

ii. The joint density of  $f(x,y)$  must equal one.

$$\int \int_0^1 c(x^2 + 3y) dx dy = 1$$

$$\begin{aligned} \int \int_D f(x,y) &= c \int_0^1 \left[ \left[ \frac{x^3}{3} + 3xy \right]_{x=0}^{x=1} \right] dy \\ &= c \int_0^1 \left( \frac{1}{3} + 3y \right) dy \\ &= c \left[ \frac{1}{3}y + \frac{3}{2}y^2 \right]_{y=0}^{y=1} \\ &= c \left( \frac{1}{3} + \frac{3}{2} \right) \\ &= 11c/6 \end{aligned}$$

$$11c/6 = 1 \quad \text{so} \quad \boxed{c = 6/11}$$

iii. We can find the marginal density of  $X$  by integrating the joint density function across all values  $\in \mathbb{R}(Y)$

$$f_X(x) = \int_D f(x,y) dy = \int_0^1 c(x^2 + 3y) dy = c \int_0^1 (x^2 + 3y) dy$$

$$= c \left[ \frac{x^2 y}{2} + \frac{3y^2}{2} \right]_{y=0}^{y=1} = c \left( x^2 + \frac{3}{2} \right)$$

$$f_X(x) = \begin{cases} \frac{6}{11} (x^2 + \frac{3}{2}) & \text{if } x \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$

iv.  $f_Y(y | X=\frac{1}{2}) = \frac{f(\frac{1}{2}, y)}{P_X(\frac{1}{2})} = \frac{\frac{6}{11}(\frac{1}{4} + 3y)}{\frac{6}{11}(\frac{1}{4} + \frac{3}{2})} = \frac{\frac{1}{4} + 3y}{\frac{1}{4} + \frac{3}{2}} = \frac{4}{7}(\frac{1}{4} + 3y) = \frac{1}{7} + \frac{12}{7}y$

$$f_Y(y | X=\frac{1}{2}) = \begin{cases} \frac{1}{7} + \frac{12}{7}y & \text{if } y \in [0, 1] \\ 0 & \text{otherwise} \end{cases}$$

v.  $X$  and  $Y$  are independent iff  $f_Y(y | X=\frac{1}{2}) = f_Y(y)$

$$f_Y(y) = \int_D f(x,y) dx = \int_0^1 c(x^2 + 3y) dx = c \left[ \frac{x^3}{3} + 3xy \right]_{x=0}^{x=1} = c(\frac{1}{3} + 3y) = \frac{6}{11}(\frac{1}{3} + 3y) = \frac{2}{11} + \frac{18}{11}y$$

$f_Y(y) \neq f_Y(y | X=\frac{1}{2})$  so the marginal distribution of  $Y$  depends on the value of  $X$ , and thus  $X$  and  $Y$  are dependent

2.i. The marginal density  $f_X(x)$  is given by integrating the joint density function  $f(x,y)$  across all  $y$  in the domain.

$$f_X(x) = \int_0^x 3x \, dy \quad \text{for } x \in [0,1]$$

$$= 3xy \Big|_{y=0}^{y=x} = 3x^2$$

$$f_X(x) = \begin{cases} 3x^2 & \text{if } x \in [0,1] \\ 0 & \text{otherwise} \end{cases}$$

ii. The conditional density  $f_Y(y|X=\frac{1}{2})$  is given by

$$f_Y(y|X=\frac{1}{2}) = \frac{f(\frac{1}{2}, y)}{f_X(\frac{1}{2})} = \frac{\frac{3}{2}}{\frac{3}{2} \cdot \frac{1}{2}^2} = \begin{cases} 2 & \text{for } y \in [0, \frac{1}{2}] \\ 0 & \text{otherwise} \end{cases}$$

compute  $\int f(x,y) dx$  for practice

iii. ①  $E[X|Y=y] = \int_x f_X(x|Y=y) dx$

We begin by finding  $f_X(x|Y=y)$

$$f_X(x|Y=y) = \frac{f(x,y)}{\int_x f(x,y) dx} \rightarrow \text{marginal of } X$$

$$= \frac{3x}{\int_y^x 3x \, dx} = \frac{3x}{\frac{3}{2}x^2 \Big|_y^x} = \frac{3x}{\frac{3}{2} - \frac{3}{2}y^2}$$

