

Lecture 13

Longhai Li, October 21, 2021

Plan:

1. Correlation P ✓ S 2.4.
2. Inclp. ✓ S 2.5

P Correlation Coefficient : Covariance.

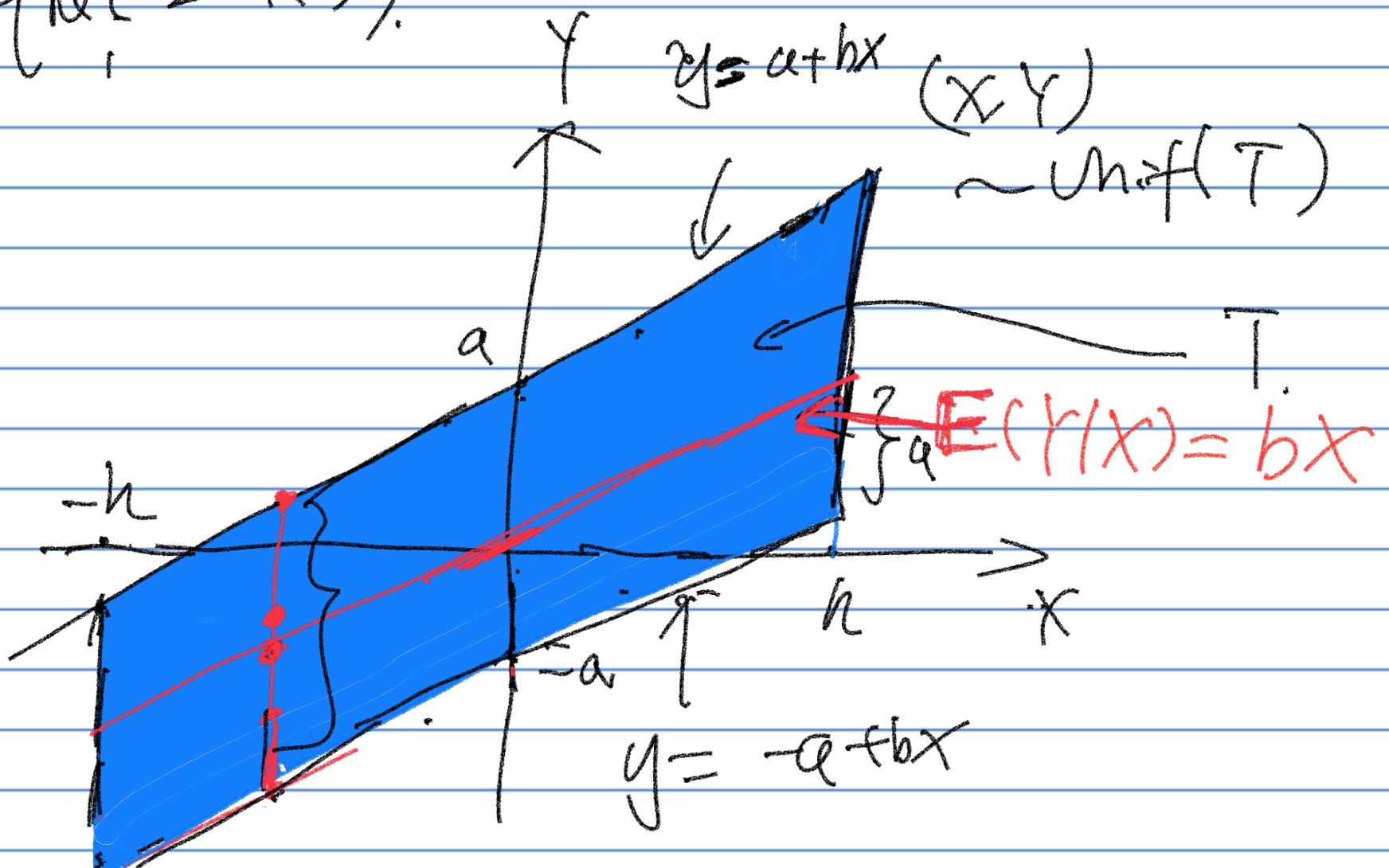
$$\begin{aligned}\text{Cov}(X, Y) &= E((X - \mu_x) \cdot (Y - \mu_y)) \\ &= E(XY) - E(X) \cdot E(Y)\end{aligned}$$

$$\mu_x = E(X), \quad \mu_y = E(Y)$$

$$\text{Cov}(X, X) = V(X)$$

$$\rho = \frac{\text{Cov}(X, Y)}{SD(X) \cdot SD(Y)} = \frac{\text{Cov}(X, Y)}{\sigma_X \cdot \sigma_Y}$$

