

goal: compare & combine optimal controller and estimator for linear systems with Gaussian noise

ref: Stengel ch 5.3

duality - i.e. comparison of LQ controller & KF estimator

$$\begin{aligned} \bullet \text{ consider } x_{s+1} &= A_s x_s + B_s u_s + \delta_s, \quad E[\delta_s] = 0, \quad \text{Cov}[\delta_s] = \bar{Q}_s \\ y_s &= C_s x_s + \eta_s, \quad E[\eta_s] = 0, \quad \text{Cov}[\eta_s] = \bar{R}_s \end{aligned}$$

$$\text{control cost} \quad \sum_{s=0}^t x_s^T Q_s x_s + u_s^T R_s u_s$$

$$\text{optimal controller} \quad u_s^* = -K_s x_s$$

where  $K_s$  solves Riccati DE:

$$K_s = (B_s^T P_{s+1} B_s + R_s)^{-1} B_s^T P_{s+1} A_s$$

$$P_s = (A_s - B_s K_s)^T P_{s+1} (A_s - B_s K_s)$$

$$\text{estimation error} \quad \sum_{s=0}^t \delta_s^T \bar{Q}_s^{-1} \delta_s + \eta_s^T \bar{R}_s^{-1} \eta_s$$

$$\text{optimal estimator} \quad \hat{x}_s = \tilde{x}_s + L_s (y_s - C_s \tilde{x}_s)$$

where  $L_s$  comes from Kalman filter:

$$\tilde{x}_s = A_{s-1} \hat{x}_{s-1} + B_{s-1} u_{s-1}$$

$$L_s = \tilde{P}_s C_s^T (C_s \tilde{P}_s C_s^T + \bar{R}_s)^{-1}$$

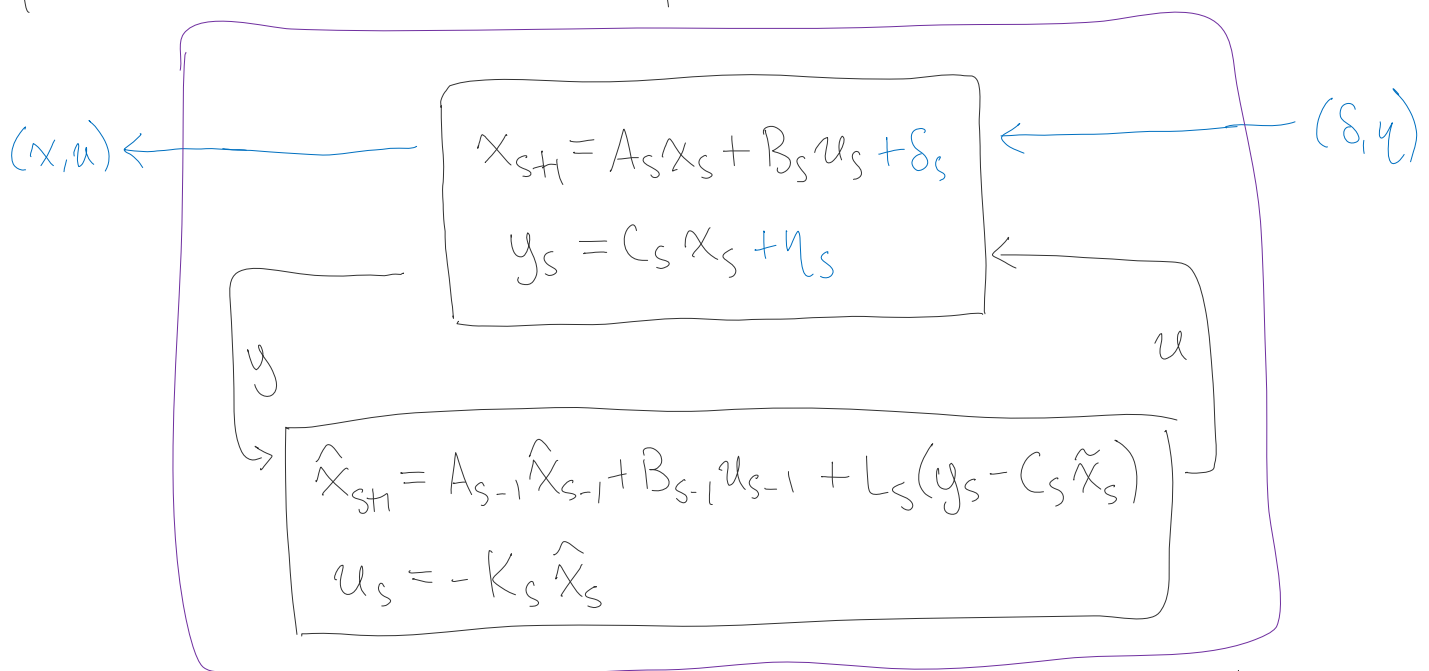
$$P_s = (A_s - B_s K_s)^T P_{s+1} (A_s - B_s K_s) + K_s^T R_s K_s + Q_s$$

$$\begin{aligned} L_s &= P_s C_s^T (C_s P_s C_s^T + R_s)^{-1} \\ \tilde{P}_s &= A_{s-1} \hat{P}_{s-1} A_{s-1}^T + \bar{Q}_s \\ \hat{P}_s &= (I - L_s C_s) \tilde{P}_s (I - L_s C_s)^T + L_s \bar{R}_s L_s^T \end{aligned}$$

\* solutions are eerily similar! mathematically: dual  
 - in continuous-time, on infinite horizon (time-invariant):

$$K = LQR(A, B, Q, R) \quad ; \quad L^T = LQR(A^T, C^T, \bar{Q}, \bar{R})$$

separation - i.e. combination of LQ controller & KF estimator



$\{K_s\}$  minimizes  $\sum_{s=0}^t x_s^T Q_s x_s + u_s^T R_s u_s$   
 $\{L_s\}$  minimizes  $\sum_{s=0}^t \delta_s^T \bar{Q}_s^{-1} \delta_s + \eta_s^T \bar{R}_s^{-1} \eta_s$

even though controller & estimator are designed separately,  
 $e = x - \hat{x} \rightarrow 0, \hat{x} \rightarrow 0$  as  $t \rightarrow \infty$

\* furthermore, this linear-quadratic-Gaussian (LQG) regulator

minimizes  $E \left[ \sum_{s=0}^t x_s^T Q_s x_s + u_s^T R_s u_s \right]$  where expectation is taken

minimizes  $E \left[ \sum_{s=0}^t x_s^T Q_s x_s + u_s^T R_s u_s \right]$  where expectation is taken  
with respect to  $\{\delta_s\}, \{\eta_s\}$

— termed "separation principle" in stochastic optimal control  
(or "certainty equivalence")