Then plug  $f_X(x|Y=y)$  into ①

$$E[X|Y=y] = \int_y^1 x \left( \frac{3x}{\frac{3}{2} - \frac{3}{2}y^2} \right) dx = \frac{1}{\frac{3}{2} - \frac{3}{2}y^2} \int_y^1 3x^2 dx$$

$$= \frac{1}{\frac{3}{2} - \frac{3}{2}y^2} \times \frac{x^3}{3} \Big|_y^1 = \frac{1 - y^3}{\frac{3}{2} - \frac{3}{2}y^2}$$

3.

Named distribution

$$Y \sim N(\mu, \sigma^2) \quad (\text{normal w/ mean } \mu, \text{ variance } \sigma^2)$$

CDF

$$\begin{aligned} P(Y \leq y) &= P(\mu + \sigma X \leq y) \\ &= P(X \leq \frac{y-\mu}{\sigma}) \\ &= \Phi\left(\frac{y-\mu}{\sigma}\right) \end{aligned}$$

PDF:  $f_Y(y) = \frac{d}{dy} \Phi\left(\frac{y-\mu}{\sigma}\right)$  derivative of CDF

$$= \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(y-\mu)^2}{2\sigma^2}}$$

4.1. IF  $X \sim \text{Unif}(0, 1)$ ,

$$\begin{aligned} E[e^{2X}] &= \int_0^1 f_X(x) e^{2x} dx && \text{by Law of Unconscious Statistician} \\ &= \int_0^1 e^{2x} dx && \text{since } f_X(x)=1 \text{ for } X \sim \text{Unif}(0, 1) \text{ on } [0, 1] \\ &= \frac{1}{2} e^{2x} \Big|_0^1 \\ &= \frac{1}{2} (e^2 - 1) \end{aligned}$$

$$\text{4iii. } \mathbb{E}[e^{rx}] = \int_{-\infty}^{\infty} e^{-(\frac{x-\mu}{\sigma})^2/2} e^{rx} dx$$

$$= \int_{-\infty}^{\infty} e^{-[\frac{(x-\mu)^2}{2\sigma^2} - \frac{x\mu}{\sigma^2} + (\frac{\mu}{\sigma})^2] + rx} dx \quad (1)$$

We will now complete the square in the exponent

$$\text{The integrand above} = -\left[\frac{(x-\mu)^2}{2\sigma^2} - \left(\frac{\mu}{\sigma^2} + r\right)x + \left(\frac{\mu}{\sigma}\right)^2\right] \quad (2)$$

To complete the square, we wish to find some term,  $y$

$$\text{s.t. } 2\left(\frac{x-\mu}{\sigma^2}\right)(y) = \left(\frac{\mu}{\sigma^2} + r\right)x$$

$$\text{Solving for } y, \frac{2}{\sigma^2}y = \left(\frac{\mu}{\sigma^2} + r\right), \text{ so } y = \frac{1}{2\sigma^2}\left(\frac{\mu}{\sigma} + r\sigma^2\right) = \frac{1}{2\sigma}\left(\frac{\mu + r\sigma^2}{\sigma}\right)$$

$$\text{So (2)} = -\left[\left(\frac{(x-\mu)^2}{2\sigma^2} - \left(\frac{\mu + r\sigma^2}{2\sigma}\right)\right)^2 - \frac{\mu + r\sigma^2}{2\sigma} + \left(\frac{\mu}{\sigma}\right)^2\right] \quad (3)$$

$$\text{Let } u = \frac{x-\mu}{\sigma} - \frac{\mu + r\sigma^2}{2\sigma}. \text{ Then } du = \frac{1}{\sigma}dx$$

