

In some places you might want to use the Leibniz Integral Rule, which states that when everything is nice and differentiable we have

$$\frac{d}{dx} \int_{u(x)}^{v(x)} g(x, t) dt = g(x, v(x)) \frac{dv}{dx} - g(x, u(x)) \frac{du}{dx} + \int_{u(x)}^{v(x)} \frac{\partial}{\partial x} g(x, t) dt.$$

- 1.** In class we noted that from the expected value rule for joint pdfs, which says

$$\mathbb{E}(g(X, Y)) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X,Y}(x, y) dx dy,$$

it follows that

$$\mathbb{E}(X + Y) = \mathbb{E}(X) + \mathbb{E}(Y).$$

- (a) Show that expected value rule implies the last identity above.
- (b) Let  $Z = X + Y$ . Given  $z$ , sketch the subset  $W$  of the  $x$ - $y$  plane with the property that  $Z \leq z$  if and only if  $(X, Y) \in W$ .
- (c) Show that

$$f_Z(z) = \int_{-\infty}^{\infty} f_{X,Y}(x, z-x) dx.$$

(Suggestion: find the cdf of  $Z$  first, then differentiate. The result of (b) might help.)

- (d) Compute  $\mathbb{E}(Z)$  using the result of (c) along with the definition

$$\mathbb{E}(Z) = \int_{-\infty}^{\infty} z f_Z(z) dz.$$

Verify that your answer agrees with what you know to be true, namely that  $\mathbb{E}(Z) = \mathbb{E}(X) + \mathbb{E}(Y)$ .

- 2.** This problem is about marginal pdfs not determining joint pdfs.

- (a) Suppose  $X$  and  $Y$  are jointly uniform on the unit square  $[0, 1] \times [0, 1]$ , i.e.

$$f_{X,Y}(x, y) = \begin{cases} 1 & \text{when } 0 \leq x, y \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

Find the marginal pdfs  $f_X(x)$  and  $f_Y(y)$ .

- (b) Now suppose instead that  $X$  and  $Y$  have joint pdf

$$f_{X,Y}(x, y) = \begin{cases} 2x(1-y) + 2y(1-x) & \text{when } 0 \leq x, y \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$

Find the marginal pdfs  $f_X(x)$  and  $f_Y(y)$ .

- 3.** Sam breaks a stick of length 1 at a point  $X$  distributed uniformly over the interval  $(0, 1)$ . As a function of  $p \in (0, 1)$ , find the expected length of the piece that contains the point  $p$ .

- 4.** The time in minutes that it takes Maddy to commute to work is a continuous random variable  $X$  with cdf  $F_X(x)$ . If she is  $s$  minutes late to work, she incurs a cost  $\alpha s$ , where  $\alpha > 0$ . If she is  $s$  minutes early, she incurs a cost  $\beta s$ , where  $\beta > 0$ .

- (a) Maddy would like to determine how many minutes before the start of work she should depart so as to minimize her expected cost. Show that the optimal solution  $t^*$  satisfies

$$F_X(t^*) = \frac{\alpha}{\alpha + \beta}.$$

- (b) How can we interpret the parameter  $\gamma = \alpha/(\alpha + \beta)$ ?

Now suppose  $X$  is uniformly distributed over the interval  $[0, t_{\max}]$ .

- (c) Find the mean  $\mu = \mathbb{E}(X)$  and standard deviation  $\sigma = \sqrt{\text{Var}(X)}$ .
- (d) Calculate  $t^*$  from part (a) in terms of  $\gamma$ ,  $\mu$ , and  $\sigma$ .
- (d) Describe how the optimal solution  $t^*$  depends on the mean and variance of  $X$ .

- 5.** The input  $X$  to a binary communication channel is either  $+c$  or  $-c$  with respective probabilities  $1/4$  and  $3/4$ . Here,  $c \geq 0$  is given. The channel output is

$$Y = X + N,$$

where  $N$  is Gaussian with zero mean and variance 1. In parts (b) and (c), please provide closed-form expressions for your answers. In part (c), you may express your answer in terms of the standard normal cdf  $\Phi$ .

- (a) Find the conditional pdf  $f_{Y|A}(y)$ , where  $A$  is the event  $\{X = x\}$ . What kind of pdf is this?
- (b) Find the marginal pdf  $f_Y(y)$ . Plot  $f_Y(y)$  on the interval  $-10 \leq y \leq 10$  for  $c = 0$  and  $c = 5$ .
- (c) Find  $\mathbb{P}(\{X = c\} \mid \{Y > 0\})$ . What happens to this probability as  $c$  gets larger? Why does this make intuitive sense?