Example (2.4.3).



$$f(x,y) = \begin{cases} \frac{1}{4ah} & \text{a+bx} < y < a+bx, h < x < h \\ 0 & \text{o.w.} \end{cases}$$

$$X \sim \text{Unif}(-h, h)$$

$$Y|X=x \sim \text{Unif}(a+bx, a+bx)$$

$$E(Y|X) = \frac{a+bx}{2} = \boxed{bx}$$

$$E(Y) = E(E(Y|X)) = b \cdot E(X) = 0$$

$$E(X) = 0$$

$$E(XY) = E(E(XY|X))$$

$$= E(X \cdot E(Y|X))$$

$$= E(X \cdot bX)$$

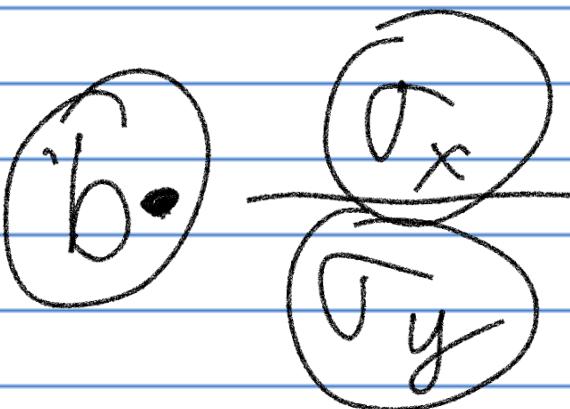
$$= b E(X^2) \stackrel{=0}{=} 0$$

$$\begin{aligned} \text{Cov}(X, Y) &= E(XY) - E(X) \cdot E(Y) \\ &= b E(X^2) \end{aligned}$$

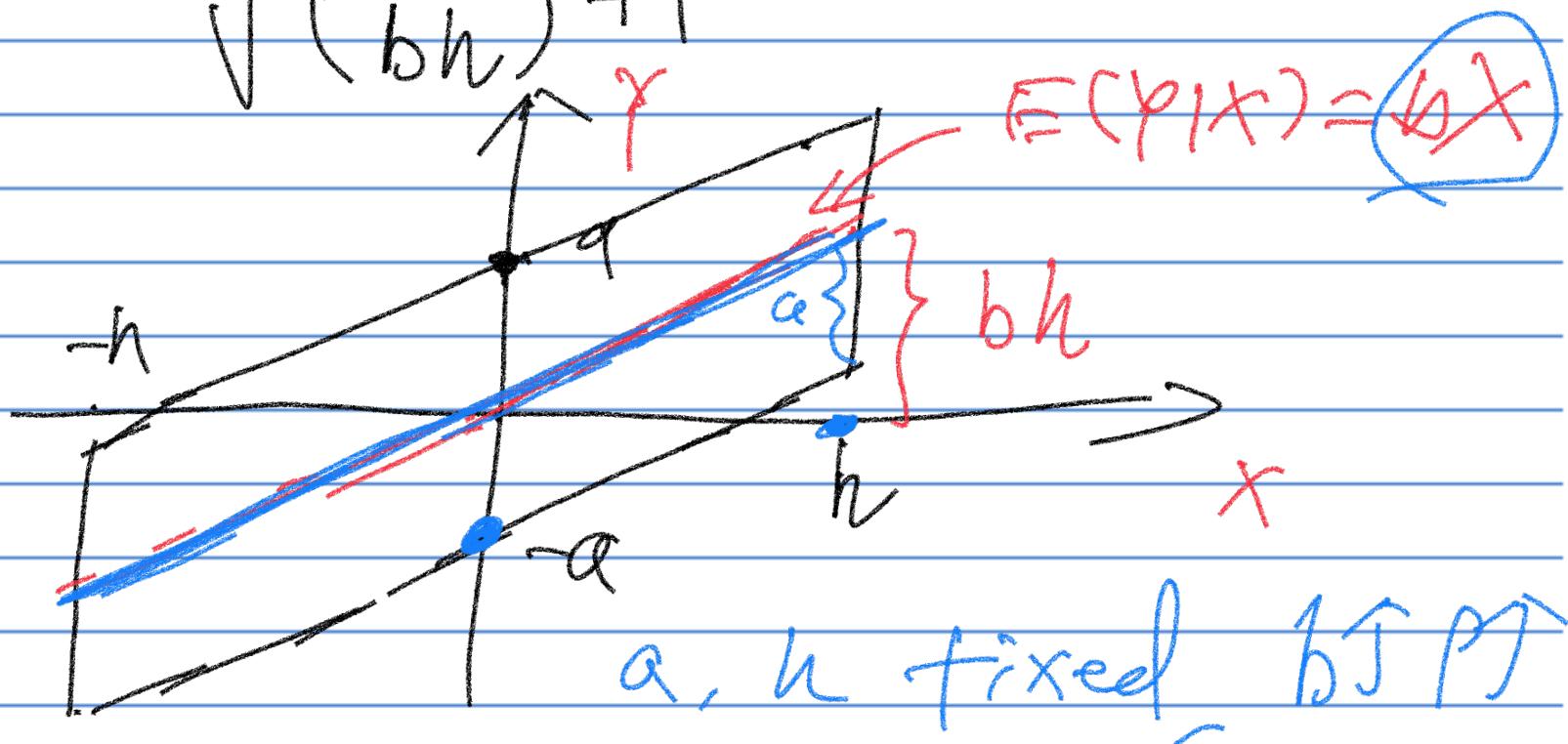
$$\text{Cov}(X, Y) = b \cdot E(X^2) = b \cdot V(X)$$

