# ME424 HW6

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December 15, 2020

## 1

Given

$$X = \begin{bmatrix} X_1 & X_2 & X_3 \end{bmatrix}^T$$

$$\Sigma_X = \begin{bmatrix} 2 & 0 & \sigma_{13} \\ 0 & 2 & \sigma_{23} \\ \sigma_{31} & \sigma_{32} & 2 \end{bmatrix}$$

## 1.1

$$W \sim N(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix})$$

Given Z = 10.

$$\begin{split} \mu_{W|Z=10} &= \mu_W + \Sigma_{WZ} \Sigma_Z^{-1} (10 - \mu_Z) \\ &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix} 2^{-1} * 10 \\ &= \begin{bmatrix} 5 \\ 5 \end{bmatrix} \\ \Sigma_{W|Z=10} &= \Sigma_X - \Sigma_{XZ} \Sigma_Z^{-1} \Sigma_{ZX} \\ &= \begin{bmatrix} 3/2 & -1/2 \\ -1/2 & 3/2 \end{bmatrix} \end{split}$$

If I observed Z = -10 instead, the conditional mean of W will be negative according to the formula above. On the other hand,  $\Sigma_{WZ}$  shows that W is positive correlated with Z, and  $\Sigma_{Z}^{-1}$  shows that the measurement of Z has little noisy. Thus, the conditional mean of W is likely to be negative if Z was observed negative.

#### 1.2

 $X_1$  is independent of  $X_2$  because  $X_1$  and  $X_2$  are jointly Gaussian and  $E(X_1X_2^T) = E(X_1)E(X_2)^T = 0$ ,  $Cov(X_1, X_2) = 0$ . Define

$$Q = (W|Z = 10)$$

We have

$$Q \sim N(\begin{bmatrix} 5 \\ 5 \end{bmatrix}, \begin{bmatrix} 3/2 & -1/2 \\ -1/2 & 3/2 \end{bmatrix})$$

Since  $Cov(Q_1, Q_2) = -1/2$ , they are not independent. If  $\sigma_{13} = \sigma_{31} = 0$  then  $Cov(Q_1, Q_2) = 0$ , so they are independent.

## 1.3

The probability density function for W|Z=10 is

$$g(x) = \frac{1}{2\pi |\Sigma|^{-\frac{1}{2}}} \exp^{-\frac{1}{2}(X-\mu)^T \Sigma (X-\mu)}$$

The MATLAB code is shown below.

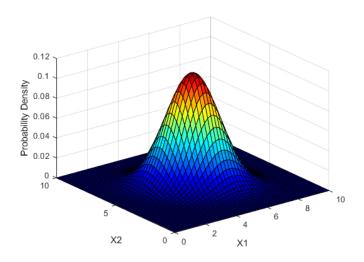


Figure 1: PDF for W—Z = 10

 $\mathbf{2}$ 

2.1

i

$$\begin{array}{ccc} x_1 & 1 & 2 \\ p(x_1|X_2=-1,X_3=3) & \frac{2}{7} & \frac{5}{7} \end{array}$$

ii

$$\begin{array}{cccc} x_1, x_3 & 1, 3 & 2, 2 & 2, 3 \\ p(x_1, x_3 | X_2 = -1) & \frac{1}{4} & \frac{1}{8} & \frac{5}{8} \\ P(X_3 = 3 | X_2 = -1) = \frac{7}{8} \\ \\ p(x_1 | X_2 = -1, X_3 = 3) = \frac{p(x_1, x_3 | X_2 = -1)}{P(X_3 = 3 | X_2 = -1)} \\ p(x_1 | X_2 = -1, X_3 = 3) & \frac{1}{2} & \frac{2}{7} & \frac{5}{7} \end{array}$$

2.2

$$\hat{X}_{1MMSE} = E(x_1|X_2 = -1, X_3 = 3)$$
$$= \frac{12}{7}$$

3

When  $\mu = [1, 2]^T$ ,

$$\mu_{X|Y=1}^{(1)} = \mu_x + \Sigma_{XY} \Sigma_Y^{-1} (1 - \mu_Y)$$
$$= 1 + (-1) \times \frac{1}{2} \times (1 - 2)$$
$$= \frac{3}{2}$$

When  $\mu = [2, 1]^T$ ,

$$\mu_{X|Y=1}^{(2)} = \mu_x + \Sigma_{XY} \Sigma_Y^{-1} (1 - \mu_Y)$$
  
= 2

Thus,

$$X_{MMSE} = \mu_{X|Y=1}^{(1)} Prob(\mu = \mu^{(1)}) + \mu_{X|Y=1}^{(2)} Prob(\mu = \mu^{(2)})$$
= 1.9

4

 $\operatorname{Given}$ 

$$\begin{split} X &\sim N(\begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix}) \\ V &\sim N(0,3) \\ Z &= HX + V, \ where \ H = \begin{bmatrix} 1 \ 2 \end{bmatrix} \end{split}$$

## 4.1

We have

$$\begin{split} E(Z) &= HE(X) + E(V) \\ &= 5 \\ Cov(Z) &= Cov(HX + V, HX + V) \\ &= HCov(X)H^T + Cov(V) \\ &= 8 \\ Coc(X, Z) &= Cov(X, HX + V) \\ &= Cov(X)H^T \\ &= \begin{bmatrix} -1 \\ 3 \end{bmatrix} \\ Cov(Z, X) &= Cov(HX + V, X) \\ &= HCov(X) \\ &= [-1\ 3] \end{split}$$

Thus

$$E(X|Z=4) = \mu_X + \Sigma_{XZ} \Sigma_Z^{-1} (4 - \mu_Z)$$

$$= \begin{bmatrix} 9/8 \\ 13/8 \end{bmatrix}$$

$$Cov(X|Z=4) = \Sigma_X - \Sigma_{XZ} \Sigma_Z^{-1} \Sigma_{ZX}$$

$$= \begin{bmatrix} 7/8 & -5/8 \\ -5/8 & 7/8 \end{bmatrix}$$

## 4.2

I need update the mean and covariance of X given Z = 4.

$$\begin{cases} E(X) \\ Cov(X) \end{cases} \xrightarrow{measurement \ update} \begin{cases} E(X|Z=4) \\ Cov(X|Z=4) \end{cases}$$

According to Kalman formula,

$$K = \Sigma_X H^T (H \Sigma_X H^T + \Sigma_V)^{-1}$$

$$= \begin{bmatrix} -1/8 \\ 3/8 \end{bmatrix}$$

$$E(X|Z=4) = \mu_X + K(4 - H\mu_X)$$

$$= \begin{bmatrix} 9/8 \\ 13/8 \end{bmatrix}$$

$$Cov(X|Z=4) = (I - KH)\Sigma_X$$

$$= \begin{bmatrix} 7/8 & -5/8 \\ -5/8 & 7/8 \end{bmatrix}$$