

# 1 Probability ①

Random variable.  $x$ .  $E(x)$ ,  $Var[x]$

$$R_{vv}(z) = \sigma_v^2 \delta(z)$$

Auto-correlation  $E[x(t)x(t+z)]$ , white noise, power spectral density is constant

# 2 Kalman Filter ②, ③, ④

$$\dot{x} = Ax + Bu + \bar{F}v \Rightarrow \text{disturbance}$$

dynamic:  $y = Cx + w \rightarrow \text{noise}$



observer/filter  $\dot{\hat{x}} = A\hat{x} + Bu + k_f(y - C\hat{x})$

$$\hat{y} = C\hat{x} + w$$

*a gain / dependence of new measurement*

$$\min_{k_f} E[(x - \hat{x})(x - \hat{x})^T]$$

$$\dot{P} = AP + PA^T - PC^T W^{-1} C P + F V F^T$$

steady-state  $\dot{P} = 0$   $k_f = P C^T W^{-1}$

$w$ : co-variance of  $w$

$v$ : co-variance of  $v$ .

$x$ : 0 unrelated  $w$ ,  $v$

# 3 Properties ④

1) Frequency response  $y \rightarrow \hat{y}$   $G_{kf}(s) = C \cdot (sI - A + k_f c)^{-1} k_f$

2) duality between LQR and Kalman Filtering  $A^T P + PA - PBR^{-1}B^T P + Q = 0$

# 4 LQG design ⑤

$$\begin{cases} \dot{x} = Ax + Bu + Fv \\ y = Cx + w \end{cases}$$

v.s. LQR

$$\dot{x} = Ax + Bu$$

$$J = \int_0^{\infty} (x^T Q x + u^T R u) dt$$

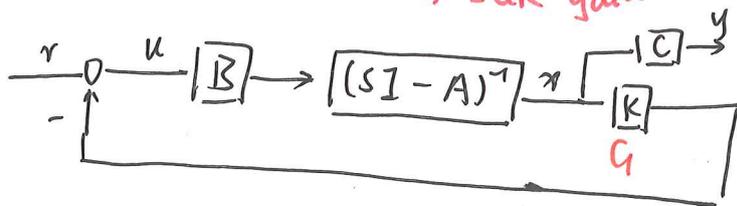
$$u = -Gx$$

$\Rightarrow$  LQR gain

$$J = E \left[ \int_0^{\infty} (x^T Q x + u^T R u) dt \right]$$

$$u = -G \hat{x}$$

$\hat{x}$  is the estimation of  $x$



$$L(x) = K(sI - A)^{-1} B$$

$K(G)$

$$L_{lqg}(s) = G(sI - A + BG + k_f c)^{-1} k_f \cdot C(sI - A)^{-1} B$$

① Separation Theorem.

② Cannot guarantee stability margin

# 5 Loop Transfer Recovery ⑥

$$F = B, \quad v = v_0 + q^2 I \quad w = w_0$$

$$\lim_{q \rightarrow \infty} L_{lqg}(s) = L_{lqr}(s)$$