$$E(X) = 0, \quad V(X) = E(X^2)$$

$$P = \frac{\text{Cov}(X, Y)}{\sqrt{V(X) V(Y)}} = \frac{b \cdot V(X)}{\sqrt{V(X) V(Y)}}$$

$$= \frac{b \cdot h}{\sqrt{a^2 + b^2 h^2}}$$


$$P = \sqrt{\left(\frac{a}{bh}\right)^2 + 1}$$



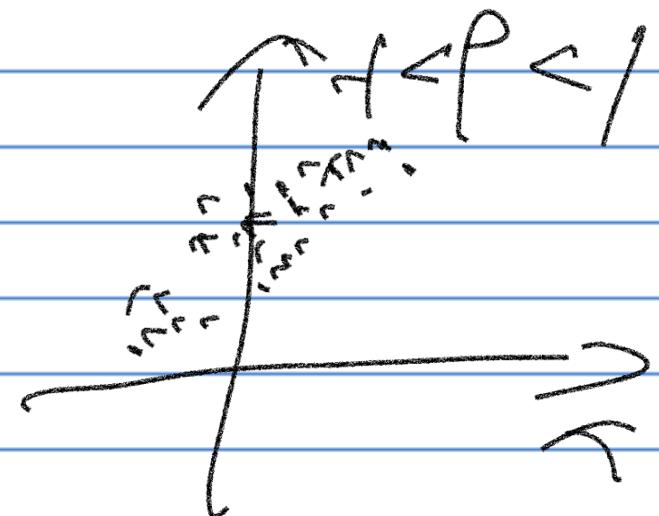
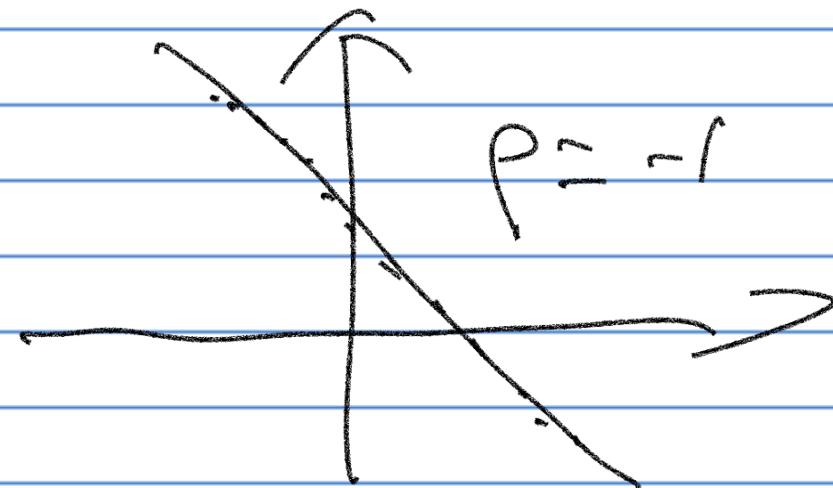
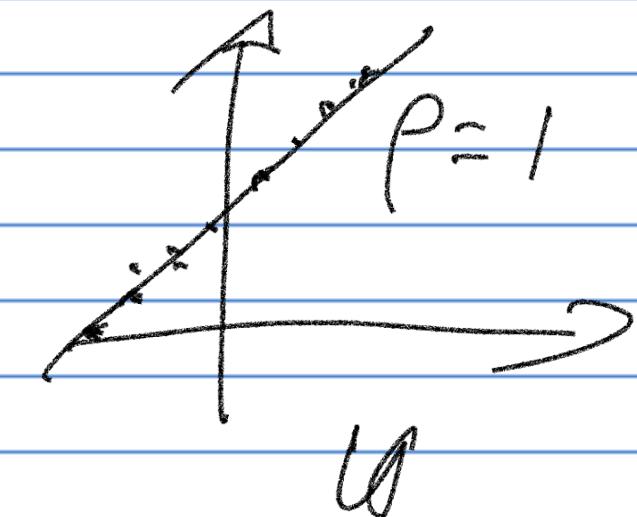
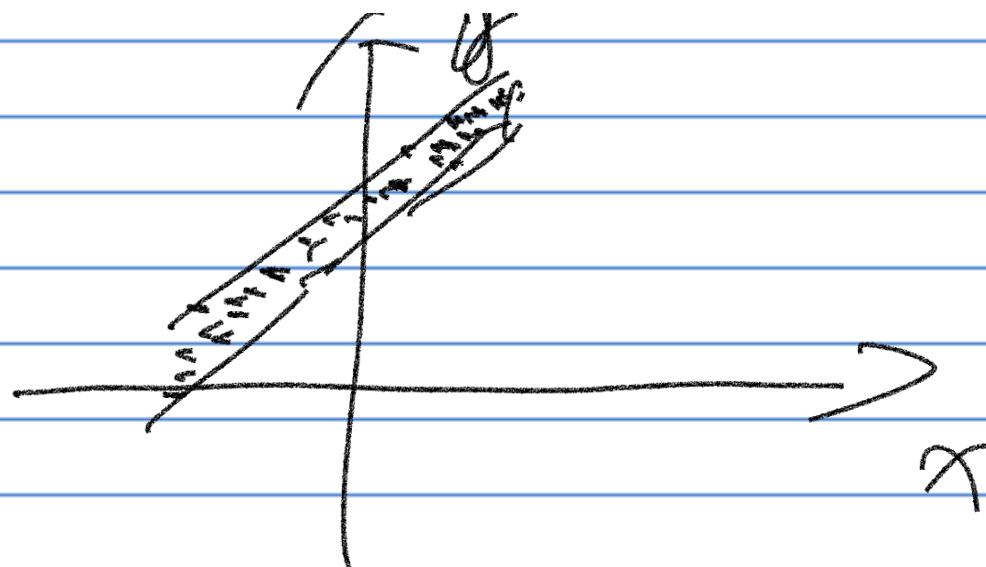
Remarks about ρ :

