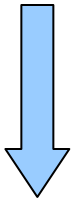




Multiple r.v.'s

- Several r.v.'s at a time
 - *Measuring different quantities simultaneously*
 - Engine oil pressure, RPM, generator voltage
 - *Repeated measurement of the same quantity*
 - Sampling a waveform, such as EEG, speech

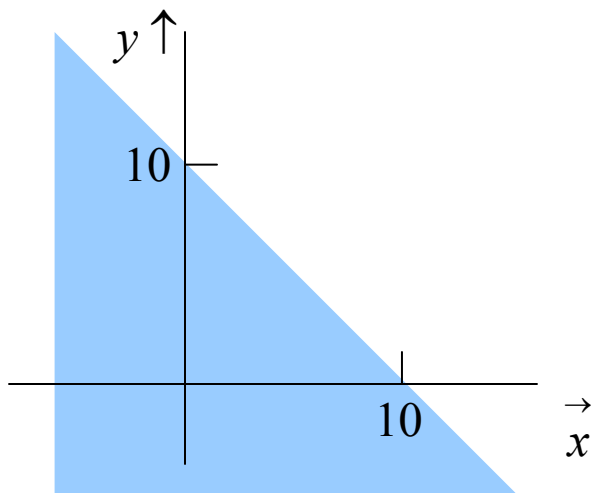


- Joint behavior of two or more r.v.'s
 - *Independence of sets of r.v.'s*
 - *Correlation if not independent*

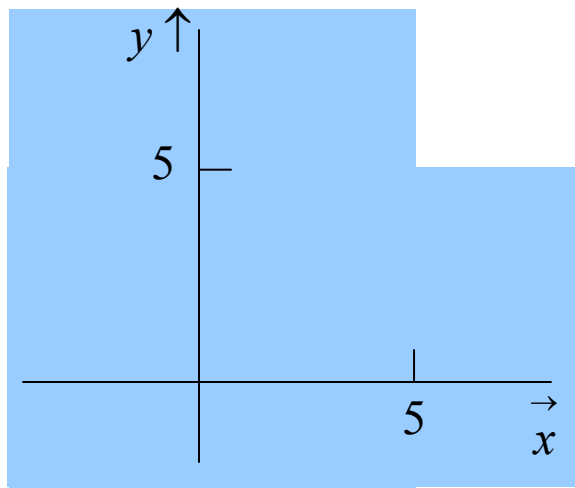
Events & probabilities

- Each event involving an n -dimensional r.v. $\mathbf{X}=(X_1, X_2, \dots, X_n)$ has a corresponding region in an n -dimensional real space
 - E.g. 2-D r.v. $\mathbf{X}=(X, Y)$*

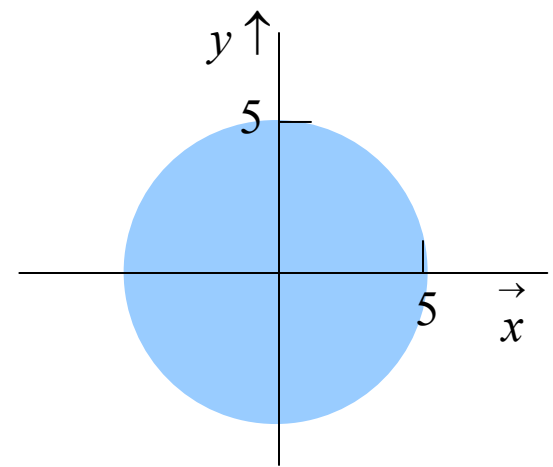
$$A = \{X + Y \leq 10\}$$



$$B = \{\min(X, Y) \leq 5\}$$



$$C = \{X^2 + Y^2 \leq 25\}$$





Independence

- Intuitively, if r.v.'s X and Y are “independent,” then events that involve only X should be independent of events that involve only Y . In other words, if A_1 is any event that involves X only and A_2 is any event that involves Y only, then

$$P[X \in A_1, Y \in A_2] = P[X \in A_1]P[Y \in A_2]$$

- In general, n r.v.'s are independent if

$$P[X_1 \in A_1, \dots, X_n \in A_n] = P[X_1 \in A_1] \cdots P[X_n \in A_n]$$

where A_k is an event that involves X_k only

if r.v.'s are independent, knowing the probabilities of the r.v.'s in isolation suffices to specify probabilities of joint events



