

## KALMAN EXERCISE

*Exercise* (Linear prediction of an AR(1) observed with additive noise)

Consider an AR(1) real process  $Z_t$ <sup>1</sup> satisfying the following canonical equation:

$$\forall t \in \mathbb{N}, Z_{t+1} = \phi Z_t + \eta_t \quad (1)$$

where  $(\eta_t)_{t \geq 0}$  is a centered white noise with known variance  $\sigma^2$  and  $\phi$  is a known constant. The process  $(Z_t)_{t \geq 0}$  is not directly observed. Instead for all  $t \geq 1$ , one gets the following sequence of observations:

$$Y_t = Z_t + \epsilon_t \quad (2)$$

where  $(\epsilon_t)_{t \geq 1}$  is a centered white noise with known variance  $\rho^2$ , that is uncorrelated with  $(\eta_t)$  and  $Z_0$ . We wish to solve the filtering problem, that is, to compute the orthogonal projection of  $Z_t$  on the space  $H_t^Y = \text{span}\{Y_1, \dots, Y_t\}$ , iteratively in  $t$ .

We denote  $\hat{Z}_{t|t} = \text{proj}(Z_t | H_t^Y)$  this projection and  $P_{t|t} = \mathbb{E} \left[ (Z_t - \hat{Z}_{t|t})^2 \right]$  the corresponding projection error variance<sup>2</sup>. Similarly, let  $\hat{Z}_{t+1|t} = \text{proj}(Z_{t+1} | H_t^Y)$  be the best linear predictor and  $P_{t+1|t} = \mathbb{E} \left[ (Z_{t+1} - \hat{Z}_{t+1|t})^2 \right]$  the linear prediction error variance.

1. Show that  $Z_0$  is a centered random variable and computes its variance  $\sigma_0^2$  using the Corollary 3.1.3 and that  $Z_0$  and  $(\eta_t)_{t \geq 0}$  are uncorrelated.<sup>3</sup>
2. Using the evolution (state) equation (1), show that

$$\hat{Z}_{t+1|t} = \phi \hat{Z}_{t|t} \quad \text{and} \quad P_{t+1|t} = \phi^2 P_{t|t} + \sigma^2$$

3. Let us define the innovation by  $I_{t+1} = Y_{t+1} - \text{proj}(Y_{t+1} | H_t^Y)$ . Using the observation equation (2), show that  $I_{t+1} = Y_{t+1} - \hat{Z}_{t+1|t}$
4. Prove that  $\mathbb{E}[I_{t+1}^2] = P_{t+1|t} + \rho^2$
5. Give the arguments that shows

$$\hat{Z}_{t+1|t+1} = \hat{Z}_{t+1|t} + k_{t+1} I_{t+1}$$

$$\text{where } k_{t+1} = \mathbb{E}[Z_{t+1} I_{t+1}] / \mathbb{E}[I_{t+1}^2] \quad ^4$$

6. Using the above expression of  $I_{t+1}$ , show that  $\mathbb{E}[Z_{t+1} I_{t+1}] = P_{t+1|t}$

<sup>1</sup>the same exercise can be apply to a complex AR(1) process  $Z_t$ . Try by yourself to see what could be the slight difference in that case.

<sup>2</sup>in complex case:  $P_{t|t} = \mathbb{E} \left[ |Z_t - \hat{Z}_{t|t}|^2 \right]$

<sup>3</sup>Hint: decompose  $Z_t$  as  $F_\phi(B) \eta_t$  where  $F_\phi(B)$  is a rational polynom fraction depends on the backshift operator and then decompose  $F_\phi(B) \eta_t$  as an infinite sum.

<sup>4</sup> $k_{t+1}$  is the Kalman gain filter

7. Why is the following equation correct ?

$$P_{t+1|t+1} = P_{t+1|t} - \mathbb{E} \left[ (k_{t+1} I_{t+1})^2 \right]$$

Deduce that  $P_{t+1|t+1} = (1 - k_{t+1}) P_{t+1|t}$ .

8. Provide the complete set of equations for computing  $\hat{Z}_{t|t}$  and  $P_{t|t}$  iteratively for all  $t \geq 1$  (Including the initial conditions.)
9. *Bonus*: Study the asymptotic behavior of  $P_{t|t}$  as  $t \rightarrow \infty$ .