

Theorem:  $\text{Cov}(aX+b, Y) = a \text{Cov}(X, Y)$

Recall:  $\text{Var}(aX+b) = a^2 \text{Var}(X)$

Pf.

$$\begin{aligned}
 \text{Cov}(aX+b, Y) &= E[(aX+b) - E[aX+b]](Y - EY)] \\
 &= E[(aX+b - aEX - b)(Y - EY)] \\
 &= E[a(X - EX)(Y - EY)] \\
 &= a E[(X - EX)(Y - EY)] \\
 &= a \text{Cov}(X, Y)
 \end{aligned}$$

Corollaries: ①  $\text{Cov}(X, cY+d) = c \text{Cov}(X, Y)$

②  $\text{Cov}(aX+b, cY+d) = ac \text{Cov}(X, Y).$

Theorem:

$$\underline{\text{Cor}(aX+b, cY+d) = \text{Sign}(a)\text{Sign}(c)\text{Cor}(X, Y)}$$

$$\text{Sign}(x) = \begin{cases} 1 & x > 0 \\ 0 & x = 0 \\ -1 & x < 0 \end{cases}$$

Ex.  $\text{Cor}(-5X, Y) = -\text{Cor}(X, Y)$

pf.  $a, c \neq 0$

if  $x \neq 0$  then  $\text{sign}(x) = \frac{x}{|x|}$

$$\text{Cor}(aX+b, cY+d) = \frac{\text{Cov}(aX+b, cY+d)}{\sqrt{\text{Var}(aX+b) \text{Var}(cY+d)}}$$

$$= \frac{ac \text{Cov}(X, Y)}{\sqrt{a^2 \text{Var}(X) c^2 \text{Var}(Y)}}$$

$$= \frac{a}{|a|} \frac{c}{|c|} \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}}$$

$$= \text{Sign}(a) \text{Sign}(c) \text{Cor}(X, Y).$$

Claim:

$$-1 \leq \text{Cor}(X, Y) \leq 1$$

Pf-  $\tilde{X} = \frac{X - EX}{\sqrt{\text{Var}(X)}} = \underbrace{\frac{1}{\sqrt{\text{Var}(X)}}}_{a > 0} X - \underbrace{\frac{EX}{\sqrt{\text{Var}(X)}}}_b = aX + b$

$$E[\tilde{X}] = \frac{EX - EX}{\sqrt{\text{Var}(X)}} = 0$$

$$\text{Var}(\tilde{X}) = \left( \frac{1}{\sqrt{\text{Var}(X)}} \right)^2 \text{Var}(X) = 1$$

$$\tilde{Y} = \frac{Y - EY}{\sqrt{\text{Var}(Y)}}$$

Notice:  $\text{Cor}(\tilde{X}, \tilde{Y}) = \text{Cor}(X, Y) \quad (*)$

Also:  $\text{Cor}(\tilde{X}, \tilde{Y}) = \frac{\text{Cov}(\tilde{X}, \tilde{Y})}{\sqrt{\underbrace{\text{Var}(\tilde{X})}_1 \underbrace{\text{Var}(\tilde{Y})}_1}} = \frac{\text{Cov}(\tilde{X}, \tilde{Y})}{1}$

Chain:  $\text{Cor}(X, Y) = \text{Cor}(\tilde{X}, \tilde{Y}) = \text{Cov}(\tilde{X}, \tilde{Y})$

Consider

$$\text{Var}(\tilde{X} \pm \tilde{Y}) = \text{Var}(\tilde{X}) + \text{Var}(\tilde{Y}) \pm 2\text{Cov}(\tilde{X}, \tilde{Y})$$

Consider

$$\begin{aligned}\underline{\text{Var}(\tilde{X} \pm \tilde{Y})} &= \underbrace{\text{Var}(\tilde{X})}_1 + \underbrace{\text{Var}(\tilde{Y})}_1 \pm 2 \underbrace{\text{Cov}(\tilde{X}, \tilde{Y})}_{\text{Cor}(\tilde{X}, \tilde{Y})} \\ &= 2 \pm 2 \text{Cor}(\tilde{X}, \tilde{Y}) \geq 0\end{aligned}$$