Pairs of discrete r.v.'s

vector r.v. $\mathbf{X} = (X, Y)$

$$S = \left\{ (x_j, y_k), j = 1, 2, \dots, k = 1, 2, \dots \right\}$$

joint probability mass function:

$$p_{X,Y}(x_j, y_k) = P\left[\{X = x_j\} \cap \{Y = y_k\}\right]$$

$$\triangleq P\left[X = x_j, Y = y_k\right] \quad j = 1, 2, \dots \quad k = 1, 2, \dots$$

gives probability of occurrence of pairs (x_j, y_k)

For any event A :
$$P[\mathbf{X} \in A] = \sum_{(x_j, y_k) \in A} \sum p_{X,Y}(x_j, y_k)$$

$$\sum_{j=1}^{\infty} \sum_{k=1}^{\infty} p_{X,Y}(x_j, y_k) = P[S] = 1$$

Marginal PMF's

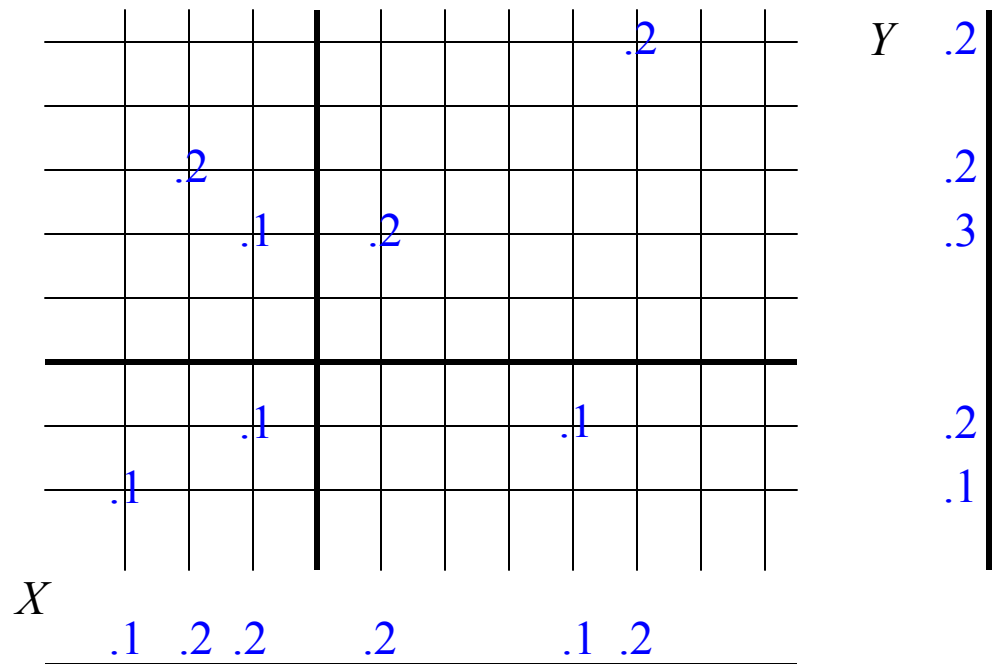
$$p_X(x_j) = p_{X,Y}(x_j, \text{any } y_k)$$

$$= P\left[\left\{\{X = x_j\} \cap \{Y = y_1\}\right\} \cup \left\{\{X = x_j\} \cap \{Y = y_2\}\right\} \cup \dots\right]$$

