
Quiz**April 14th, 2010****Introduction to Recursive Filtering and Estimation**
(151-0566-00)**Prof. R. D'Andrea**

Solutions

Duration: 45 minutes**Number of Problems:** 4**Permitted Aids:** None.Use only the prepared sheets for your solutions.

Solution 1**a) $c = 1$.**

In order for $f(x)$ to be a valid probability density function, the following property has to be satisfied:

$$\int_{-\infty}^{\infty} f(x) dx = 1.$$

That is, the constant c has to be chosen such that the area under the given function $f(x)$ is unity. By symmetry,

$$\int_{-\infty}^{\infty} f(x) dx = 4 \int_0^{0.5} f(x) dx = 4 A_{\text{triangle}} = c \cdot 1 = 1,$$

where the area of the four triangles $4 A_{\text{triangle}}$ result in a rectangle with side lengths of c and 1. Thus, the constant c is chosen to be $c = 1$.

b) $E[\mathbf{x}] = 1$.

Intuitively, the expected value can be obtained by viewing the probability density function $f(x)$ as a continuous mass distribution along a line and determining the corresponding center of mass.

When applying the definition of the expected value,

$$E[x] = \int_{-\infty}^{\infty} x f(x) dx,$$

the symmetry of the given function $f(x)$ can be exploited by shifting the function to the left and introducing $\xi = x - 1$,

$$\begin{aligned} E[x] &= \int_0^2 x f(x) dx \\ &= \int_{-1}^1 (\xi + 1) f(\xi + 1) d\xi = \int_{-1}^1 \xi g(\xi) d\xi + \int_{-1}^1 g(\xi) d\xi = 0 + 1 = 1, \end{aligned}$$

where $g(\xi) = f(\xi + 1)$ represents the left-shifted function. We use the fact that $\xi g(\xi)$ is an odd function and

$$\int_{-1}^1 g(\xi) d\xi = 1$$

was calculated in part a).

c) No.

The maximum likelihood estimate is given by

$$\hat{x}_{ML} = \arg \max_x f(x) = \{0.5, 1.5\},$$

whereas the expected value $E[x] = 1$.

d) Yes.

In the scalar case, the variance is defined as

$$\text{Var}(x) = E[(x - \bar{x})^2],$$

where $\bar{x} = E[x]$. For the given probability density function, that is

$$\text{Var}(x) = \int_0^2 \underbrace{(x-1)^2}_{\leq 1} f(x) dx \leq \int_0^2 f(x) dx = E[x] = 1.$$

Note that the two cases, $x = 0$ and $x = 2$, where $(x-1)^2 = 1$, have no contribution since $f(0) = f(2) = 0$. Thus, the variance $\text{Var}(x)$ is strictly smaller than 1.

e) **Yes.**

For any constant value b , the following relation holds:

$$E[x - b] = E[x] - b \quad (\text{law of the unconscious statistician}).$$

Thus,

$$\text{Var}(x - b) = E \left[\left(x - b - (E[x] - b) \right)^2 \right] = E \left[(x - E[x])^2 \right] = \text{Var}(x).$$

Problem 2**25%**

Please circle either ‘Yes’ or ‘No’ in the following questions.

Consider the joint probability density function $f(x, y)$ for the continuous random variables x and y :

$$f(x, y) = 6x^2y, \quad x, y \in [0, 1] .$$

a) The probability density function of x is $f(x) = x^2$.

Yes

No

b) The cumulative distribution function of y is $F(y) = y^2$.

Yes

No

c) The random variables x and y are independent.

Yes

No

d) The conditional probability density function of x conditioned on y is $f(x|y) = x^2$.

Yes

No

e) The probability of $x \leq 1$ and $y \leq 0.5$ is $\Pr(x \leq 1, y \leq 0.5) = 0.25$.

Yes

No

Note: For each question you get: +5 % for a correct answer, -5 % for an incorrect answer and 0 % for no answer. If you change your mind, please cross out all options (‘Yes’ and ‘No’) and write either ‘Yes’ or ‘No’ alongside, or leave it blank.