So  $1 \pm \text{Cor}(\tilde{X}, \tilde{Y}) \geq 0$

$1 + \text{Cor}(\tilde{X}, \tilde{Y}) \geq 0$

↓

$$\boxed{-1 \leq \text{Cor}(\tilde{X}, \tilde{Y})}$$

$1 - \text{Cor}(\tilde{X}, \tilde{Y}) \geq 0$

↓

$$\boxed{\text{Cor}(\tilde{X}, \tilde{Y}) \leq 1}$$

Theorem: Short-Cut formula for Covariance

$$\text{Cov}(X, Y) = E[XY] - E[X]E[Y]$$

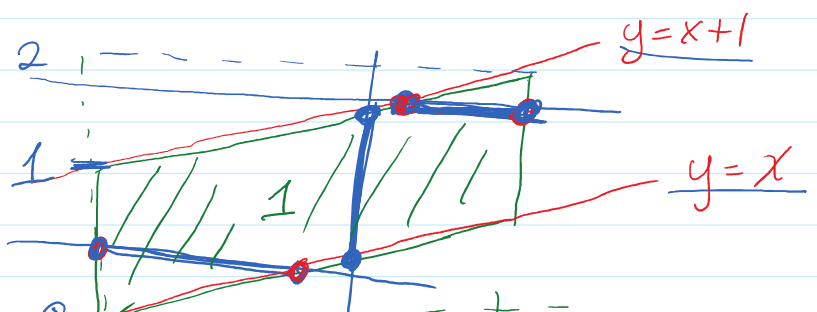
Analogy:  $\text{Var}(X) = E[X^2] - (EX)^2$

Ex.  $f(x, y) = 1$  for  $0 < x < 1$   
 $x < y < x+1$

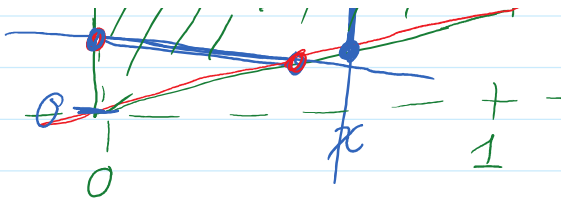
We had calculated

$$\boxed{E[XY] = 7/12}$$

What is Cov/Cor?



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Marginal of X

$$f_X(x) = \int_{\mathbb{R}} f(x,y) dy = \int_{y=x}^{y=x+1} 1 dy = (x+1) - x = 1$$

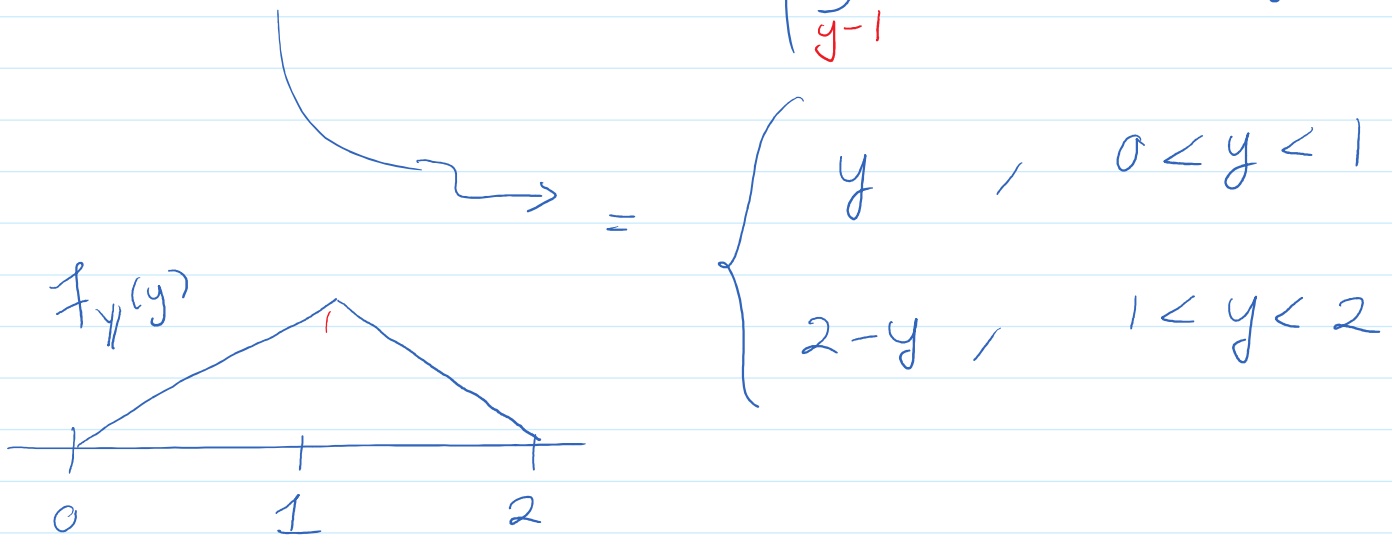
So  $f_X(x) = 1$  for  $0 < x < 1$

i.e.  $X \sim U(0,1)$

So  $E[X] = 1/2$  and  $Var(X) = 1/12$

Marginal of Y

$$f_Y(y) = \int_{\mathbb{R}} f(x,y) dx = \begin{cases} \int_0^y 1 dx & 0 < y < 1 \\ \int_{y-1}^1 1 dx & 1 < y < 2 \end{cases}$$



