

# MTH 383/683: Homework #6

Due Date: October 27, 2023

## 1 Problems for Everyone

1. **Conditional Expectation of Continuous Random Variables** Let  $X, Y$  be two random variables with joint density  $f(x, y)$  on  $\mathbb{R}^2$  and assume  $f(x, y) > 0$  for all  $x, y \in \mathbb{R}$ . Show that the conditional expectation  $\mathbb{E}[Y|X]$  equals  $h(X)$  where  $h$  is the function

$$h(x) = \frac{\int_{-\infty}^{\infty} yf(x, y)dy}{\int_{-\infty}^{\infty} f(x, y)dy}.$$

**Hint:** To prove this you need to show both properties of conditional expectation.

2. **Exercises on  $\sigma$ -fields** The Borel sets of  $\mathbb{R}$ , denoted  $\mathcal{B}(\mathbb{R})$ , is the smallest  $\sigma$ -algebra on  $\mathbb{R}$  containing intervals of the form  $(a, b]$ . That is,  $\mathcal{B}(\mathbb{R})$  contains all possible unions and intersections of intervals of the form  $(a, b]$ .
  - Show that all singletons  $\{b\}$  are in  $\mathcal{B}(\mathbb{R})$  by writing  $\{b\}$  as the infinite intersection of intervals of the form  $(b - 1/n, b + 1/n]$ .
  - Prove that all open intervals  $(a, b)$  and closed intervals  $[a, b]$  are in  $\mathcal{B}(\mathbb{R})$ .
3. **Another Look at Conditional Expectation for Gaussians** Let  $(X, Y)$  be a Gaussian vector with mean 0 and covariance matrix

$$C = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix},$$

where  $\rho \in (-1, 1)$ .

- Use Equation (4.5) in the text to show that  $\mathbb{E}[Y|X] = \rho X$ .
- Write down the joint PDF  $f(x, y)$  of  $(X, Y)$ .
- Show that

$$\int_{-\infty}^{\infty} yf(x, y)dy = \rho x \text{ and } \int_{-\infty}^{\infty} f(x, y)dy = 1.$$

- Use problem #1 on this homework to show that  $\mathbb{E}[Y|X] = \rho X$ .

4. **Gaussian Conditioning** Consider the Gaussian vector  $(X_1, X_2, X_3)$  with mean 0 and covariance matrix

$$C = \begin{bmatrix} 2 & 2 & 0 \\ 2 & 4 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

- (a) Prove that  $X_3$  is independent of  $X_2$  and  $X_1$ .
  - (b) Compute  $\mathbb{E}[X_2|X_1]$ .
  - (c) Write  $X_2$  as a linear combination of  $X_1$  and a random variable independent of  $X_1$ .
  - (d) Compute  $\mathbb{E}[e^{aX_2}|X_1]$  for any  $a \in \mathbb{R}$ .
5. Let  $B_t$  be a standard Brownian motion. Verify that  $M_t = B_t^2 - t$  is a martingale for the Brownian filtration.

## Homework #6

#1

Let  $X, Y$  be two random variables with joint density  $f(x, y)$  on  $\mathbb{R}^2$  and assume  $f(x, y) > 0$  for all  $x, y \in \mathbb{R}$ . Show that

$$\mathbb{E}[Y|X] = h(X)$$

where

$$h(x) = \frac{\int_{-\infty}^{\infty} y f(x, y) dy}{\int_{-\infty}^{\infty} f(x, y) dy}$$

Solution:

Let  $g(X)$  be any function. Therefore,

$$\begin{aligned}
 \mathbb{E}[g(X)\mathbb{E}[Y|X]] &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x) \left( \frac{\int_{-\infty}^{\infty} z f(x, z) dz}{\int_{-\infty}^{\infty} f(x, w) dw} \right) f(x, y) dy dx \\
 &= \int_{-\infty}^{\infty} g(x) \int_{-\infty}^{\infty} z f(x, z) dz \left( \int_{-\infty}^{\infty} f(x, y) dy \right) dx \\
 &= \int_{-\infty}^{\infty} g(x) \int_{-\infty}^{\infty} z f(x, z) dz \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x) z f(x, z) dz \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x) y f(x, y) dy \\
 &= \mathbb{E}[g(X)Y].
 \end{aligned}$$

Thus,

$$\mathbb{E}[Y|X] = \mathbb{E}[g(X)Y].$$

#2.

