

ROB310-Problem Set 8

Problem 8.1

a) $E[X] = \sum_{x \in X} x f(x)$, $f(x) = ax^2$, $a = \frac{1}{6}$

$$\Rightarrow (-1)(\frac{1}{6}) + (0)(\frac{1}{6}) + (1)(\frac{1}{6}) + (2)(\frac{2}{3}) = \frac{4}{3}$$

$$\begin{aligned} \text{Var}[X] &= E[(X - E[X])^2] = E[(X - E[X])(X - E[X])] \\ &= E[X^2 - 2XE[X] + E[X]^2] = E[X^2] - 2E[XE[X]] + E[E[X]^2] \\ &= E[X^2] - E[X]^2 \\ &\Rightarrow \sum_{x \in X} x^2 f(x) - (\frac{4}{3})^2 = (1)(\frac{1}{6}) + (0)(\frac{1}{6}) + (1)(\frac{1}{6}) + (4)(\frac{2}{3}) - (\frac{16}{9}) \\ &= 11/9 \end{aligned}$$

b) $E[X] = \int_X x f(x) dx$, $f(x) = ax^2$, $a = \frac{1}{3}$

$$\Rightarrow \frac{1}{3} \int_{-1}^2 x^3 dx = \frac{1}{3} \left[\frac{x^4}{4} \right]_{-1}^2 = \frac{5}{4}$$

$$\begin{aligned} \text{Var}[X] &= E[X^2] - E[X]^2 = \int_{-1}^2 x^2 f(x) dx - (\frac{5}{4})^2 \\ &= \frac{1}{3} \int_{-1}^2 x^4 dx - \frac{25}{16} = \frac{1}{3} \left[\frac{x^5}{5} \right]_{-1}^2 - \frac{25}{16} = 51/80 \end{aligned}$$

c) The probability that y is in a small interval $[\bar{y}, \bar{y} + \Delta y]$.

$$\Rightarrow \Pr(y \in [\bar{y}, \bar{y} + \Delta y]) = \underbrace{\int_{\bar{y}}^{\bar{y} + \Delta y} f_y(y) dy}_{\approx f_y(\bar{y}) \Delta y \text{ as } \Delta y \rightarrow 0.}$$

\rightarrow w/ $\bar{x} = g(\bar{y})$, using Taylor's: $g(\bar{y} + \Delta y) \approx g(\bar{y}) + \frac{dg}{dy}(\bar{y}) \Delta y = \bar{x} + \Delta x$, $\Delta x < 0$.

$\rightarrow \Delta x < 0$ b/c $g(y) = \text{dec.}$

$$\rightarrow y \in [\bar{y}, \bar{y} + \Delta y] \Rightarrow g(y) \in [g(\bar{y}), g(\bar{y} + \Delta y)]$$

$$\Rightarrow x \in [\bar{x} + \Delta x, \bar{x}]$$

\rightarrow if Δy & Δx are small: $x \in [\bar{x} + \Delta x, \bar{x}] = y \in [\bar{y}, \bar{y} + \Delta y]$

$$\Rightarrow \Pr(y \in [\bar{y}, \bar{y} + \Delta y]) = f_y(\bar{y}) \Delta y$$

$$= \Pr(x \in [\bar{x} + \Delta x, \bar{x}]) = -f_x(\bar{x}) \Delta x$$

$$\Rightarrow f_y(\bar{y}) \Delta y = -f_x(\bar{x}) \Delta x \Rightarrow f_x(\bar{x}) = f_y(\bar{y}) / \left(-\frac{\Delta x}{\Delta y}\right)$$

$$\text{As } \Delta y \rightarrow 0 \text{ \& } \Delta x \rightarrow 0: \frac{\Delta x}{\Delta y} \rightarrow \frac{dg(\bar{y})}{dy}$$

$$\therefore f_x(\bar{x}) = f_y(\bar{y}) / \left(-\frac{dg(\bar{y})}{dy}\right) \Rightarrow f_x(x) = f_y(y) / \left(-\frac{dg(y)}{dy}\right)$$

→ Change of variables formula.

Problem 8.2

a) → Change of variables: $f_y(y) = \frac{f_x(x)}{\left(\frac{dg(x)}{dx}\right)}$

→ g = continuously differentiable, strictly monotonically inc. func.

$$\Rightarrow y = g(x) \Rightarrow dy = \frac{dg(x)}{dx} dx$$

$$\rightarrow W/ Y = \{y \mid \exists x \in X: y = g(x)\}$$

$$\Rightarrow E[Y] = \int_Y y f_y(y) dy = \int_X g(x) \left(\frac{f_x(x)}{\left(\frac{dg(x)}{dx}\right)}\right) \left(\frac{dg(x)}{dx} dx\right)$$

$$\Rightarrow E[Y] = \int_X g(x) f(x) dx$$

b) W/ Law of the unconscious statistician, vector random variable $\begin{bmatrix} x \\ y \end{bmatrix}$ w/ PDF $f_{xy}(x, y)$:

$$\Rightarrow E[g(x)h(y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x)h(y) f_{xy}(x, y) dx dy$$

