

AA ECE ME 548: Linear Multivariable Control  
Prof Burden TA Tinu Spring 2020

- today:
- ☒ exam 1 results, solution, regrade procedure
  - ☒ HW 6 (optimal estimation) overview & Q's
  - ☒ lec 7 (Kalman Filter) Q's
  - ~~☐ ~10 min break + discussion & follow-up~~
  - ☐ Prof Burden OH

---

what is covariance?

• let  $x: \Omega \rightarrow \mathbb{R}^2$  be a random vector

$$E[x] \in \mathbb{R}^2 \text{ -- the mean (or average)}$$
$$= \begin{bmatrix} E[x_1] \\ E[x_2] \end{bmatrix}$$

$$\text{Cov}[x]^T = E[(x - E[x])(x - E[x])^T]^T$$

$$\begin{aligned}
 \text{Cov}[x] &= E[(x - E[x])(x - E[x])^T] \\
 &= E\left[\left((x - E[x])(x - E[x])^T\right)^T\right] \\
 &= E[(x - E[x])(x - E[x])^T] = \text{Cov}[x]
 \end{aligned}$$

- let  $z = x - E[x]$  so  $\text{Cov}[x] = \text{Cov}[z]$

$$\begin{aligned}
 \text{Cov}[z] &\in \mathbb{R}^{2 \times 2} \quad \text{Cov}[z] = \begin{bmatrix} E[z_1 \cdot z_1] & E[z_1 \cdot z_2] \\ \times & E[z_2 \cdot z_2] \end{bmatrix} \\
 &= \begin{bmatrix} \text{Var}[z_1] & \text{corr}[z_1, z_2] \\ \times & \text{Var}[z_2] \end{bmatrix}
 \end{aligned}$$

• if  $\underbrace{z_1, z_2}_{\text{if independent}} \sim \mathcal{N}(0, \sigma^2)$ ,  $\text{Var}[z_1] = \text{Var}[z_2] = \sigma^2$   
 $E[z_1 \cdot z_2] = 0$

$$\Rightarrow \text{Cov}[z] = \begin{bmatrix} \sigma^2 & 0 \\ 0 & \sigma^2 \end{bmatrix} = \sigma^2 \cdot I$$

geometric intuition for covariance

• let  $x: \Omega \rightarrow \mathbb{R}^n$  be random vector  $E[x] = \mu$   $\text{Cov}[x] = \Sigma$

• let  $x: \Omega \rightarrow \mathbb{R}^n$  be random vector,  $E[x] = \mu$ ,  $\text{Cov}[x] = \Sigma$

fact:  $\Sigma^T = \Sigma \geq 0$  so  $\exists U$  s.t.  $U^T U = I$ , i.e.  $U^{-1} = U^T$   
and  $U \Sigma U^T = D$ , diagonal  
(i.e.  $D = \text{diag}\{\lambda_1, \dots, \lambda_n\}$ )

• define  $z = U x$  so  $E[z] = U \mu$ ,  $\text{Cov}[z] = U \Sigma U^T = D$   
i.e.  $\text{Var}[z_i] = \lambda_i \geq 0$  and  $\underbrace{\text{corr}[z_i, z_j]}_{E[z_i z_j]} = 0, i \neq j$

---

least squares

given  $z = Cx + \eta$ , estimate  $\hat{x}$  to minimize  $\|z - C\hat{x}\|$

\* if  $E[\eta] = 0$  then  $E[\hat{x}] = E[x]$

so  $E[x - \hat{x}] = 0$

and  $\text{Cov}[x - \hat{x}]$  is as small as possible using  
 $\text{Cov}[\eta]$  as a weighting matrix