

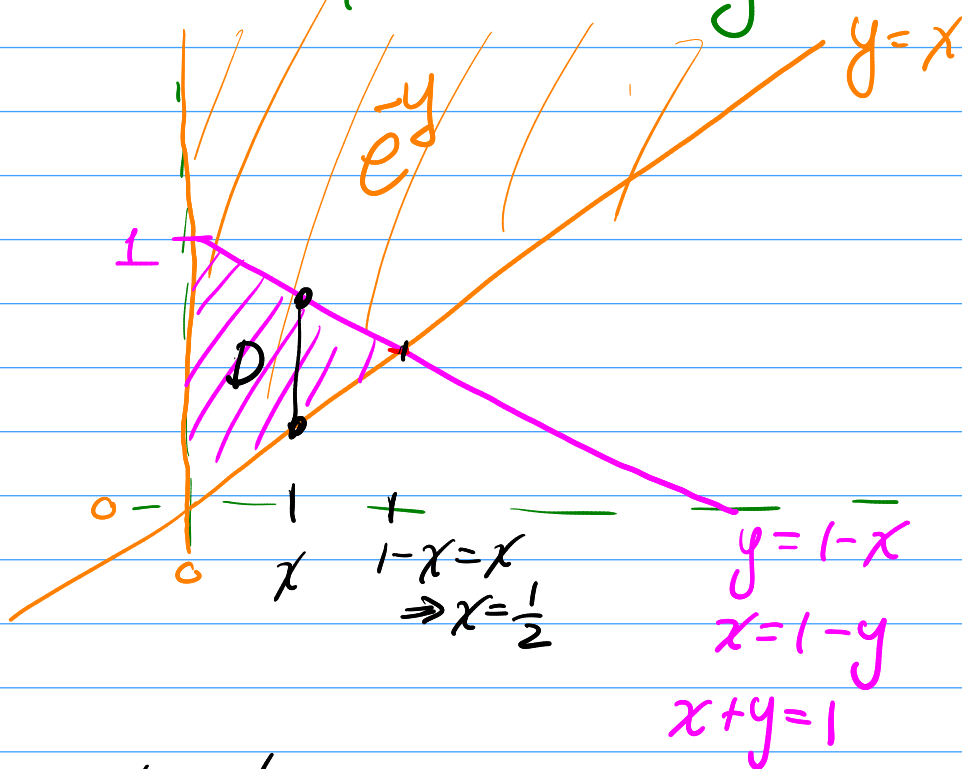
Lecture 17

Ex.

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

$$P(X+Y \geq 1)$$

$$= \iint_C f(x, y) dx dy$$



$$= 1 - \underbrace{P(X+Y < 1)}$$

$$= \iint_D f(x, y) dx dy = \int_0^{\frac{1}{2}} \int_x^{1-x} e^{-y} dy dx$$

$$= \int_0^{\frac{1}{2}} \left[-e^{-y} \right]_x^{1-x} dx$$

$$= \int_0^{\frac{1}{2}} -e^{-(1-x)} + e^{-x} dx$$

$$= \int_0^{\frac{1}{2}} e^{-x} - \frac{1}{e} e^x dx$$

$$= \left[-e^{-x} - \frac{1}{e} e^x \right]_0^{\frac{1}{2}}$$

$$= -e^{-\frac{1}{2}} - \frac{1}{e} e^{\frac{1}{2}} - \left(-1 - \frac{1}{e} \right)$$

Final ans: 

$$P(X+Y \geq 1) = 1 -$$

Defn: Bivariate Expectation

If (X, Y) is a bivariate RV and

$$g: \mathbb{R}^2 \rightarrow \mathbb{R}$$

then

$$E[g(X, Y)] = \begin{cases} \sum_x \sum_y g(x, y) f(x, y) & \text{(discrete)} \\ \iint g(x, y) f(x, y) dx dy & \text{(cts)} \end{cases}$$