$$= \sum_{k=1}^{\infty} p_{X,Y}(x_j, y_k)$$

$$p_Y(y_k) = \sum_{j=1}^{\infty} p_{X,Y}(x_j, y_k)$$

**marginal PMF's are
insufficient – in general -
for specifying joint PMF**



Ex 4.6 tossing loaded dice

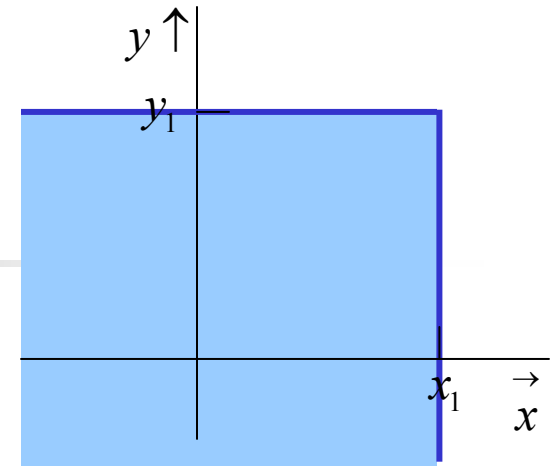
joint PMF in units of $\frac{1}{42}$

$j \backslash k$	1	2	3	4	5	6				
1	2	1	1	1	1	1	}	7	1	
2	1	2	1	1	1	1		7	1	
3	1	1	2	1	1	1		7	1	
4	1	1	1	2	1	1		7	1	
5	1	1	1	1	2	1		7	1	
6	1	1	1	1	1	2		7	1	
	}									
	7	7	7	7	7	7		/42	/6	
	1	1	1	1	1	1				

marginals give no clue

Joint CDF of X and Y

- Basic building block:



semi-infinite rectangle: $\{(x, y) : \{x \leq x_1\} \cap \{y \leq y_1\}\}$

joint cumulative distribution function of X and Y :

$$\begin{aligned} F_{X,Y}(x_1, y_1) &= P[\{x \leq x_1\} \cap \{y \leq y_1\}] && \text{product-form event} \\ &= P[X \leq x_1, Y \leq y_1] \end{aligned}$$

long-term proportion of times in which the outcome (x, y) falls in the blue rectangle, or the probability mass in it



properties of the joint CDF

i. $F_{X,Y}(x_1, y_1) \leq F_{X,Y}(x_2, y_2) \quad x_1 \leq x_2, y_1 \leq y_2$
non-decreasing in NE direction

ii. $F_{X,Y}(-\infty, y_1) = F_{X,Y}(x_2, -\infty) = 0$
neither X nor Y takes on values $< -\infty$

iii. $F_{X,Y}(\infty, \infty) = 1$
 X and Y can take on only values $< \infty$

iv. $F_X(x) = F_{X,Y}(x, \infty) = P[X \leq x, Y < \infty] = P[X \leq x]$

$$F_Y(y) = F_{X,Y}(\infty, y) = P[X < \infty, Y \leq y] = P[Y \leq y]$$

marginal cumulative distribution functions

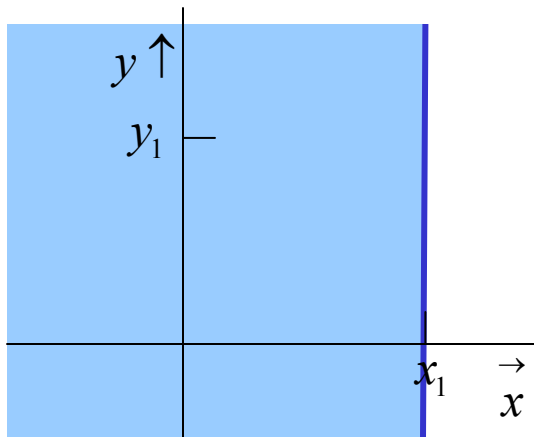
v. $\lim_{x \rightarrow a^+} F_{X,Y}(x, y) = F_{X,Y}(a, y)$

