F}_j^Y] = \frac{\dot{E}[\epsilon(x_j(\omega)) \prod_{i=0}^j Y_i(x_i(\omega), Y_{[0,i]})]}{\dot{E}[\prod_{i=0}^j Y_i(x_i(\omega), Y_{[0,i]})]}$$

Introduce the relation

$$L_j(x(\omega), y) = \prod_{i=0}^j Y_i(x_i(\omega), Y_{[0,i]})$$

and note that

$$\begin{aligned} & \dot{E}[\epsilon(x_j(\omega)) L_j(x(\omega), y) | \mathcal{F}_{j-1}^X] = \\ & = L_{j-1}(x(\omega), y) \dot{E}[\epsilon(x_j(\omega)) Y_j(x_j(\omega), Y_{[0,j]}) | \mathcal{F}_{j-1}^X] \\ & = L_{j-1}(x(\omega), y) \underbrace{\int_{\Omega} \epsilon(u) Y_j(u, Y_{[0,j]}) \Delta_j(x_{j-1}(\omega), Y_{[0,j]}, du)}_{\text{in}} \end{aligned}$$

Then,

$$\begin{aligned} E[\epsilon(x_j) | \mathcal{F}_j^Y] & = \frac{\dot{E}[\epsilon(x_j(\omega)) L_j(x(\omega), y)]}{\dot{E}[L_j(x(\omega), y)]} \\ & = \frac{\dot{E}[L_{j-1}(x(\omega), y) \int_{\Omega} \epsilon(u) Y_j(u, Y_{[0,j]}) \Delta_j(x_{j-1}(\omega), Y_{[0,j]}, du)]}{\dot{E}[L_{j-1}(x(\omega), y) \int_{\Omega} Y_j(u, X_{[0,j]}) \Delta_j(x_{j-1}(\omega), Y_{[0,j]}, du)]} \end{aligned}$$

$$= (*1) \frac{\dot{E}[L_{j-1}(x(\omega), y)]}{\dot{E}[L_{j-1}(x(\omega), y)]}$$

$$= \frac{\int_{\Omega} \epsilon(u) Y_j(u, Y_{[0,j]}) \Delta_j(x_{j-1}(\omega), Y_{[0,j]}, du) | \mathcal{F}_{j-1}^Y}{\int_{\Omega} Y_j(u, Y_{[0,j]}) \Delta_j(x_{j-1}(\omega), Y_{[0,j]}, du) | \mathcal{F}_{j-1}^Y}$$

Let  $\pi_j(dx)$  be the regular conditional distribution of  $X_j$  given  $\mathcal{F}_j^Y$ . Then, using Fatou

$$\int_{\mathbb{R}} \mathbb{E}(x) \pi_j(dx) = \mathbb{E}[\mathbb{E}(X_j) | \mathcal{F}_j^Y]$$

$$= \frac{\int_{\mathbb{R}} \int_{\mathbb{R}}^u \mathbb{E}(u) Y_j(u, Y_{[0,j]}) \Delta_j(x, Y_{[0,j-1]}, du) \pi_{j-1}(dx)}{\int_{\mathbb{R}} \int_{\mathbb{R}} Y_j(u, Y_{[0,j]}) \Delta_j(x, Y_{[0,j-1]}, du) \pi_{j-1}(dx)}$$

$$= \int_{\mathbb{R}}^u \mathbb{E}(u) \frac{\int_{\mathbb{R}}^x Y(u, Y_{[0,j]}) \Delta_j(x, Y_{[0,j-1]}, du) \pi_{j-1}(dx)}{\int_{\mathbb{R}} Y(u, Y_{[0,j]}) \Delta_j(x, Y_{[0,j-1]}, du) \pi_{j-1}(dx)}$$