→ Since x and y are independent:

$$\Rightarrow \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x) h(y) f_x(x) f_y(y) dx dy$$

$$\Rightarrow \int_{-\infty}^{\infty} g(x) f_x(x) dx \int_{-\infty}^{\infty} h(y) f_y(y) dy = E[g(x)] E[h(y)]$$

→ Similarly for discrete random variables:

$$\Rightarrow \sum_{x \in X} \sum_{y \in Y} g(x) h(y) f_{xy}(x, y) \dots \sum_{x \in X} g(x) f_x(x) \sum_{y \in Y} h(y) f_y(y)$$

w/ $\bar{x} = E[x]$ {

$$\begin{aligned} \text{Var}[X] &= \int_{-\infty}^{\infty} (x - \bar{x})^2 f(x) dx = \int_{-\infty}^{\infty} (x^2 - 2x\bar{x} + \bar{x}^2) f(x) dx \\ &= \int_{-\infty}^{\infty} x^2 f(x) dx - 2\bar{x} \int_{-\infty}^{\infty} x f(x) dx + \bar{x}^2 \int_{-\infty}^{\infty} f(x) dx \\ \text{Var}[X] &= E[X^2] - 2\bar{x}^2 + \bar{x}^2 = E[X^2] - \bar{x}^2 \end{aligned}$$

$$\rightarrow \text{Var}[X] \geq 0 \quad \therefore E[X^2] = \text{Var}[X] + \bar{x}^2 \geq \bar{x}^2$$

→ will be equality if $\text{Var}[X] = 0$.

Problem 8.3

$$\Pr(5 < X < 15) = \Pr(|X - 10| < 5)$$

$$\Rightarrow 1 - \Pr(|X - 10| \geq 5)$$

→ Chebyshev's inequality: $\Pr(|X - \mu| \geq a) \leq \frac{\sigma^2}{a^2}$
w/ $\mu = 10$, $\sigma^2 = 15$

$$\Rightarrow \Pr(|X - 10| \geq 5) \leq \frac{15}{25} = \frac{3}{5}$$

$$\Rightarrow \Pr(5 < X < 15) = 1 - \Pr(|X - 10| \geq 5) = 1 - \frac{3}{5} = \frac{2}{5}$$

Problem 8.4

→ 2 random variables: ^① box selected $x \in \{1, 2\}$.

$$\rightarrow f_x(1) = f_x(2) = 1/2$$

^② black or white marble $y \in \{b, w\}$.

$$\rightarrow f_{y|x}(b|1) = f_{y|x}(w|1) = 1/2$$

$$\rightarrow f_{y|x}(b|2) = \frac{2}{3}$$

$$\rightarrow f_{y|x}(w|2) = \frac{1}{3}$$

→ w/ total probability theorem:

$$f_y(b) = f_{y|x}(b|1) f_x(1) + f_{y|x}(b|2) f_x(2)$$

$$= \left(\frac{1}{2}\right)\left(\frac{1}{2}\right) + \left(\frac{2}{3}\right)\left(\frac{1}{2}\right) = \frac{7}{12}$$

Problem 8.5

a) → Integration of PDF = 1

$$\Rightarrow \int_{-\infty}^{\infty} f_{n_1}(\bar{n}_1) d\bar{n}_1 = \alpha_1 \left(\int_{-1}^0 (1 + \bar{n}_1) d\bar{n}_1 + \int_0^1 (1 - \bar{n}_1) d\bar{n}_1 \right)$$

$$= \alpha_1 \left(1 - \frac{1}{2} + 1 - \frac{1}{2} \right) = 1 \Rightarrow \boxed{\alpha_1 = 1}$$

$$\Rightarrow \int_{-\infty}^{\infty} f_{n_2}(\bar{n}_2) d\bar{n}_2 = \alpha_2 \left(\int_{-2}^0 \left(1 + \frac{1}{2}\bar{n}_2\right) d\bar{n}_2 + \int_0^2 \left(1 - \frac{1}{2}\bar{n}_2\right) d\bar{n}_2 \right)$$

$$= \alpha_2 (2 - 1 + 2 - 1) = 1 \Rightarrow \boxed{\alpha_2 = \frac{1}{2}}$$

b) → x = uniformly distributed $\Rightarrow f_x(x) = \begin{cases} \frac{1}{10} & \text{for } -5 \leq x \leq 5 \\ 0 & \text{otherwise.} \end{cases}$

→ b/c r.v.'s are independent:

$$f_{z_1, z_2|x}(z_1, z_2|x) = f_{z_1|x}(z_1|x) f_{z_2|x}(z_2|x) (*)$$

$$\rightarrow z_1 - x = \eta_1, \quad z_2 - x = \eta_2 :$$

$$\Rightarrow f_{z_1|x}(z_1|x) = \begin{cases} \alpha_1(1 - z_1 + x) & \text{for } 0 \leq z_1 - x \leq 1 \\ \alpha_1(1 + z_1 - x) & \text{for } -1 \leq z_1 - x \leq 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\Rightarrow f_{z_2|x}(z_2|x) = \begin{cases} \alpha_2(1 - \frac{1}{2}z_2 + \frac{1}{2}x) & \text{for } 0 \leq z_2 - x \leq 2 \\ \alpha_2(1 + \frac{1}{2}z_1 - \frac{1}{2}x) & \text{for } -2 \leq z_2 - x \leq 0 \\ 0 & \text{otherwise} \end{cases}$$

