

Lecture 17

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

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

$$\downarrow x+y=1 \Rightarrow y=1-x$$

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

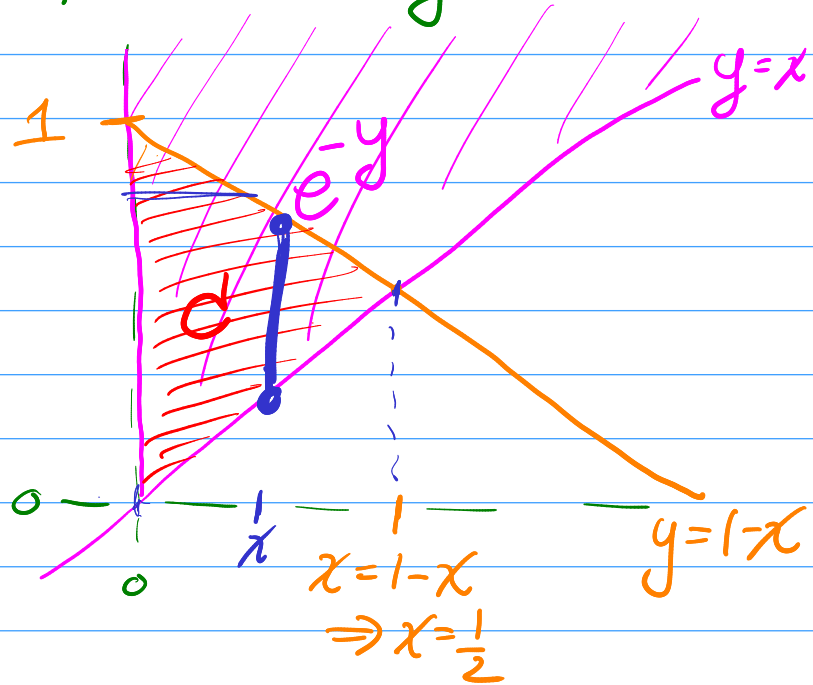
$$= \int_0^{\frac{1}{2}} \int_x^{1-x} e^{-y} dy dx$$

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

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

$$= \int_0^{\frac{1}{2}} -\frac{1}{e} e^x + e^{-x} dx = \left[-\frac{1}{e} e^x - e^{-x} \right]_0^{\frac{1}{2}}$$

$$= -\frac{1}{e} e^{\frac{1}{2}} - e^{-\frac{1}{2}} + \frac{1}{e} + 1$$



Defn: Bivariate Expectation

If (X, Y) is a Biv. RV and

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

then

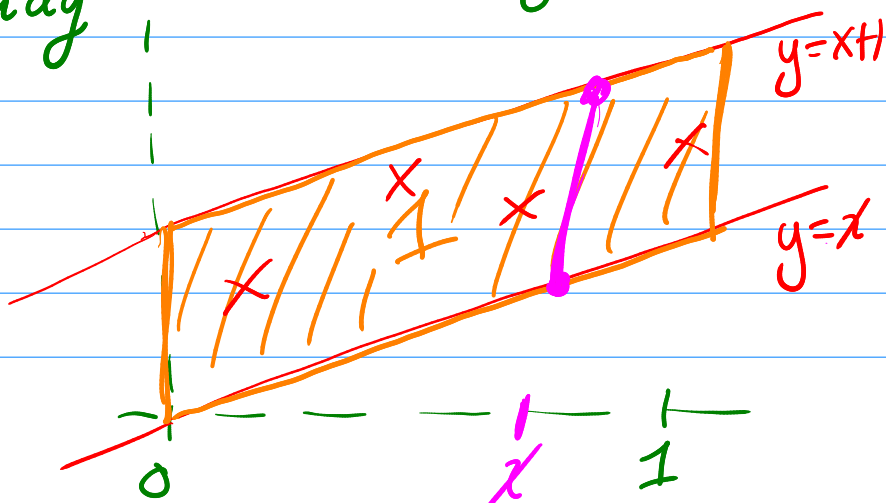
$$E[g(X, Y)] = \begin{cases} \sum_x \sum_y g(x, y) f(x, y) & \text{(discrete)} \\ \iint g(x, y) f(x, y) dx dy & \text{(cts)} \end{cases}$$