Substituting  $u$  and (3) into (1),

$$e^{\left[\left(\frac{\mu}{\sigma}\right)^2 - \frac{\mu + r\sigma^2}{2\sigma}\right] \int_{-\infty}^{\infty} e^{u^2} \sigma \sqrt{2} du} = \sigma \sqrt{2} e^{\left[\left(\frac{\mu}{\sigma}\right)^2 - \frac{\mu + r\sigma^2}{2\sigma}\right] \int_{-\infty}^{\infty} e^{u^2} du}$$

I believe I failed to complete the square properly, because my solution terminates in an integral I can't solve.

5

5i. Compute  $f_x(x)$ We begin by computing  $F_x(x)$ 

$$F_x(x) = \Pr(X \leq x) = 1 - \Pr(X > x)$$

$$= 1 - \Pr(U_1 > x) \cap (U_2 > x)$$

$$= 1 - \Pr(U_1 > x)^2$$

$$= 1 - (1-x)^2$$

since  $U_1, U_2$  are iidsince  $U_i \sim \text{Uniform}(0,1)$ 

$$\text{Now } f_x(x) = \frac{d}{dx} (1 - (1-x)^2) = -2(1-x)(-1)$$

$$\begin{cases} 2-2x & \text{if } x \in [0,1] \\ 0 & \text{otherwise} \end{cases}$$

5ii. Compute  $\Pr(X < \frac{1}{4} | Y \geq \frac{1}{2})$ 

Per the definition of conditional probability, this is

$$\frac{\Pr(X < \frac{1}{4}, Y \geq \frac{1}{2})}{\Pr(Y \geq \frac{1}{2})} \quad (1)$$

To compute  $\Pr(Y \geq \frac{1}{2})$ , we must find  $F_Y(y)$ 

$$F_Y(y) = \Pr(Y \leq y) = \Pr(U_1 \leq y, U_2 \leq y)$$

$$= \Pr(U_1 \leq y)^2 \quad \text{since } U_1, U_2 \text{ are iid}$$

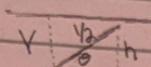
$$= y^2 \quad \text{by } U_i \sim [0,1] \quad \text{since } U_i \sim \text{Uniform}(0,1)$$

$$\Pr(Y \geq \frac{1}{2}) = 1 - F_Y(\frac{1}{2}) = 1 - (\frac{1}{2})^2 = \frac{3}{4}$$

To compute  $\Pr(X < \frac{1}{4}, Y \geq \frac{1}{2})$ , we note that to satisfy both conditions,one of the uniform r.v.s  $< \frac{1}{4}$  and the other  $\geq \frac{1}{2}$ . Since they are iid  $\sim \text{Unif}(0,1)$ ,the probability of this happening is  $(\frac{1}{4})(\frac{1}{2})(2 \text{ vors}) = \frac{1}{4}$ 

$$\text{Plugging everything back into (1), } \Pr(X < \frac{1}{4}, Y \geq \frac{1}{2}) = \frac{\frac{1}{4}}{\frac{3}{4}} = \boxed{\frac{1}{3}}$$

6i.

diagram: 

For a needle to touch the crack, we need  $h \geq Y$ , which means

$$\left(\frac{1}{2}\right) \sin \theta \geq Y$$

$$Y \leq \frac{1}{2} \sin \theta$$

$\Theta \sim \text{Unif}(0, \pi/2)$ ,  $Y \sim \text{Unif}(0, 1)$  and they are independent, so the following graph reflects their joint density.