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HW9 ECE 3100

①

$$\mathbb{E}(g(X, Y)) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} g(x, y) f_{X,Y}(x, y) dx dy$$

(a)  $g(X, Y) = X + Y$

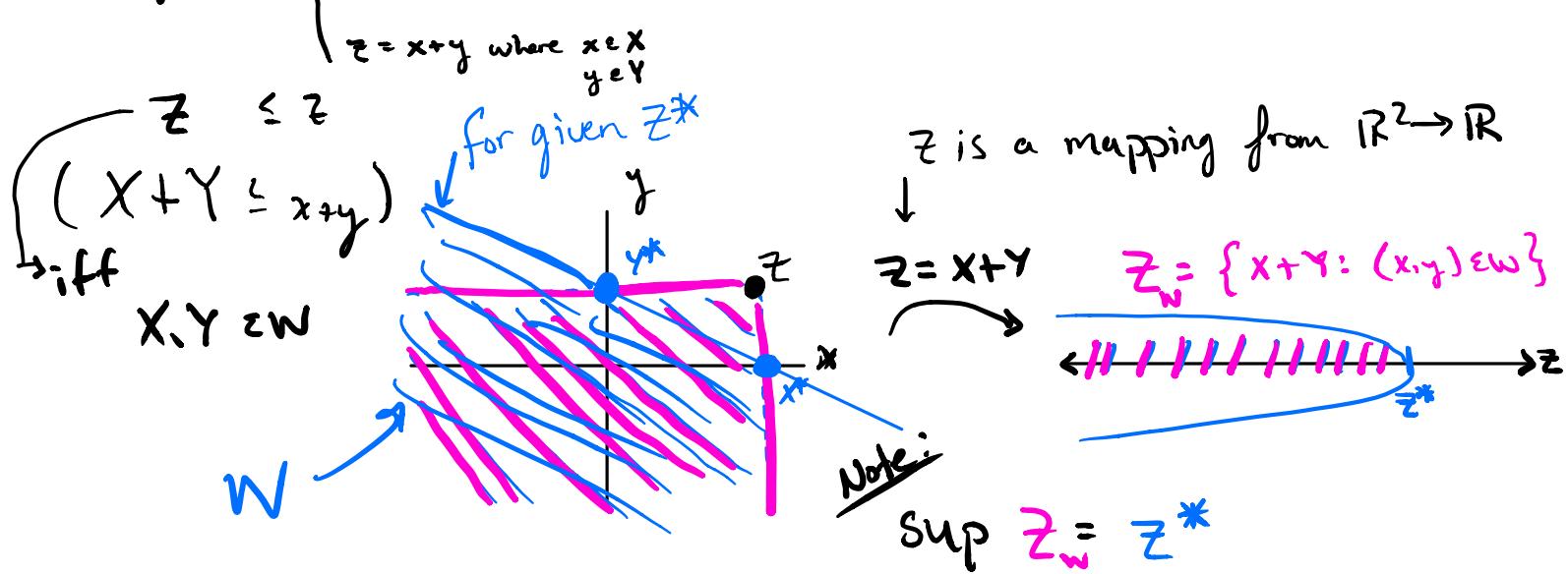
$$\mathbb{E}(X + Y) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} (x+y) f_{X,Y}(x, y) dx dy$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} x \underbrace{\left( f_{X,Y}(x, y) dy \right)}_{f_X(x)} dx + \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} y \underbrace{\left( f_{X,Y}(x, y) dx \right)}_{f_Y(y)} dy$$

$$= \mathbb{E}(X) + \mathbb{E}(Y)$$

(b)  $Z = X + Y$  ( $Z = g(X, Y)$ )  $g: \mathbb{R}^2 \rightarrow \mathbb{R}$

given  $z^*$ , sketch  $W \subset \mathbb{R}^2$  w/ property  $Z \leq z$  iff  $(x, y) \in W$



(c) Show

$$f_z(z) = \int_{-\infty}^{+\infty} f_{x,y}(x, z-x) dx$$

$$F_z(z) = P(\{Z \leq z\}) = \int_{x=-\infty}^{+\infty} \int_{y=-\infty}^{z-x} f_{x,y}(x, y) dy dx$$

$$f_z(z) = \frac{d}{dz} F_z(z) = \int_{x=-\infty}^{+\infty} \left( \frac{d}{dz} \int_{y=-\infty}^{z-x} f_{x,y}(x, y) dy \right) dx$$

$$= \int_{x=-\infty}^{+\infty} f_{x,y}(x, z-x) dx \quad \} \text{ By Leibniz Rule}$$