1. $-1 \leq \rho \leq 1$.

$$|\text{Cov}(X, Y)| \leq SD(X) \cdot SD(Y)$$

$C = S$ They correlate.

2. $\rho = \pm 1$, Perfect (linear relationship).



Thm:

If $E(Y|X)$ is a linear function

then $E(Y|X) = u_y + \rho \frac{\sigma_y}{\sigma_x} (X - u_x)$

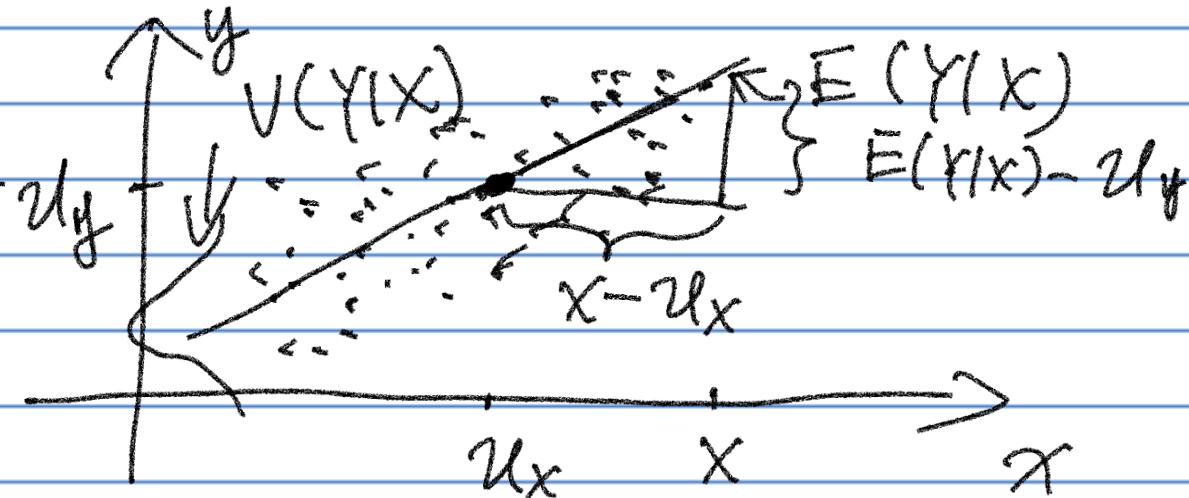
$$\frac{E(Y|X) - u_y}{\sigma_y} = \rho \cdot \frac{X - u_x}{\sigma_x}$$