$$\lim_{y \rightarrow b^+} F_{X,Y}(x, y) = F_{X,Y}(x, b)$$

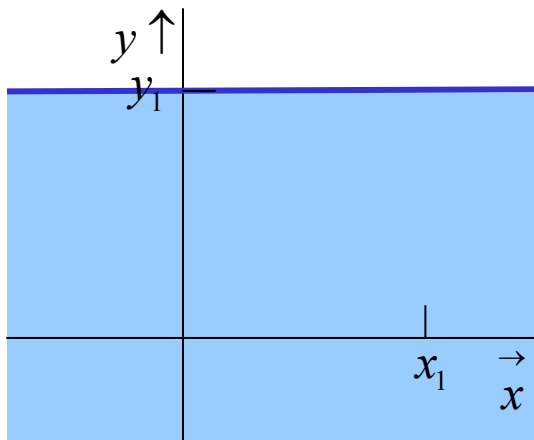
continuous from N & E

Ex 4.8 marginals for continuous joint r.v.'s

$$\text{Given } F_{X,Y}(x,y) = (1 - e^{-\alpha x})(1 - e^{-\beta y})u(x)u(y)$$



$$F_X(x) = \lim_{y \rightarrow \infty} F_{X,Y}(x,y) = (1 - e^{-\alpha x})u(x)$$



$$F_Y(y) = \lim_{x \rightarrow \infty} F_{X,Y}(x,y) = (1 - e^{-\beta y})u(y)$$

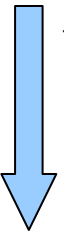
marginal CDFs are exponential distributions with parameters α and β respectively



Joint CDF in terms of joint PDF & v.v.

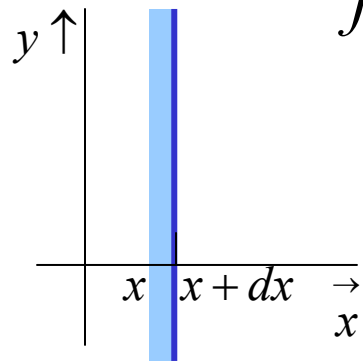
$$F_{X,Y}(x, y) = \int_{-\infty}^x \int_{-\infty}^y f_{X,Y}(x', y') dx' dy'$$

if jointly continuous r.v.'s


$$f_{X,Y}(x, y) = \frac{\partial^2 F_{X,Y}(x, y)}{\partial x \partial y}$$

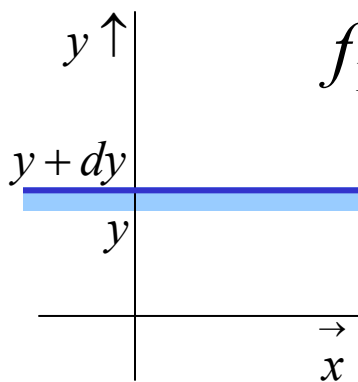
if the CDF is discontinuous
- or the partial derivatives are discontinuous -
then the joint PDF does not exist

Marginal PDF's from joint PDF



$$f_X(x) = \frac{d}{dx} F_{X,Y}(x, \infty) = \frac{d}{dx} \int_{-\infty}^x \left\{ \int_{-\infty}^{\infty} f_{X,Y}(x', y') dy' \right\} dx'$$

$$= \int_{-\infty}^{\infty} f_{X,Y}(x, y') dy' \quad \text{concentrating mass on x-axis}$$



$$f_Y(y) = \frac{d}{dy} F_{X,Y}(\infty, y) = \int_{-\infty}^{\infty} f_{X,Y}(x', y) dx'$$

concentrating mass on y-axis

marginal PDF's are obtained by integrating out the other variable(s)

$$f_{X,Y}(x,y) \equiv \begin{cases} 1 & (x,y) \in [0,1] \times [0,1] \\ 0 & \text{o.w.} \end{cases}$$