$$\left[\underline{\text{Uni!}} \quad E[g(X)] = \int g(x) f(x) dx \right]$$

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

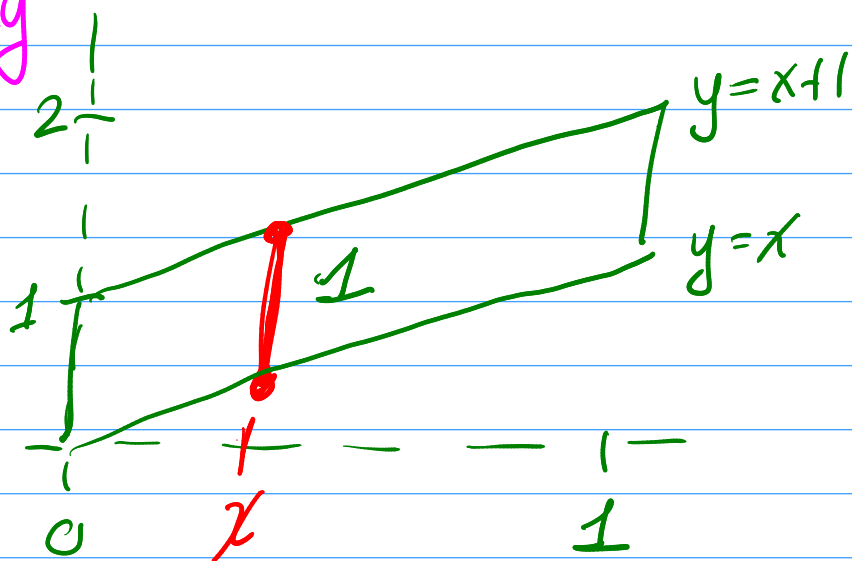
$$E[XY] = \iint xy f(x,y) dx dy$$

$$= \int_0^1 \int_x^{x+1} xy (1) dy dx$$

$$= \int_0^1 x \left(\int_x^{x+1} y dy \right) dx$$

$$= \int_0^1 x \left[\frac{y^2}{2} \right]_x^{x+1} dx = \int_0^1 x \left(\frac{(x+1)^2 - x^2}{2} \right) dx$$

$$= \dots = 7/12$$



Theorem! Biv. Exp. is Linear.

If $g_1: \mathbb{R}^2 \rightarrow \mathbb{R}$, $g_2: \mathbb{R}^2 \rightarrow \mathbb{R}$ and $a, b \in \mathbb{R}$
 then

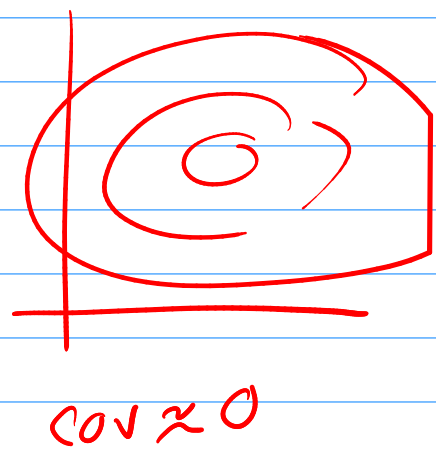
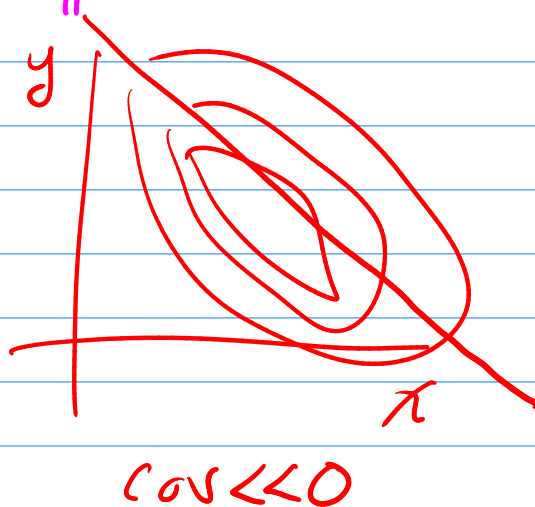
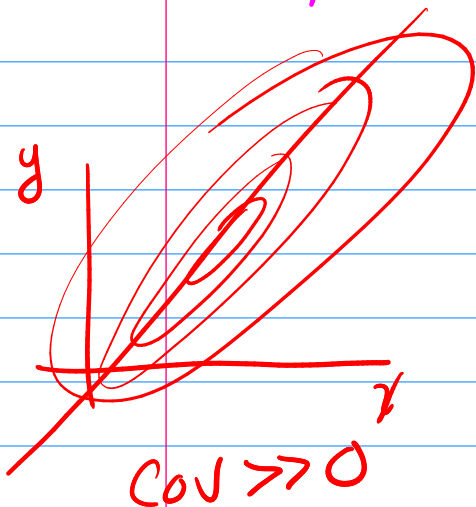
$$E[ag_1(x,y) + bg_2(x,y)] = a E[g_1(x,y)] + b E[g_2(x,y)]$$

Defn: Covariance

The covariance between X and Y is

$$\begin{aligned}\text{Cov}(X, Y) &= E[(X - EX)(Y - EY)] \\ &= E[(X - \mu_X)(Y - \mu_Y)]\end{aligned}$$

Claim: Cov measures how linearly related X and Y are.



Note:

① $\text{Var}(X) = E[(X - EX)^2]$
 $= \text{Cov}(X, X)$

② Covariance is scale sensitive:
 $\text{Cov}(5X, Y) = 5 \text{Cov}(X, Y)$

Defn: Correlation

Basically cov re-scaled to btwn
-1 and 1.

$$\begin{aligned}\text{Cor}(X, Y) &= \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}} \\ &= \frac{\text{Cov}(X, Y)}{\text{sd}(X) \text{sd}(Y)}.\end{aligned}$$

idea: $\text{Cor} \approx 1$, strong lin. rel.

$\text{Cor} \approx -1$, strong (neg.) lin. rel.

