

Lecture 20: State-space models - Kalman filters

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Derivation of Kalman filter - Predict

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Predict

$$0: p(x_0) = \mathcal{N}(t_0, V_0)$$

$$1: p(x_1 | u_0) = \int \underbrace{p(x_1 | x_0, u_0)}_{\mathcal{N}(Ax_0 + Bu_0, Q)} \underbrace{p(x_0)}_{\mathcal{N}(t_0, V_0)} dx_0$$

$$= \mathcal{N}(x_1 | At_0 + Bu_0, AV_0A^T + Q)$$

$$x_1 = Ax_0 + Bu_0 + z_1, z_1 \sim \mathcal{N}(0, Q)$$

$$E[x_1] = E[Ax_0 + Bu_0] + E[z_1]$$

$$= A \cdot E[x_0] + Bu_0 = At_0 + Bu_0$$

$$C[x_1] = C[Ax_0 + Bu_0 + z_1]$$

$$= C[Ax_0] + C[z_1] + C[Ax_0, z_1]$$

$$= C[Ax_0 + z_1] = \underbrace{C[Ax_0]}_{AV_0A^T} + C[z_1] + C[Ax_0, z_1]$$

$$\dots$$

$$n: p(x_{n-1} | y_{1:n-1}, u_{0:n-2}) = \mathcal{N}(x_{n-1} | t_{n-1}, V_{n-1})$$

Assume you know this.

$$p(x_n | y_{1:n-1}, u_{0:n-2}, u_{n-1}) = \int \underbrace{p(x_n | x_{n-1}, u_{n-1})}_{\mathcal{N}(Ax_{n-1} + Bu_{n-1}, Q)} \underbrace{p(x_{n-1} | y_{1:n-1}, u_{0:n-2})}_{\mathcal{N}(x_{n-1} | t_{n-1}, V_{n-1})} dx_{n-1}$$

$$= \mathcal{N}(x_n | At_{n-1} + Bu_{n-1}, AV_{n-1}A^T + Q)$$

$$x_n = Ax_{n-1} + Bu_{n-1} + z_n, z_n \sim \mathcal{N}(0, Q)$$

$$E[x_n] = At_{n-1} + Bu_{n-1}; C[x_n] = AV_{n-1}A^T + Q$$

If $p(x_{n-1} | y_{1:n-1}, u_{0:n-2}) = \mathcal{N}(x_{n-1} | t_{n-1}, V_{n-1})$

Then Predict:

$$p(x_n | y_{1:n-1}, u_{0:n-2}, u_{n-1}) = \mathcal{N}(x_n | At_{n-1} + Bu_{n-1}, AV_{n-1}A^T + Q)$$