Solution 2

a) **No.**

The desired probability density function $f(x)$ is obtained by marginalization of the joint probability density function $f(x, y)$,

$$f(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_0^1 f(x, y) dy = 3x^2. \quad (1)$$

b) **Yes.**

Similar to a), we calculate

$$f(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_0^1 f(x, y) dx = 2y. \quad (2)$$

The cumulative distribution function $F(y)$ is given by

$$F(y) = \int_{-\infty}^y f(\tilde{y}) d\tilde{y} = y^2, \quad y \in [0, 1]. \quad (3)$$

c) **Yes.**

The random variables x and y are said to be independent if

$$f(x, y) = f(x)f(y).$$

With (1) and (2), the independence for the given function $f(x, y)$ is shown,

$$f(x)f(y) = 3x^2 \cdot 2y = 6x^2y = f(x, y).$$

d) **No.**

Because of the independence of x and y shown above, we obtain

$$f(x|y) = f(x) = 3x^2.$$

e) **Yes.**

Recalling the independence of x and y , Equation (3) and $x, y \in [0, 1]$, we get

$$Pr(x \leq 1, y \leq 0.5) = F_{x,y}(1, 0.5) = F_x(1)F_y(0.5) = 1 \cdot (0.5)^2 = 0.25.$$

Problem 3**25%**

A bag contains twelve coins: four coins are fair, six coins are biased and show tails with a probability $2/3$, and two coins are two-headed. A coin is chosen at random from the bag. The coin is tossed. It happens to be heads.

What is the probability that the chosen coin was a two-headed one?

Solution 3

We introduce the random variable $x \in \{r, b, w\}$ representing the outcome of choosing a coin at random from the bag. A fair coin is denoted by r , a biased coin by b , and w denotes a two-headed coin. The probabilities are as follows:

$$f_x(r) = \frac{1}{3}, \quad f_x(b) = \frac{1}{2}, \quad f_x(w) = \frac{1}{6}.$$

A second random variable $y \in \{h, t\}$ is introduced representing the outcome of flipping a coin, which can be either heads, $y = h$, or tails, $y = t$. When tossing the previously chosen coin, the conditional probabilities of obtaining heads are

$$f_{y|x}(h|r) = \frac{1}{2}, \quad f_{y|x}(h|b) = \frac{1}{3}, \quad f_{y|x}(h|w) = 1.$$

Bayes Theorem yields the desired value,

$$f_{x|y}(w|h) = \frac{f_{y|x}(h|w) f_x(w)}{f_y(h)},$$

where

$$f_y(h) = f_{y|x}(h|w) f_x(w) + f_{y|x}(h|b) f_x(b) + f_{y|x}(h|r) f_x(r)$$

with the total probability theorem. Given the numerical values above, we get

$$f_{x|y}(w|h) = \frac{1}{3}.$$

Note that $f_{x|y}(w|h) = f_{x|y}(b|h) = f_{x|y}(r|h)$.

Problem 4

25%

Consider the following estimation problem: an object B moves randomly on a circle with radius 1. From a given observation point P at location $x = L$, the angle ϕ can be measured, which describes the angle from the horizontal line to the line segment connecting P and B , see Figure 1. The goal is to estimate the location of the object B .