$$[\text{uni: } E[g(x)] = \int g(x) f(x) dx]$$

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

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

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



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

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

$$= \dots = 7/12$$

Theorem: Biv. Exp. is Linear

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

$$E[ag_1(X, Y) + bg_2(X, Y)]$$

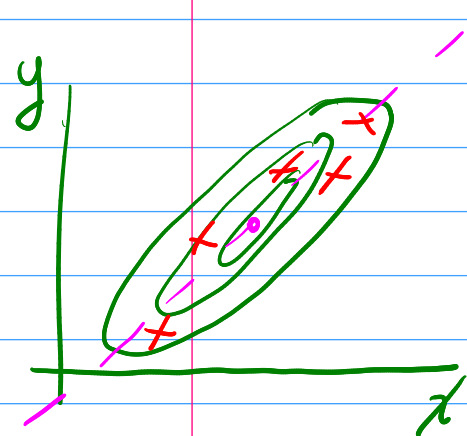
$$= aE[g_1(X, Y)] + bE[g_2(X, Y)].$$

Defn: Covariance

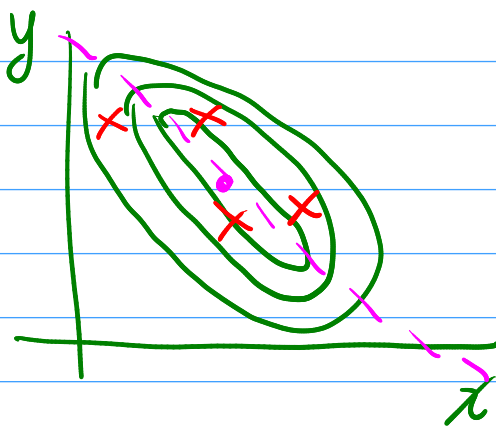
The covariance between X and Y is

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

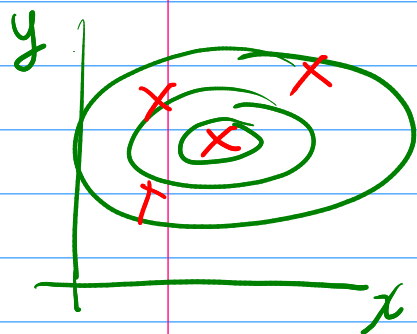
Claim: Cov measures strength/direction
of linear rels. (btwn X and Y)



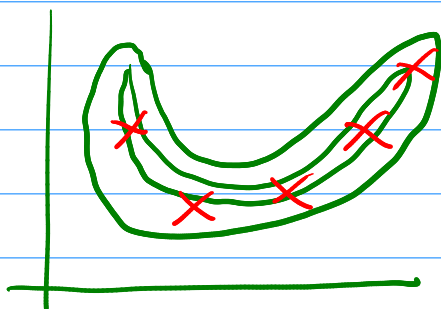
$\text{Cov} > 0$



$\text{Cov} < 0$



$\text{Cov} \approx 0$



$\text{Cov} \approx 0$

(X, Y related, non-linearly)

Notes:

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

② Covariance is scale sensitive:

$$\text{Cov}(5X, Y) = 5 \text{Cov}(X, Y).$$

Defn: Correlation

Basically, re-scaled cov. to be
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 strong lin. rel.

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

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

pf. $Z = aX + bY$

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

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

$$= E[(aX + bY - aEX - bEY)^2]$$

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

$$\downarrow \underbrace{\hspace{10em}}_{\alpha \quad \beta}$$
$$(\alpha + \beta)^2 = \alpha^2 + \beta^2 + 2\alpha\beta$$

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

$$\begin{aligned}&= a^2 \underbrace{E[(X - EX)^2]}_{\text{Var}(X)} + b^2 \underbrace{E[(Y - EY)^2]}_{\text{Var}(Y)} \\ &\quad + 2ab \underbrace{E[(X - EX)(Y - EY)]}_{\text{Cov}(X, Y)}\end{aligned}$$

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

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

pf.

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

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:

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

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

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

pf. $a, c \neq 0$

note: $x \neq 0$

$$\text{Sgn}(x) = \frac{x}{|x|}$$

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

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

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

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

pf WLOG assume

$$EX = EY = 0$$

$$\text{Var} X = \text{Var} Y = 1$$

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

$\underbrace{\hspace{10em}}_{1 \quad 1}$

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

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

↙
 $1 + \text{Cor} \geq 0$

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

↘
 $1 - \text{Cor} \geq 0$

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

Variance Short-cut!

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

Covariance Short-Cut

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