Ex 4.10 CDF for jointly uniform r.v.'s

$$F_{X,Y}(x,y) = \int_{-\infty}^x \int_{-\infty}^y f_{X,Y}(x',y') dx' dy'$$

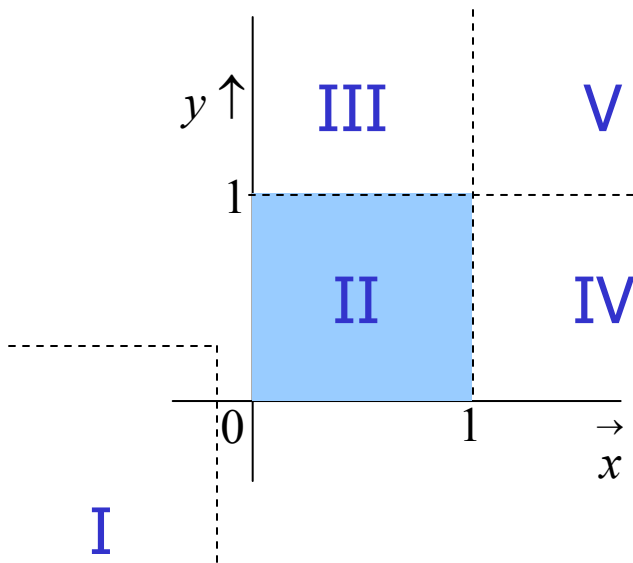
$$= 0 \quad \text{for } (x,y) \in I$$

$$= \int_0^x \int_0^y 1 dx' dy' = xy \quad \text{for } (x,y) \in II$$

$$= \int_0^x \int_0^1 1 dx' dy' = x \quad \text{for } (x,y) \in III$$

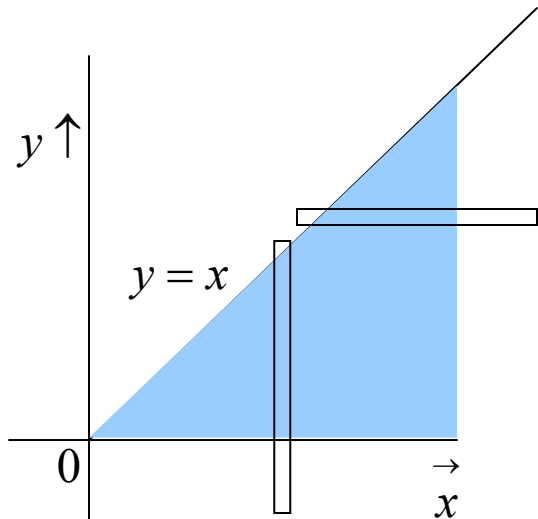
$$= \int_0^1 \int_0^y 1 dx' dy' = y \quad \text{for } (x,y) \in IV$$

$$= \int_0^1 \int_0^1 1 dx' dy' = 1 \quad \text{for } (x,y) \in V$$



Ex 4.11 finding marginal PDF's

$$f_{X,Y}(x,y) = \begin{cases} ce^{-x}e^{-y} & 0 \leq y \leq x < \infty \\ 0 & \text{o.w.} \end{cases}$$



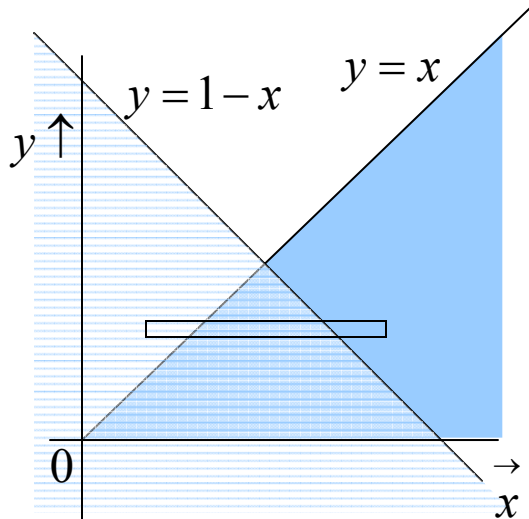
$$\begin{aligned} 1 &= \int_0^{\infty} \int_0^x ce^{-x}e^{-y} dy dx = \int_0^{\infty} ce^{-x} \int_0^x e^{-y} dy dx \\ &= \int_0^{\infty} ce^{-x} \frac{e^{-x} - 1}{-1} dx = \frac{c}{2} \Rightarrow c = 2 \end{aligned}$$