The Borel sets of  $\mathbb{R}$ , denoted  $\mathcal{B}(\mathbb{R})$  is the smallest  $\sigma$ -algebra on  $\mathbb{R}$  containing intervals of the form  $(a,b]$ .

(a) Show that singletons  $\{b\}$  are in  $\mathcal{B}(\mathbb{R})$ .

(b) Prove that all open intervals  $(a,b)$  and closed intervals  $[a,b]$  are in  $\mathcal{B}(\mathbb{R})$ .

Solution:

(a) Since  $\{b\} = \bigcap_{n=1}^{\infty} (b - \frac{1}{n}, b + \frac{1}{n}]$  it follows that  $\{b\} \in \mathcal{B}(\mathbb{R})$ .

(b) Since  $[a,b] = (a,b] \cup \{a\}$  and  $(a,b) = (a,b] \cap \{b\}^c$  it follows that  $(a,b), [a,b] \in \mathcal{B}(\mathbb{R})$ . ■

#3.

Let  $(X, Y)$  be a Gaussian vector with mean  $0$  and covariance matrix

$$C = \begin{bmatrix} 1 & s \\ s & 1 \end{bmatrix}$$

where  $s \in (-1, 1)$ .

(a) Show that  $\mathbb{E}[Y|X] = sX$ .

(b) Write down the joint P.D.F. of  $(X, Y)$ .

(d) Use problem #1 to show that  $\mathbb{E}[Y|X] = sX$ .

Solution:

(a)  $\mathbb{E}[Y|X] = \frac{\mathbb{E}[YX]}{\mathbb{E}[X^2]} X = \frac{s}{1} X = sX$ .

(b).  $f(x, y) = \frac{1}{2\pi\sqrt{1-s^2}} e^{-(x^2 - 2sxy + y^2)/2(1-s^2)}$

(c) Computing, we have that

$$\begin{aligned} \int_{-\infty}^{\infty} y f(x, y) dy &= \frac{1}{2\pi\sqrt{1-s^2}} \int_{-\infty}^{\infty} y e^{-(x^2-2sy+y^2)/(1-s^2)} dy \\ &= \frac{1}{2\pi\sqrt{1-s^2}} \int_{-\infty}^{\infty} y e^{-x^2/(1-s^2)} e^{-(y^2-2sy+s^2x^2)/(1-s^2)} e^{s^2x^2/(1-s^2)} dy \\ &= \frac{e^{-(1-s^2)x^2}}{2\pi\sqrt{1-s^2}} \int_{-\infty}^{\infty} e^{-(y-sx)^2/(1-s^2)} dy \\ &= \frac{e^{-(1-s^2)x^2}}{\sqrt{2\pi}} sx. \end{aligned}$$

$$\begin{aligned} \int_{-\infty}^{\infty} f(x, y) dy &= \frac{1}{2\pi\sqrt{1-s^2}} \int_{-\infty}^{\infty} e^{-(1-s^2)x^2} e^{-(y-sx)^2/(1-s^2)} dy \\ &= \frac{1}{\sqrt{2\pi}} e^{-(1-s^2)x^2} \end{aligned}$$

(d) From problem #1 we have that

$$\mathbb{E}[Y|X] = h(X),$$

where

$$\begin{aligned} h(x) &= \int_{-\infty}^{\infty} y f(x, y) dy \\ &= \int_{-\infty}^{\infty} y e^{-(1-s^2)x^2} e^{-(y-sx)^2/(1-s^2)} dy \\ &= sx. \end{aligned}$$

Therefore,

$$\mathbb{E}[Y|X] = sX.$$

#4.