$$(d) |E(Z)| = \int_{z=-\infty}^{z=+\infty} z f_z(z) dz = \int_{z=-\infty}^{z=+\infty} z \left( \int_{x=-\infty}^{x=+\infty} f_{x,y}(x, z-x) dx \right) dz$$

$$= \int_{x=-\infty}^{x=+\infty} \left( \int_{z=-\infty}^{z=+\infty} z f_{x,y}(x, z-x) dz \right) dx$$

$$= \int_{x=-\infty}^{x=+\infty} \int_{y=-\infty}^{y=+\infty} (x+y) f_{x,y}(x, y) dy dx$$

$$= |E(X+Y)| = |E(X)| + |E(Y)|$$

Let

$$y = z - x$$

$$dy = dz$$

$$z = x+y$$

②

(a)  $X, Y$  are jointly uniform on  $[0,1] \times [0,1]$ .

i.e

$$f_{X,Y}(x,y) = \begin{cases} 1, & 0 \leq x, y \leq 1 \\ 0, & \text{else} \end{cases}$$

Thus

$$f_X(x) = \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dy = \begin{cases} \int_0^1 1 dy = y \Big|_0^1 = 1, & 0 \leq x \leq 1 \\ 0, & \text{else} \end{cases}$$

$$f_Y(y) = \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dx = \begin{cases} \int_0^1 1 dx = x \Big|_0^1 = 1, & 0 \leq y \leq 1 \\ 0, & \text{else} \end{cases}$$

Thus,  $X \sim Y \sim \text{Uniform}[0,1]$ (b)  $X, Y$  have joint pdf

$$f_{X,Y}(x,y) = \begin{cases} 2x(1-y) + 2y(1-x), & 0 \leq x, y \leq 1 \\ 0, & \text{else} \end{cases}$$

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dy = \int_0^1 2x(1-y) + 2y(1-x) dy \\ &= 2x(y - \frac{1}{2}y^2) + (1-x)y^2 \Big|_0^1 \\ &= 2x\left(1 - \frac{1}{2}\right) + (1-x) - (0+0) \\ &= x + 1 - x \\ &= 1 \quad \text{when } x \in [0,1] \end{aligned}$$

$$\begin{aligned}
 f_Y(y) &= \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dx = \int_0^1 2x(1-y) + 2y(1-x) dx \\
 &= x(1-y) + 2y\left(x - \frac{1}{2}x^2\right) \quad \Big|_{x=0}^{x=1} \\
 &= 1-y + 2y\left(1 - \frac{1}{2}\right) - (0+0) \\
 &= 1-y+y \\
 &= 1 \quad \text{when } y \in [0,1]
 \end{aligned}$$

Moral:  $X, Y$  can be individually uniform on  $[0,1]$  but not jointly uniform on  $[0,1] \times [0,1]$

③ Sam breaks a stick of length 1 at a point  $X$  distributed uniformly over the interval  $(0, 1)$ . As a function of  $p \in (0, 1)$ , what's the expected length of the piece that contains the point  $p$ .

Know: - Uniform on  $(0, 1)$

So

$$f_X(x) = \begin{cases} 1, & x \in (0, 1) \\ 0, & \text{else} \end{cases}$$

$$\mathbb{E}(X) = \int_0^1 x \, dx = \frac{1}{2}$$

$\frac{b+a}{2}$   
expected location  
to break stick

Let  $Y$  be the length of the piece that contains the point  $p$ .

Then

$$Y = g(X) = \begin{cases} X, & X > p \\ 1-X, & X < p \end{cases}$$

So

$$\begin{aligned} \mathbb{E}(Y) &= \int_0^p (1-x) \, dx + \int_p^1 x \, dx \\ &= x - x^2 \Big|_0^p + \frac{1}{2} x^2 \Big|_p^1 \\ &= p(1-p) + \frac{1}{2} \end{aligned}$$

(4) Time it takes Maddy to get to work is a continuous rv  $X$  w/ CDF  $F_X(x)$ .

- If  $s$  minutes LATE to work, incurs cost  $\alpha s$ ,  $\alpha > 0$
- If  $s$  minutes EARLY to work, incurs cost  $\beta s$ ,  $\beta > 0$ .

(a) Time she should depart to MINIMIZE cost. Call this time  $t^*$ .

optimal solution  
satisfies

$$\rightarrow F_X(t^*) = \frac{\alpha}{\alpha + \beta}$$

let the time she arrives =  $t_a$ .

let the time work starts =  $t_s$ .

let the time she departs =  $t_d$ .