$$f_X(x) = \int_0^x 2e^{-x}e^{-y} dy = 2e^{-x}(1 - e^{-x}) \quad 0 \leq x < \infty$$

$$f_Y(y) = \int_y^{\infty} 2e^{-x}e^{-y} dx = 2e^{-y}(e^{-y}) = 2e^{-2y} \quad 0 \leq y < \infty$$

are these PDF's?

Ex 4.12 Find $P[X + Y \leq 1]$ in EX 4.11



$$\begin{aligned} P[X + Y \leq 1] &= \int_0^{0.5} \int_y^{1-y} 2e^{-x} e^{-y} dx dy \\ &= \int_0^{0.5} 2e^{-y} \int_y^{1-y} e^{-x} dx dy \\ &= \int_0^{0.5} 2e^{-y} \left[e^{-y} - e^{-(1-y)} \right] dy \\ &= -e^{-1} + 1 - e^{-1} = 1 - 2e^{-1} \end{aligned}$$

Ex 4.13 Find marginal PDFs of jointly Gaussian PDF

$$f_{X,Y}(x,y) = \frac{1}{2\pi\sqrt{1-\rho^2}} e^{-\frac{(x^2 - 2\rho xy + y^2)}{2(1-\rho^2)}}$$

$$f_X(x) = \frac{e^{-\frac{x^2}{2(1-\rho^2)}}}{2\pi\sqrt{1-\rho^2}} \int_{-\infty}^{\infty} e^{-\frac{(-2\rho xy + y^2)}{2(1-\rho^2)}} dy$$

completing the square

$$= \frac{e^{-\frac{x^2(1-\rho^2)}{2(1-\rho^2)}}}{2\pi\sqrt{1-\rho^2}} \int_{-\infty}^{\infty} e^{-\frac{(\rho^2 x^2 - 2\rho xy + y^2)}{2(1-\rho^2)}} dy$$

$$= \frac{e^{-x^2/2}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \frac{1}{\sqrt{1-\rho^2} \sqrt{2\pi}} e^{-\frac{(y-\rho x)^2}{2(1-\rho^2)}} dy = \frac{e^{-x^2/2}}{\sqrt{2\pi}} = N(0,1)$$

it's a PDF

symmetry

$$f_Y(y) = f_X(x)$$



Independence of two r.v.'s

discrete r.v.'s X and Y are independent

⇔ iff

joint PMF is product of marginal PMFs

$$\forall x_j, y_k$$

Ex 4.15 another look at Ex 4.6 loaded dice

joint PMF in units of $\frac{1}{42}$

$j \backslash k$	1	2	3	4	5	6	
1	2	1	1	1	1	1	}
2	1	2	1	1	1	1	
3	1	1	2	1	1	1	
4	1	1	1	2	1	1	
5	1	1	1	1	2	1	
6	1	1	1	1	1	2	
	1	1	1	1	1	1	/6

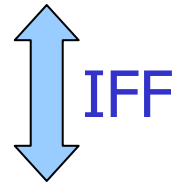
marginals give no clue

not independent

$$p_{X,Y}(x_j, y_k) \neq p_X(x_j) p_Y(y_k) \quad \forall x_j, y_k$$



Independence of two r.v.'s in general



$$p_{X,Y}(x_j, y_k) = p_X(x_j) p_Y(y_k) \quad \forall x_j, y_k \quad \text{discrete}$$