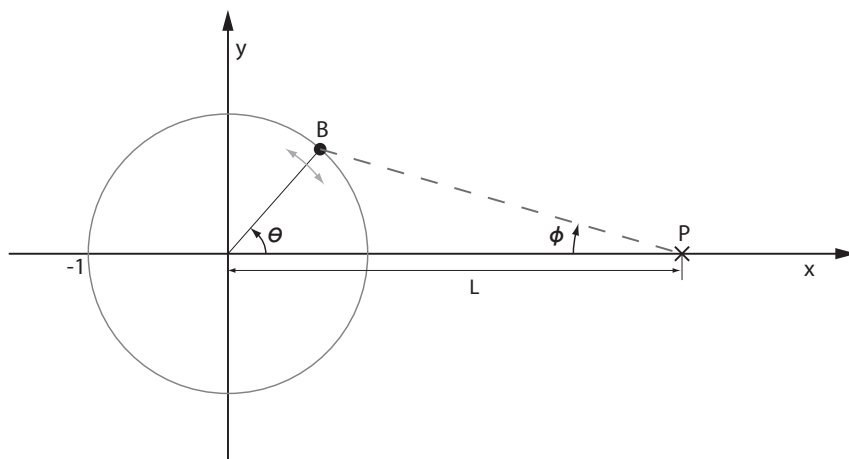


Figure 1

- a) The object B can only move in discrete steps. The object's location at time k is given by $x(k) \in \{0, 1, \dots, N - 1\}$, where

$$\theta(k) = 360 \frac{x(k)}{N}.$$

The angle $\theta(k)$ is in degrees. The dynamics are

$$x(k) = \text{mod}(x(k - 1) + v(k), N), \quad k = 1, 2, \dots$$

with $v(k) \in \{-1, 0, 1\}$. The object B moves clockwise, $v(k) = -1$, with probability p , it moves counter-clockwise, $v(k) = 1$, with probability p , and stays at its location, $v(k) = 0$, with probability $(1 - 2p)$, $p < 0.5$.

Note that $\text{mod}(N, N) = 0$ and $\text{mod}(-1, N) = N - 1$.

Determine the conditional probability density function $f(x(k)|x(k - 1))$.

- b) The angle sensor sitting at point P returns a measurement z of the angle ϕ corrupted by an additive noise w , that is

$$z = \phi + w.$$

The sensor noise w has the following probability density function:

$$f(w) = \begin{cases} \frac{1}{2} \left(1 + \frac{1}{2}w\right) & \text{for } -2 \leq w \leq 0 \\ \frac{1}{2} \left(1 - \frac{1}{2}w\right) & \text{for } 0 \leq w \leq 2 \\ 0 & \text{otherwise.} \end{cases}$$

The angle ϕ and the noise w are in degrees.

- Write ϕ as a function of θ .
- Determine the conditional probability density function $f(z|\theta)$.

Solution 4

- a) The conditional probability $f(x(k)|x(k-1))$ is directly derived from the object's dynamics,

$$f(x(k)|x(k-1)) = \begin{cases} p & \text{if } x(k) = \text{mod}(x(k-1) + 1, N) \\ p & \text{if } x(k) = \text{mod}(x(k-1) - 1, N) \\ 1 - 2p & \text{if } x(k) = x(k-1) \\ 0 & \text{otherwise.} \end{cases}$$

- b) The position of object B is given by its coordinates (x_B, y_B) , which can be expressed in terms of θ ,

$$x_B = \cos \theta, \quad y_B = \sin \theta.$$

For the angle ϕ , the following relation is derived

$$\tan \phi = \frac{y_B}{L - x_B} = \frac{\sin \theta}{L - \cos \theta}.$$

Thus,

$$\begin{aligned} \phi &= \arctan \left(\frac{\sin \theta}{L - \cos \theta} \right) \\ &= h(\theta), \end{aligned}$$

where the abbreviation $h(\cdot)$ is used in the following derivations.

Given the angle θ , the conditional probability density function of z can be determined using the relationship

$$w = z - \phi = z - h(\theta)$$

and the given probability distribution of w , resulting in

$$f(z|\theta) = \begin{cases} \frac{1}{2} \left(1 + \frac{1}{2}(z - h(\theta)) \right) & \text{for } -2 \leq z - h(\theta) \leq 0 \\ \frac{1}{2} \left(1 - \frac{1}{2}(z - h(\theta)) \right) & \text{for } 0 \leq z - h(\theta) \leq 2 \\ 0 & \text{otherwise.} \end{cases}$$