

Stat-134, Section 02
Fall 2002

Solution of The Extra Credit Problem Set

1. Observe that from definition $X = R \cos \Theta$ and $Y = R \sin \Theta$.

- (a) **Step-1** : The range of values of Θ is $[0, 2\pi)$.
Step-2 : Fix $\alpha \in [0, 2\pi)$,

$$\begin{aligned} F_{\Theta}(\alpha) &= \mathbf{P}(\Theta \leq \alpha) \\ &= \int_{\theta \leq \alpha} \int \frac{1}{2\pi} e^{-\frac{1}{2}(x^2+y^2)} dx dy \\ &= \int_0^{\alpha} \int_0^{\infty} \frac{1}{2\pi} e^{-\frac{r^2}{2}} r dr d\theta \\ &= \left(\int_0^{\alpha} \frac{d\theta}{2\pi} \right) \left(\int_0^{\infty} r e^{-\frac{r^2}{2}} dr \right) \\ &= \frac{\alpha}{2\pi} \end{aligned}$$

So the CDF of Θ is given by

$$F_{\Theta}(\alpha) = \begin{cases} 0 & \text{if } \alpha \leq 0 \\ \frac{\alpha}{2\pi} & \text{if } 0 < \alpha \leq 2\pi \\ 1 & \text{if } \alpha > 2\pi. \end{cases}$$

Clearly, $\Theta \sim \text{Unif}(0, 2\pi)$.

- (b) Observe that (Z, W) is nothing but a rotation of (X, Y) by an angle of $\theta = \frac{\pi}{4}$. Now recall the following fact proved in the class
Fact : If X and Y are i.i.d Normal(0, 1) random variables and (Z, W) is a rotation of (X, Y) then (Z, W) are also i.i.d Normal(0, 1).
So using this we conclude that (Z, W) are i.i.d. Normal(0, 1) random variables.

2. (a) Notice that $F : \mathbb{R} \rightarrow (0, 1)$ is defined as $F(x) = \int_{-\infty}^x f(t) dt$, where $f(t) > 0$ for all $-\infty < t < \infty$. Thus F is a strictly increasing function, and hence $F^{-1} : (0, 1) \rightarrow \mathbb{R}$ exists.
(b) Observe that the range of values of $Y = F^{-1}(U)$ is \mathbb{R} . Fix $-\infty < y < \infty$, then

$$\begin{aligned} F_Y(y) &= \mathbf{P}(Y \leq y) \\ &= \mathbf{P}(F^{-1}(U) \leq y) \\ &= \mathbf{P}(U \leq F(y)) \\ &= F(y) \end{aligned}$$

The last equality follows from the fact that $U \sim \text{Unif}(0, 1)$, and $F(y) \in (0, 1)$. So the CDF of $Y = F^{-1}(U)$, is F . In otherwords Y has the same distribution as of X .

(c) First note that from definition $W = F(X) \in (0, 1)$. Fix $w \in (0, 1)$.

$$\begin{aligned}
 F_W(w) &= \mathbf{P}(W \leq w) \\
 &= \mathbf{P}(F(X) \leq w) \\
 &= \mathbf{P}(X \leq F^{-1}(w)) \\
 &= F(F^{-1}(w)) \\
 &= w
 \end{aligned}$$

So clearly $W \sim \text{Unif}(0, 1)$.

3. Suppose U_1, U_2, U_3, \dots be the **i.i.d.** sequence of $\text{Unif}(0, 1)$ random variables produced by the computer routine.

(a) Define $W_1 := -\frac{1}{\lambda} \log U_1$, and use change of variable formula to conclude that $W_1 \sim \text{Exponential}(\lambda)$ (done in class).

(b) For each $i \geq 1$, define $W_i := -\frac{1}{\lambda} \log U_i$. So using part (a), W_1, W_2, W_3, \dots are **i.i.d.** $\text{Exponential}(\lambda)$. Let $T_r := W_1 + W_2 + \dots + W_r$, where $r \geq 1$. Define N to be the first r such that $T_{r+1} \geq 1$; that is, $N = 0$ if $T_1 \geq 1$; $N = 1$ if $T_1 < 1$ but $T_2 \geq 1$, $N = 2$ if $T_2 < 1$ but $T_3 \geq 1$, and so on. Naturally, N is nothing but the number of arrivals in $[0, 1)$ for the Poisson arrival process of rate λ defined through the Exponential variables W_1, W_2, W_3, \dots , so clearly $N \sim \text{Poisson}(\lambda)$.

4. First note that $\mathbf{E}[Z] = \mathbf{E}[(X - Y)^2] = \mathbf{E}[X^2] + \mathbf{E}[Y^2] - 2\mathbf{E}[XY]$. Now using the *Law of Iterated Expectations* and the first given condition we get

$$\begin{aligned}
 \mathbf{E}[XY] &= \mathbf{E}[\mathbf{E}[XY|Y]] \\
 &= \mathbf{E}[Y\mathbf{E}[X|Y]] \\
 &= \mathbf{E}[Y^2]
 \end{aligned}$$

Similarly using the second condition we will get $\mathbf{E}[XY] = \mathbf{E}[X^2]$. So $\mathbf{E}[Z] = 0$, hence using the given math fact we conclude that $\mathbf{P}(Z = 0) = 1 \Leftrightarrow \mathbf{P}(X = Y) = 1$.