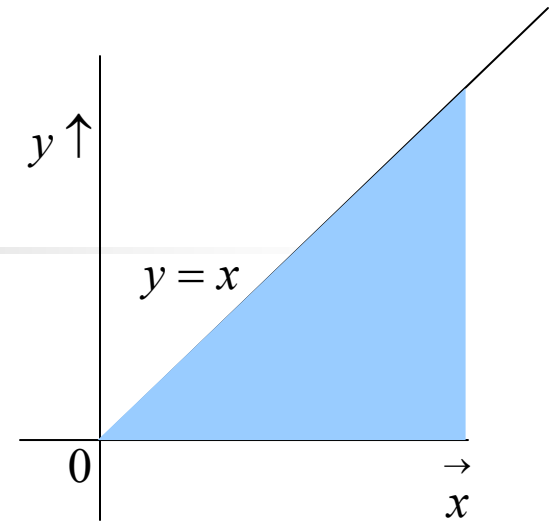
$$F_{X,Y}(x, y) = F_X(x) F_Y(y) \quad \forall x, y$$

$$f_{X,Y}(x, y) = f_X(x) f_Y(y) \quad \forall x, y \quad \text{jointly continuous}$$

Ex 4.17 back to Ex 4.11

joint

$$f_{X,Y}(x,y) = \begin{cases} 2e^{-x}e^{-y} & 0 \leq y \leq x < \infty \\ 0 & \text{o.w.} \end{cases}$$



$$f_X(x) = \int_0^x 2e^{-x}e^{-y} dy = 2e^{-x}(1 - e^{-x}) \quad 0 \leq x < \infty$$

marginal PDFs

$$f_Y(y) = \int_y^\infty 2e^{-x}e^{-y} dx = 2e^{-y}(e^{-y}) = 2e^{-2y} \quad 0 \leq y < \infty$$

product? $2e^{-x}e^{-y} \neq 2e^{-x}(1 - e^{-x})2e^{-2y}$

X and Y are not independent



Ex 4.18 back to Ex 4.13

$$\left. \begin{aligned} f_{X,Y}(x,y) &= \frac{1}{2\pi\sqrt{1-\rho^2}} e^{-\frac{(x^2-2\rho xy+y^2)}{2(1-\rho^2)}} \\ f_X(x) &= \frac{e^{-x^2/2}}{\sqrt{2\pi}} & f_Y(y) &= \frac{e^{-y^2/2}}{\sqrt{2\pi}} \\ f_X(x)f_Y(y) &= \frac{e^{-x^2/2}}{\sqrt{2\pi}} \frac{e^{-y^2/2}}{\sqrt{2\pi}} = \frac{e^{-\frac{(x^2+y^2)}{2}}}{2\pi} \end{aligned} \right\} \equiv \text{iff } \rho = 0$$

Conditional probability

$$P[Y \in A | X = x] = \frac{P[Y \in A, X = x]}{P[X = x]}$$

X discrete

$$F_Y(y | x_k) = \frac{P[Y \leq y, X = x_k]}{P[X = x_k]} \text{ for } P[X = x_k] > 0$$

if derivative exists

$$f_Y(y | x_k) = \frac{d}{dy} F_Y(y | x_k)$$

independence

$$F_Y(y | x) = F_Y(y)$$

$$f_Y(y | x) = f_Y(y)$$

$$P[Y \in A | X = x_k] = \int_{y \in A} f_Y(y | x_k) dy$$

Conditional probability

$$P[Y \in A | X = x] = \frac{P[Y \in A, X = x]}{P[X = x]}$$

X and Y discrete

$$p_Y(y_j | x_k) = P[Y = y_j | X = x_k] = \frac{P[X = x_k, Y = y_j]}{P[X = x_k]}$$