Put  $x = t_s - t_d$ .

$$s = t_a - t_s$$

If  $s < 0$ , cost is  $\alpha s$

If  $s > 0$ , cost is  $\beta s$

The cost is a function of commute time is thus

$$c(x) = \begin{cases} \alpha(x - t_a), & t_a \leq x \\ \beta(t_a - x), & t_a > x \end{cases} \quad \leftarrow x = t_s - t_d$$

Minimize cost.

$$IE(c(x)) = \int_0^\infty c(x) f_x(x) dx$$

$$= \int_0^{t_a} \beta(t-x) f_x(x) dx + \int_{t_a}^\infty \alpha(x-t) f_x(x) dx$$

$$= \int_0^{t_a} \beta t f_x(x) dx - \int_0^{t_a} \beta x f_x(x) dx + \int_{t_a}^\infty \alpha x f_x(x) dx - \int_{t_a}^\infty \alpha t f_x(x) dx$$

To minimize,

$$\frac{d}{dx} \left[ \int_0^t \beta t f_x(x) dx - \int_0^t \beta x f_x(x) dx + \int_t^\infty \alpha x f_x(x) dx - \int_t^\infty \alpha t f_x(x) dx \right] = 0$$

$$\cancel{\beta t f_x(t)} + \cancel{\beta F_x(t)} - \cancel{\beta t F_x(t)} + \alpha F_x(t) - \alpha = 0$$

$$(\alpha + \beta) F_x(t) = \alpha$$

$$F_x(t) = \frac{\alpha}{\alpha + \beta} \quad \leftarrow \text{call this } t^*$$

$$(b) f = \frac{\alpha}{\alpha + \beta}$$

(c)  $X \sim \text{Uniform}[0, t_{\max}]$

$$\mu = \frac{b+a}{2} = \frac{t_{\max}}{2}, \quad \sigma^2 = \frac{(b-a)^2}{12} = \frac{t_{\max}^2}{12} \Rightarrow \sigma = \frac{t_{\max}}{2\sqrt{3}}$$

(d)  $F_X(x) = \frac{x-a}{b-a} = \frac{x}{t_{\max}}$

$$F_X(t^*) = \frac{t^*}{t_{\max}} = \gamma$$

$$t^* = \gamma t_{\max} = \sqrt{12} \sigma = 2\mu\gamma$$

(e)  $t^*$  is linear in the mean of  $X$  and also in the standard deviation of  $X$ .

⑤ Input  $X$  to binary communication channel either  $+c$  or  $-c$  with probabilities  $\frac{1}{4}, \frac{3}{4}$  respectively. Here,  $c > 0$  is given.

Channel output is

$$Y = X + N$$

where  $N$  is  $\text{Gaussian}(0, 1)$

(a)  $f_{Y|A}(y)$  where  $A = \{X=x\}$  ?

$$Y = c + N \sim \text{Gaussian}(\mu=c, \sigma^2=1)$$

Thus

$$f_{Y|A}(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-c)^2}{2}}$$

and

$$f_{Y|A^c}(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{(y+c)^2}{2}} \quad \text{i.e. } Y = N - c \sim \text{Gaussian}(-c, 1)$$

(b)

$$\begin{aligned} f_Y(y) &= f_{Y|A}(y) P(A) + f_{Y|A^c} P(A^c) \\ &= \frac{1}{4\sqrt{2\pi}} e^{-\frac{(y-c)^2}{2}} + \frac{3}{4\sqrt{2\pi}} e^{-\frac{(y+c)^2}{2}} \end{aligned}$$

$$(c) P(\{X=c\} \mid \{Y>0\}) = \frac{P(\{Y>0\} \mid \{X=c\}) P(\{X=c\})}{P(\{Y>0\})}$$

$$\begin{aligned} P(\{Y>0\} \mid \{X=c\}) P(\{X=c\}) &= \frac{1}{4} \int_0^\infty \frac{1}{\sqrt{2\pi}} e^{-\frac{(y-c)^2}{2}} dy \\ &= \frac{1}{4} \int_{y=-c}^\infty \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt \\ &= \frac{1}{4} (1 - \Phi(-c)) \end{aligned}$$

$$P(\{Y>0\}) = \int_0^\infty \frac{1}{4\sqrt{2\pi}} e^{-\frac{(y-c)^2}{2}} + \frac{3}{4\sqrt{2\pi}} e^{-\frac{(y+c)^2}{2}} dy$$

$$\frac{1}{4} \int_{y=-c}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt + \frac{3}{4} \int_{y=+c}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$$

$$= \frac{1}{4} (1 - \Phi(-c)) + \frac{3}{4} (1 - \Phi(c))$$

Thus

$$P(\{X=c\} \mid \{Y>0\}) = \frac{\frac{1}{4} (1 - \Phi(-c))}{\frac{1}{4} (1 - \Phi(-c)) + \frac{3}{4} (1 - \Phi(c))}$$