0

1

2

$$E[Y] = 1 \quad \text{and} \quad \text{Var}(Y) = \frac{1}{6}$$

So

$$\begin{aligned} \text{Cov}(X, Y) &= E[XY] - E[X]E[Y] \\ &= \frac{7}{12} - \left(\frac{1}{2}\right)(1) = \frac{1}{12} \end{aligned}$$

$$\text{Cor}(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}} = \frac{\frac{1}{12}}{\sqrt{\frac{1}{12} \cdot \frac{1}{6}}}$$

### Conditional Probability

$$P(A|B) = \frac{P(AB)}{P(B)}$$

If  $X$  and  $Y$  are discrete. Consider

$$A = \{X=x\} \quad \text{and} \quad B = \{Y=y\}$$

$$P(X=x|Y=y) = \frac{P(AB)}{P(B)} = \frac{P(X=x, Y=y)}{P(Y=y)}$$

$$P(X=x|Y=y)$$

$$= \frac{f(x,y)}{f_Y(y)}$$

Defn: Conditional PMF

If  $X$  and  $Y$  are discrete then the conditional PMF of  $X$  given  $Y=y$  is

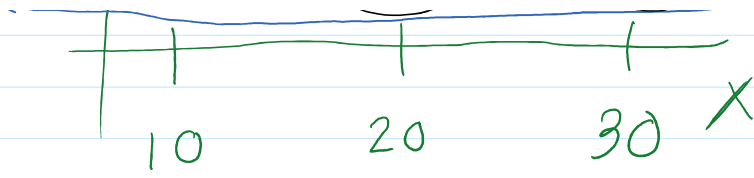
$$f(x|y) = f_{X|Y=y}(x) = \frac{f(x,y)}{f_Y(y)}$$

Basically think of  $Z = "X|Y=y"$  as a univariate RV.

Ex. Joint PMF  $\rightarrow f(x,y)$

$Y$	0	0	$4/18$
2	0	0	
1	$3/18$	$4/18$	$3/18$
0	$2/18$	$2/18$	0
	10	20	30
			$X$

$$f_Y(0) = 4/18$$



Let's get dist of  $X|Y=0$

$$f(x|0) = \frac{f(x,0)}{f_Y(0)} = \begin{cases} \frac{2/18}{4/18} = \frac{1}{2} & x=10 \\ \frac{2/18}{4/18} = \frac{1}{2} & x=20 \\ 0 & x=30 \end{cases}$$

$X|Y=0$  equally split b/w 10, 20

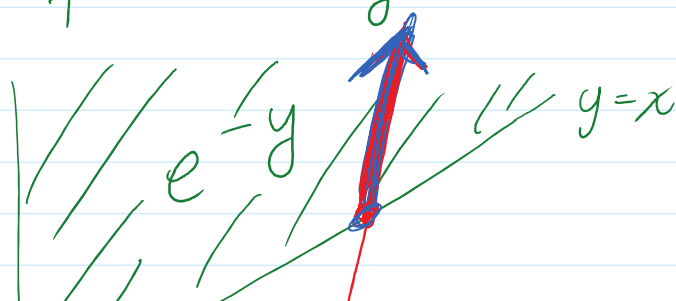
Defn: Conditional PDF

If  $X$  and  $Y$  are continuous then the conditional PDF of  $X$  given  $Y=y$  is

$$f(x|y) = \frac{f(x,y)}{f_Y(y)}$$

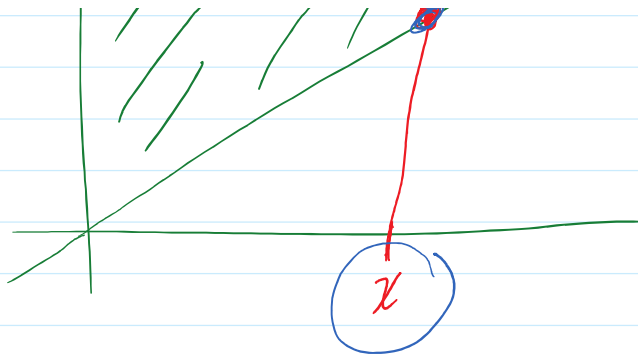
Ex.  $f(x,y) = e^{-y}$  for  $0 < x < y$

What is the PDF of  $Y|X=x$ ?





of  $\forall x = x$  :



Marginal of  $X$

$$f_X(x) = \int_{\mathbb{R}} f(x, y) dy = \int_{y=x}^{\infty} e^{-y} dy = -e^{-y} \Big|_x^{\infty}$$

$$\begin{aligned} &= 0 - (-e^{-x}) \\ &= e^{-x} \end{aligned} \quad \text{for } x > 0$$

$$X \sim \text{Exp}(\lambda = 1)$$

So

$$f(y|x) = \frac{f(x, y)}{f_X(x)} = \frac{e^{-y}}{e^{-x}} \quad \text{for } x < y$$

$$= e^{-(y-x)} \quad \text{for } y > x$$

called shifted Exponential dist

$\backslash \text{Exp}(1)$

$\backslash \text{Shifted Exp}$