\rightarrow w/ Bayes' Theorem: $f_{z_1, z_2|x}(z_1, z_2|x)$ w/ \otimes

$$f_{x|z_1, z_2}(x|z_1, z_2) = \frac{f_{z_1|x}(z_1|x) f_{z_2|x}(z_2|x) f_x(x)}{\underbrace{\left(\int_{-5}^5 f_{z_1|x}(z_1|\bar{x}) f_{z_2|x}(z_2|\bar{x}) f_x(\bar{x}) d\bar{x} \right)}_{\text{normalization } n(z_1, z_2)}}$$

$$\rightarrow w/ z_1=0, z_2=0:$$

$$\otimes \Rightarrow \frac{1}{10} f_{z_1|x}(0|x) f_{z_2|x}(0|x) \Rightarrow \text{nor}(x).$$

$$\rightarrow w/ x \in [-5, -1]: \text{nor}(x) = 0.$$

$$\rightarrow w/ x \in [-1, 0]: \text{nor}(x) = \frac{1}{10} \alpha_1(1+x) \alpha_2(1+\frac{x}{2}) = \frac{1}{20}(1+x)(1+\frac{x}{2})$$

$$\rightarrow w/ x \in [0, 1]: \text{nor}(x) = \frac{1}{10} \alpha_1(1-x) \alpha_2(1-\frac{x}{2}) = \frac{1}{20}(1-x)(1-\frac{x}{2})$$

$$\rightarrow w/ x \in [1, 5]: \text{nor}(x) = 0.$$

$$\Rightarrow f_{x|z_1, z_2}(x|0, 0) = \frac{1}{n(0, 0)} \text{nor}(x)$$

$$\begin{aligned} \rightarrow n(0, 0) &= \int_{-1}^1 \text{nor}(x) dx = \frac{1}{20} \left[\int_{-1}^0 (1+x)(1+\frac{x}{2}) dx + \int_0^1 (1-x)(1-\frac{x}{2}) dx \right] \\ &= \frac{1}{20} \left(\frac{5}{12} + \frac{5}{12} \right) = \frac{1}{24} \end{aligned}$$

$$\therefore f_{X|Z_1, Z_2}(x|0, 0) = \begin{cases} 0 & \text{for } -5 \leq x \leq -1, 1 \leq x \leq 5 \\ \frac{12}{10} (1+x)(1+\frac{x}{2}) & \text{for } -1 \leq x \leq 0 \\ \frac{12}{10} (1-x)(1-\frac{x}{2}) & \text{for } 0 \leq x \leq 1. \end{cases}$$

c) W/ $Z_1=0, Z_2=1$:

$$nor(x) = \frac{1}{10} f_{Z_1|x}(0|x) f_{Z_2|x}(1|x)$$

$$\rightarrow \begin{matrix} \textcircled{1} \\ W/ x \in [-5, -1] \\ \textcircled{2} \\ x \in [1, 5] \end{matrix} \Rightarrow nor(x) = 0.$$

$\rightarrow W/ x \in [-1, 0]$:

$$nor(x) = \frac{1}{10} \alpha_1(1+x) \alpha_2(\frac{1}{2} + \frac{x}{2}) = \frac{1}{40} (1+x)^2$$

$\rightarrow W/ x \in [0, 1]$:

$$nor(x) = \frac{1}{10} \alpha_1(1-x) \alpha_2(\frac{1}{2} + \frac{x}{2}) = \frac{1}{40} (1-x^2)$$

$$\rightarrow n(0, 1) = \int_{-1}^1 nor(x) dx$$

$$= \frac{1}{40} \left[\int_{-1}^0 (1+x)^2 dx + \int_0^1 (1-x^2) dx \right]$$

$$= \frac{1}{40} \left(\frac{1}{3} + \frac{2}{3} \right) = \frac{1}{40}$$

$$\therefore f_{X|Z_1, Z_2}(x|0, 1) = \begin{cases} 0 & \text{for } -5 \leq x \leq -1, 1 \leq x \leq 5 \\ (1+x)^2 & \text{for } -1 \leq x \leq 0 \\ 1-x^2 & \text{for } 0 \leq x \leq 1. \end{cases}$$

d) W/ $Z_1=0, Z_2=3$:

$$nor(x) = \frac{1}{10} f_{Z_1|x}(0|x) f_{Z_2|x}(3|x)$$

→ w/ $x \in [-5, 5] \Rightarrow \text{nor}(x) = 0$.

→ this is b/c the intervals of pos. probability of $f_{z_1|x}(0|x)$ and $f_{z_2|x}(3|x)$ do not overlap.

$$\therefore f_{x|z_1, z_2}(x|0, 3) = 0.$$