$\text{Cor} \approx 0$, no lin. rel.

Theorem: $a, b \in \mathbb{R}$,

$$\begin{aligned}\text{Var}(aX + bY) &= a^2 \text{Var}(X) + b^2 \text{Var}(Y) \\ &\quad + 2ab \text{Cov}(X, Y).\end{aligned}$$

pf: $Z = aX + bY$

$$\text{Var}(Z) = E[(Z - EZ)^2]$$

$$= E[(aX + bY - \underbrace{E[aX + bY]}_{\downarrow})^2]$$

$$= E[(aX + bY - aE[X] - bE[Y])^2]$$

$$= E[(\underbrace{a(X - EX)}_{\alpha} + \underbrace{b(Y - EY)}_{\beta})^2]$$

↓

$$(\alpha + \beta)^2 = \alpha^2 + \beta^2 + 2\alpha\beta$$

$$= E[a^2(X - EX)^2 + b^2(Y - EY)^2 + 2a(X - EX)b(Y - EY)]$$

$$= a^2 E[(X - EX)^2] + b^2 E[(Y - EY)^2] + 2ab E[(X - EX)(Y - EY)]$$

$$\underbrace{a^2 E[(X - EX)^2]}_{\text{Var}(X)} + 2ab E[(X - EX)(Y - EY)]$$

$$\underbrace{\hspace{10em}}_{\text{Cov}(X, Y)}.$$

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

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

pf $\text{Cov}(aX+b, Y)$

$$= E[(aX+b - \underline{E[aX+b]}) (Y - EY)]$$

$$= E[(\cancel{aX} + \cancel{b} - aE\cancel{X} - \cancel{b})(Y - EY)]$$

$$= E[a(X - EX)(Y - EY)]$$

$$= a E[(X - EX)(Y - EY)]$$

$$= a \text{Cov}(X, Y).$$

Corollaries:

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

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

Theorem:

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

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

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

pf. $a, c \neq 0$

note: $x \neq 0$ then $\text{sgn}(x) = \frac{x}{|x|}$

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

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

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

↓ |

$$\frac{a}{|a|} \quad \downarrow \quad \text{sgn}(c) \quad \text{Cor}(X, Y)$$

$\text{sgn}(a)$

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

WLOG: assume $\text{Var}(X) = \text{Var}(Y) = 1$
 $E[X] = E[Y] = 0$

Note: $\text{Cor}(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}} = \text{Cov}(X, Y)$

Consider

$$\text{Var}(X \pm Y) = \overset{1}{\text{Var}(X)} + \overset{1}{\text{Var}(Y)} \pm 2\text{Cov}(X, Y) \geq 0$$

So $2 \pm 2\text{Cov}(X, Y) \geq 0$

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

$$1 + \text{Cor} \geq 0$$

$$\Rightarrow \boxed{\text{Cor} \geq -1}$$

$$1 - \text{Cor} \geq 0$$

$$\Rightarrow \boxed{\text{Cor} \leq 1}$$

Theorem: Short-cut Covariance

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

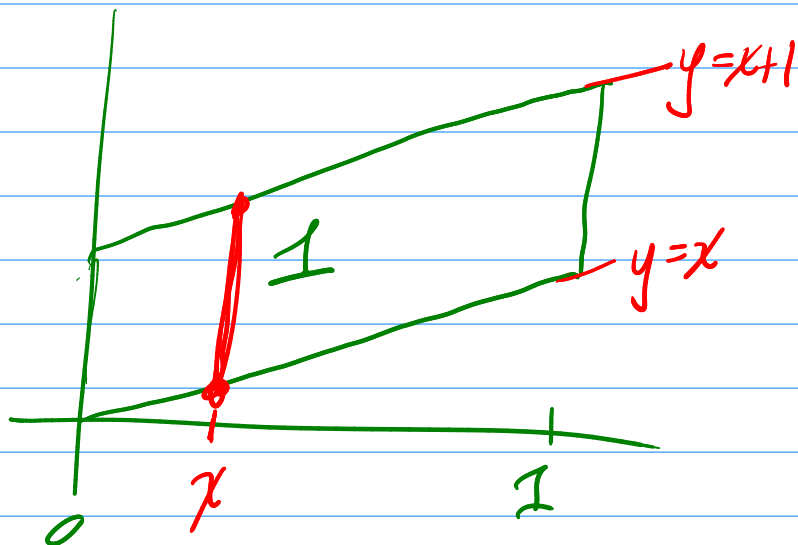
$$\left(\text{Var: } \text{Var}(X) = E[X^2] - E[X]^2 \right)$$

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

Had calc:

$$E[XY] = 7/12$$

What's the
Cov / Cor.



Marginal of X

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

$$\text{So } f_X(x) = 1 \text{ for } 0 < x < 1$$

$$X \sim U(0,1)$$

$$E[X] = \frac{1}{2}, \quad \text{Var}(X) = \frac{1}{12}.$$