Consider the Gaussian vector  $(\bar{X}_1, \bar{X}_2, \bar{X}_3)$  with mean 0 and covariance matrix

$$C = \begin{bmatrix} 2 & 2 & 0 \\ -2 & 4 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

(a) Prove that  $\bar{X}_3$  is independent of  $\bar{X}_1$  and  $\bar{X}_2$ .

(b) Compute  $\mathbb{E}[\bar{X}_2 | \bar{X}_1]$

(c) Write  $\bar{X}_2$  as a linear combination of  $\bar{X}_1$  and a random variable independent of  $\bar{X}_1$ .

(d) Compute  $\mathbb{E}[e^{a\bar{X}_2} | \bar{X}_1]$  for any  $a \in \mathbb{R}$ .

Solution:

(a) Since  $\text{Cov}(\bar{X}_1, \bar{X}_2) = \text{Cov}(\bar{X}_3, \bar{X}_1) = 0$  it follows that  $\bar{X}_3$  is independent of  $\bar{X}_1$  and  $\bar{X}_2$ .

(b).  $\mathbb{E}[\bar{X}_2 | \bar{X}_1] = \frac{\mathbb{E}[\bar{X}_2 \bar{X}_1]}{\mathbb{E}[\bar{X}_1^2]} \bar{X}_1 = \frac{2}{2} \bar{X}_1 = \bar{X}_1$

(c) Using the orthogonal decomposition we have that

$$\begin{aligned} \bar{X}_2 &= (\bar{X}_2 - \mathbb{E}[\bar{X}_2 | \bar{X}_1]) + \mathbb{E}[\bar{X}_2 | \bar{X}_1] \\ &= (\bar{X}_2 - \bar{X}_1) + \bar{X}_1 \end{aligned}$$

$$\begin{aligned} (d). \mathbb{E}[e^{a\bar{X}_2} | \bar{X}_1] &= \mathbb{E}[e^{a(\bar{X}_2 - \bar{X}_1)} e^{a\bar{X}_1} | \bar{X}_1] \\ &= e^{a\bar{X}_1} \mathbb{E}[e^{a(\bar{X}_2 - \bar{X}_1)} | \bar{X}_1] \\ &= e^{a\bar{X}_1} \mathbb{E}[e^{a(\bar{X}_2 - \bar{X}_1)}] \end{aligned}$$

Now,  $\bar{X}_2 - \bar{X}_1$  is Gaussian with mean 0 and variance.

$$\begin{aligned} \sigma^2 &= \mathbb{E}[(\bar{X}_2 - \bar{X}_1)^2] \\ &= \mathbb{E}[\bar{X}_2^2] - 2\mathbb{E}[\bar{X}_2 \bar{X}_1] + \mathbb{E}[\bar{X}_1^2] \\ &= 4 - 4 + 2 \\ &= 2. \end{aligned}$$

Therefore,  $\mathbb{E}[e^{a\bar{X}_2} | \bar{X}_1] = e^{a\bar{X}_1} e^{a^2 \sigma^2 / 2} = e^{a\bar{X}_1 + a^2}$ . ■

#5

Let  $B_t$  be a standard Brownian motion. Verify that  $M_t = B_t^2 - t$  is a martingale for the Brownian filtration.

Solution:

$$\begin{aligned} \mathbb{E}[M_t | \mathcal{F}(B_s)] &= \mathbb{E}[B_t^2 - t | \mathcal{F}(B_s)] \\ &= \mathbb{E}[(B_t - B_s + B_s)^2 | \mathcal{F}(B_s)] - t \\ &= \mathbb{E}[(B_t - B_s)^2 + 2(B_t - B_s)B_s + B_s^2 | \mathcal{F}(B_s)] - t \\ &= \mathbb{E}[(B_t - B_s)^2 | \mathcal{F}(B_s)] + 2\mathbb{E}[(B_t - B_s)B_s | \mathcal{F}(B_s)] \\ &\quad + \mathbb{E}[B_s^2 | \mathcal{F}(B_s)] - t \\ &= \mathbb{E}[(B_t - B_s)^2] + B_s \mathbb{E}[B_t - B_s] + B_s^2 - t \\ &= t - s + 0 + B_s^2 - t \\ &= B_s^2 - s. \end{aligned}$$