

let

$$x_{t-1} \sim \mathcal{N}(\mu_x, \bar{\Sigma}_x), \quad w_{t-1} \sim \mathcal{N}(\mu_w, \bar{\Sigma}_w)$$

as in:

- ① compute the parameters of the joint distribution  $P(x_t, w_{t-1})$ , assuming  $x$  and  $w$  are independent

$$\mu_{x,w} = \begin{pmatrix} ? \\ ? \end{pmatrix} \quad \bar{\Sigma}_{x,w} = \begin{pmatrix} ? & 0 \\ 0 & ? \end{pmatrix}$$

- ② consider the following affine transformation

$$x_t = \begin{pmatrix} A_t & B_t \end{pmatrix} \begin{pmatrix} x_{t-1} \\ w_{t-1} \end{pmatrix} + c_t$$

compute the parameters of  $P(x_t)$

$$\mu_{x_{t|t-1}} =$$

$$\bar{\Sigma}_{x_{t|t-1}} =$$

- ③ consider the following conditional density

$$z_t = \mathcal{N}(C_t \cdot x_{t|t-1} + d_t; \bar{\Sigma}_t) \quad , \quad \text{and consider } x_{t|t-1} = \mathcal{N}\left(\mu_{x_{t|t-1}}, \bar{\Sigma}_{t|t-1}\right)$$

apply the chain rule to compute the joint  $P(x_{t|t-1}, z_t)$

$$\mu_{xz} = \begin{pmatrix} \mu_x \\ ? \end{pmatrix}$$

$$\bar{\Sigma}_{xz} = \begin{pmatrix} \bar{\Sigma}_{xx} & \bar{\Sigma}_{xz} \\ \bar{\Sigma}_{zx} & \bar{\Sigma}_{zz} \end{pmatrix}$$

$$\begin{aligned} \bar{\Sigma}_{xx} &=? \\ \bar{\Sigma}_{xz} &=? \\ \bar{\Sigma}_{zx} &=? \\ \bar{\Sigma}_{zz} &=? \end{aligned}$$

- ④ Apply the conditioning on  $z_t = \bar{z}_t$  to the joint of the step before, and get the parameters of  $P(x_t | z_t)$

$$\mu_{x_t} = ?$$

$$\bar{\Sigma}_{x_t} = ?$$