

* Cond. Expectation as a information processing tool

X, Y

$$\min_{h(Y)} \mathbb{E} \|X - h(Y)\|^2 \Rightarrow h(Y) = \mathbb{E}[X|Y]$$

$$\mathbb{E} \left\{ \mathbb{E} \left[\|X - h(Y)\|^2 | Y \right] \right\}$$

$$X^2 + h(Y)^2 - X h(Y)$$

$$\mathbb{E} \left\{ \mathbb{E}[X^2|Y] - \mathbb{E}(X h(Y) | Y) \cdot h(Y) \right\}$$

$$\Rightarrow \min_{\text{for}} \boxed{h(Y) = \mathbb{E}(X|Y)}$$

* Exercise: $\underline{X}, \underline{Y}$ Gaussian Random Vectors;

Show that

$$\mathbb{E}[\underline{X}|\underline{Y}] = \underline{L}\underline{Y}$$

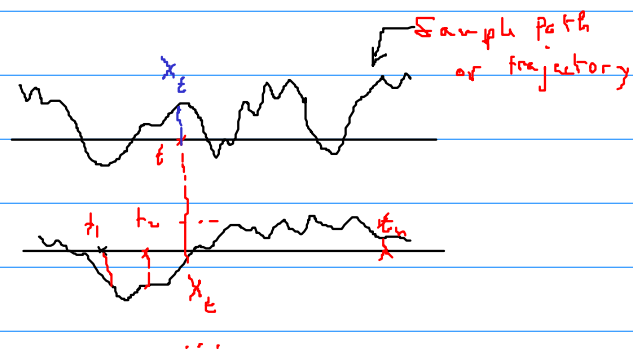
$$\left\{ \underline{L} = \underbrace{\mathbb{E}[\underline{X}\underline{Y}^T]}_{R_{XY}} \cdot \underbrace{\mathbb{E}[\underline{Y}\underline{Y}^T]}_{R_Y}^{-1} \right\}$$

(X, Y) gaussian:

$$\underline{Z} = \begin{bmatrix} X \\ Y \end{bmatrix}, \quad R_Z = \begin{bmatrix} R_X & R_{XY} \\ R_{YX} & R_Y \end{bmatrix}; \quad R_{YX} = R_{XY}^T$$

$$\mathbb{E}[X|Y] = \int X \frac{P_{X,Y}(x,y)}{P_Y(y)} dx$$

10/ Random Process: $\{X_t, t \in \mathbb{R}\}$



Independent increment process:

t_0, t_1, \dots, t_n

$$(X_{t_2} - X_{t_1}), (X_{t_3} - X_{t_2}), \dots, (X_{t_n} - X_{t_{n-1}}) \quad \text{Independent}$$

Ex: Brownian process