1

$$Y = \frac{1}{2} \sin \theta$$

We wish to find  $P(Y \leq \frac{1}{2} \sin \theta)$ ,

which is equivalent to the area under  $y = \frac{1}{2} \sin \theta$  from  $\theta = 0$  to  $\theta = \pi/2$

divided by the total area of the box

$$\text{So, } P(\text{NeedleIntersection}) = P(Y \leq \frac{1}{2} \sin \theta)$$

$$= \frac{\int_0^{\pi/2} \frac{1}{2} \sin \theta d\theta}{(1)(\pi/2)} = \frac{1}{\pi} \int_0^{\pi/2} \sin \theta d\theta$$

$$= \frac{1}{\pi} (-\cos \theta) \Big|_0^{\pi/2} = \frac{1}{\pi}$$

6ii.  $\text{Var}(H_n) = \text{Var}\left(\frac{1}{N} \sum_{k=1}^N T_k\right) = \frac{1}{N^2} \text{Var}\left(\sum_{k=1}^N T_k\right)$  Since Cov is inner product

$$\begin{aligned}
 &= \frac{1}{N^2} \sum_{k=1}^N \text{Var}(T_k) \quad \text{because the } T_k \text{ are independent} \\
 &= \frac{1}{N^2} \sum_{k=1}^N \frac{1-p}{p^2} \quad \text{Variance of geo r.v.} \\
 &= \frac{1}{N^2} \sum_{k=1}^N \frac{(1-\pi)}{(\pi)^2} \\
 &= \frac{1}{N^2} \sum_{k=1}^N \left(\frac{\pi-1}{\pi}\right) (\pi)^2 \\
 &= \frac{1}{N^2} \sum_{k=1}^N \pi(\pi-1) \\
 &= \frac{N\pi(\pi-1)}{N^2} \\
 &= \frac{\pi(\pi-1)}{N}
 \end{aligned}$$

And  $E[H_n] = \pi$

6iii. We can set this problem up using the CLT.

We want  $P(3.14158 \leq H_n \leq 3.14160) = P\left(\frac{\sum T_k - N\pi}{\sqrt{N}} \leq \frac{N\pi - N\pi + 0.001}{\sqrt{N}}\right)$  def of  $H_n$

$$\begin{aligned}
 &= P\left(\frac{N \cdot 3.14158 - N\pi}{\sqrt{N}} \leq \frac{\sum T_k - N\pi}{\sqrt{N}} \leq \frac{N \cdot 3.14160 - N\pi}{\sqrt{N}}\right) \\
 &= P\left(\frac{3.14158 - \pi}{\sqrt{N}} \leq \frac{\sum T_k - N\pi}{\sqrt{N}} \leq \frac{3.14160 - \pi}{\sqrt{N}}\right)
 \end{aligned}$$

We found in 6ii that  $\sigma^2 = \frac{\pi-1}{N}$ , so  $\sigma = \sqrt{\frac{\pi-1}{N}} = \frac{\sqrt{\pi-1}}{\sqrt{N}}$

$$\mu = E[H_n] = \pi$$

$$So P\left(\frac{(-10^{-5})(N)}{\sqrt{N}} \leq \frac{\sum T_k - N\pi}{\sigma\sqrt{N}} \leq \frac{(10^{-5})(N)}{\sqrt{\pi(\pi-1)}\sqrt{N}}\right)$$

$$= P\left(\frac{(-10^{-5})(N)}{\sqrt{\pi(\pi-1)}} \leq \frac{\sum T_k - N\pi}{\sigma\sqrt{N}} \leq \frac{(10^{-5})(N)}{\sqrt{\pi(\pi-1)}}\right)$$

$$\text{Let } a = \frac{(10^{-5})}{\sqrt{\pi(\pi-1)}}$$

$$t = \frac{1}{\sqrt{N}} \int_{-aN}^{aN} e^{-\frac{x^2}{2}} dx = \Phi(aN) - \Phi(-aN).$$

Using an online calculator, I find this expression  $\approx 0.90$  when  $N \approx 428,700$

Note: This makes  $aN \approx 1.65$ , which is the crit value for a 2-tailed T test at 90% confidence