5. Let A_i be the event that i^{th} ball goes into i^{th} box, for $1 \leq i \leq n$. We know from the Midterm Exam Problem # 3 that $\mathbf{P}(A_i) = (1 - 1/n)^n$ for any $1 \leq i \leq n$, and $\mathbf{P}(A_i \cap A_j) = (1 - 2/n)^n$ for $1 \leq i \neq j \leq n$. Let I_i be the indicator function of the event A_i . Naturally from definition $X = I_1 + I_2 + \dots + I_n$. Thus $\mathbf{E}[X] = n \times (1 - 1/n)^n$. Notice that

$$\begin{aligned}
 X^2 &= \sum_{i=1}^n I_i^2 + \sum_{1 \leq i \neq j \leq n} I_i I_j \\
 &= \sum_{i=1}^n I_i + \sum_{1 \leq i \neq j \leq n} I_{A_i \cap A_j}
 \end{aligned}$$

So by taking expectation we get

$$\begin{aligned}
 \mathbf{E}[X^2] &= \sum_{i=1}^n \mathbf{P}(A_i) + \sum_{1 \leq i \neq j \leq n} \mathbf{P}(A_i \cap A_j) \\
 &= n \times \left(1 - \frac{1}{n}\right)^n + n(n-1) \times \left(1 - \frac{2}{n}\right)^n
 \end{aligned}$$

Finally,

$$\begin{aligned}
 \mathbf{Var}(X) &= \mathbf{E}[X^2] - (\mathbf{E}[X])^2 \\
 &= n \left(1 - \frac{1}{n}\right)^n + n(n-1) \left(1 - \frac{2}{n}\right)^n - n^2 \left(1 - \frac{1}{n}\right)^{2n}
 \end{aligned}$$

6. (a) From the formula for the joint density we get that both X and Y takes values in $(0, 1)$, and have the same distribution. Fix $0 < x < 1$, then

$$\begin{aligned} f_X(x) &= \int_0^1 (2x + 2y - 4xy) dy \\ &= [2xy + y^2 - 2xy^2]_0^1 \\ &= 2x + 1 - 2x = 1. \end{aligned}$$

Thus X and Y are $\text{Unif}(0, 1)$.

- (b)

$$f_{Y|X}\left(y \middle| x = \frac{1}{4}\right) = \frac{f\left(\frac{1}{4}, y\right)}{f_X\left(\frac{1}{4}\right)} = \begin{cases} \frac{2}{4}y + 2y - \frac{4}{4}y = \frac{1}{2} + y & \text{if } 0 < y < 1, \\ 0 & \text{otherwise} \end{cases}$$

(c) $\mathbf{E}[Y|X = \frac{1}{4}] = \int_0^1 y\left(\frac{1}{2} + y\right) dy = \frac{7}{12}$

7. (a) Observe that Y only takes positive values. Fix $y > 0$,

$$\begin{aligned} f_Y(y) &= \int_0^y \lambda^3 x e^{-\lambda y} dx \\ &= \frac{\lambda^3}{2} y^2 e^{-\lambda y} \end{aligned}$$

Thus $Y \sim \text{Gamma}(3, \lambda)$. Hence $\mathbf{E}[Y] = \frac{3}{\lambda}$.

- (b) The conditional density of X given $Y = 1$ is

$$f_{X|Y}\left(x \middle| y = 1\right) = \frac{f(x, 1)}{f_Y(1)} = \begin{cases} \frac{\lambda^3 x e^{-\lambda y}}{(\lambda^3/3)y^2 e^{-\lambda y}} = 2x & \text{if } 0 < x < 1, \\ 0 & \text{otherwise.} \end{cases}$$

Hence $\mathbf{E}[X|Y = 1] = \int_0^1 x(2x) dx = \frac{2}{3}$.

8. (a) Observe that Y only takes positive values. Fix $y > 0$,

$$\begin{aligned} f_Y(y) &= \int_0^y \frac{e^{-y/2} dx}{2\pi\sqrt{x(y-x)}} \\ &= \frac{1}{2\pi} e^{-y/2} \int_0^y \frac{dx}{\sqrt{x(y-x)}} \\ &= \frac{c}{2\pi} e^{-y/2}, \end{aligned}$$

where $c = \int_0^1 ds/\sqrt{s(1-s)}$, and the last equality can be obtained by substituting $x = ys$ in the previous integral. Since f_Y is a density so it must integrate out to 1, and so without any difficult integration evaluation we can show that $c = \pi$. Hence $Y \sim \text{Exponential}(1/2)$.

- (b) The conditional density of X given $Y = 1$ is

$$f_{X|Y}\left(x \middle| y = 1\right) = \frac{f(x, 1)}{f_Y(1)} = \begin{cases} \frac{e^{-y/2}/(\pi\sqrt{x(1-x)})}{(1/2)e^{-y/2}} = \frac{1}{\pi} \frac{1}{\sqrt{x(1-x)}} & \text{if } 0 < x < 1, \\ 0 & \text{otherwise.} \end{cases}$$

Thus Given $Y = 1$, X is a random variable which takes values in $(0, 1)$. Notice (by inspection, or by using change of variable formula) that given $Y = 1$, $1 - X$ has the same distribution as that of X . Thus clearly with any difficult calculations we can conclude that $\mathbf{E}[X|Y = 1] = \frac{1}{2}$.