δ functions
 \sim PMF weights

$$= \begin{cases} \frac{p_{X,Y}(x_k, y_j)}{p_X(x_k)} & \text{for } P[X = x_k] = p_X(x_k) > 0 \\ 0 & \text{for } P[X = x_k] = p_X(x_k) = 0 \end{cases}$$

in general

$$P[Y \in A | X = x_k] = \sum_{y_j \in A} p_Y(y_j | x_k) = \sum_{y_j \in A} p_Y(y_j)$$

if also independent



Ex 4.20 back to Ex 4.14

- X is the input and Y the output of a communication channel. $P[Y < 0 | X = +1]$?

$$f_Y(y|1) \sim U[-1, 3] \implies P[Y < 0 | X = +1] = \int_{-1}^0 \frac{1}{4} dy = \frac{1}{4}$$

integrate over event



Conditional CDF of Y given $X=x$

$$F_Y(y | x) \triangleq \lim_{h \rightarrow 0} F_Y(y | x < X \leq x + h)$$

$$= \frac{P[Y \leq y, x < X \leq x + h]}{P[x < X \leq x + h]} = \frac{\int_{-\infty}^y \int_x^{x+h} f_{X,Y}(x', y') dx' dy'}{\int_x^{x+h} f_X(x') dx'}$$

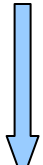
$$\simeq \frac{h \int_{-\infty}^y f_{X,Y}(x, y') dy'}{h f_X(x)} = \frac{\int_{-\infty}^y f_{X,Y}(x, y') dy'}{f_X(x)}$$

$$= \int_{-\infty}^y f_Y(y') dy' = F_Y(y) \quad \text{if independent}$$



Conditional PDF of Y given $X=x$


$$F_Y(y|x) = \frac{\int_{-\infty}^y f_{X,Y}(x, y') dy'}{f_X(x)}$$


$$f_Y(y|x) = \frac{d}{dy} F_Y(y|x) = \frac{f_{X,Y}(x, y)}{f_X(x)}$$

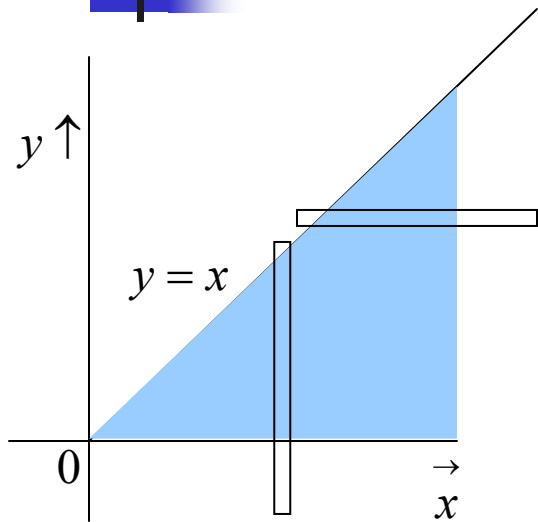
almost Bayes' rule

$$f_Y(y|x) dy = \frac{f_{X,Y}(x, y) dx dy}{f_X(x) dx}$$

if independent


$$f_Y(y|x) = f_Y(y)$$

Ex 4.21 back to Ex 4.11



$$f_{X,Y}(x,y) = \begin{cases} ce^{-x}e^{-y} & 0 \leq y \leq x < \infty \\ 0 & \text{o.w.} \end{cases}$$

$$f_X(x) = 2e^{-x}(1 - e^{-x}) \quad 0 \leq x < \infty$$

$$f_Y(y) = 2e^{-2y} \quad 0 \leq y < \infty$$

$$f_X(x|y) = \frac{f_{X,Y}(x,y)}{f_Y(y)} = \frac{2e^{-x}e^{-y}}{2e^{-2y}} = e^{-(x-y)}; y \leq x < \infty$$

$$f_Y(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)} = \frac{2e^{-x}e^{-y}}{2e^{-x}(1 - e^{-x})} = \frac{e^{-y}}{1 - e^{-x}}; 0 \leq y < x$$