$$E(V(Y|X)) = \sigma_y^2 (1 - \rho^2)$$

$$V(E(Y|X)) = \sigma_y^2 \rho^2$$

$$\rho^2 = \frac{V(E(Y|X))}{V(X)} = \frac{\sigma_y^2 \rho^2}{\sigma_x^2}$$

ρ^2 is the proportion of Variance of Y that is explained by X .



$$\rho = \frac{[E(Y|x) - u_y] / \sigma_y}{(x - u_x) / \sigma_x}$$

$$\frac{\rho \cdot \sigma_y}{\sigma_x} = \frac{E(Y|x) - u_y}{x - u_x} = \beta$$

$$\sigma_y^2 = V(Y) = \underline{E(V(Y|X))} + V(\overline{E(Y|X)})$$

Variance within
groups

Variance b/w
groups.

$$\rho^2 = \frac{V(E(Y|X))}{V(Y)}$$

Pf: Assume $E(Y|X) = a + bX$

We will link a, b to $P, \bar{u}_x, \bar{u}_y, \sigma_x, \sigma_y$

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

$$(\bar{u}_y = a + b \bar{u}_x)$$

$$\begin{aligned} E(XY) &= E(X E(Y|X)) \\ &= E(X(a + bX)) \\ &= a \bar{u}_x + b E(X^2) \end{aligned}$$

$$E(XY) = a \bar{u}_x + b \cdot (\sigma_x^2 + \bar{u}_x^2)$$

$$\text{b.c. } E(X^2) - [E(X)]^2 = \sigma_x^2$$

$$\rho = \frac{E(XY) - \bar{E}X \cdot \bar{E}Y}{\sigma_x \cdot \sigma_y}$$

$$= \frac{a \bar{u}_x + b \cdot (\sigma_x^2 + \bar{u}_x^2) - \bar{u}_x \bar{u}_y}{\sigma_x \sigma_y}$$

Solving a, b given $\rho, \mu_x, \mu_y, \sigma_x, \sigma_y$:

$$\begin{cases} a = \mu_y - \rho \cdot \frac{\sigma_y}{\sigma_x} \mu_x \\ b = \rho \cdot \frac{\sigma_y}{\sigma_x} \end{cases}$$

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

$$= \mu_y + \rho \frac{\sigma_y}{\sigma_x} (X - \mu_x)$$

$$V(E(Y|X)) = \rho^2 \cdot \frac{\sigma_y^2}{\sigma_x^2} \cdot V(X) = \rho^2 \sigma_y^2$$

Corollary:

If $E(Y|X) = a$, where a is a constant

then $P_{X,Y} = 0$.

pt: $b=0$. $b = \rho \frac{\sigma_y}{\sigma_x}$, $\rho = 0$.

$$\textcircled{R^2} = \frac{\text{SSR}}{\text{SST}} \text{ for } \geq 3 \text{ variables}$$



p^2 for bivariate variables

R^2 is a more general concept
of p^2

Independence

X & Y are indep. if

$$\underline{f(x, y)} = \underline{f_x(x)} \cdot \underline{f_y(y)} \text{ or}$$

$$f(x|y) = f_x(x) \text{ or}$$

$$f(y|x) = f_y(y)$$

Thm: If X & Y are indep.

then $P_{X,Y} = 0$. $\text{Cov}(X,Y) = 0$.

Pf: $f(x,y) = f_X(x) \cdot f_Y(y)$

$$E(X \cdot Y) = \iint \underline{x} \cdot \underline{y} \cdot \underline{f_X(x)} \cdot \underline{f_Y(y)} dxdy$$

$$= \int y \cdot f_Y(y) dy \cdot \int x \cdot f_X(x) dx$$

$$= E(Y) \cdot E(X)$$

$$\text{Cov}(X, Y) = E(X \cdot Y) - E(X) \cdot E(Y)$$

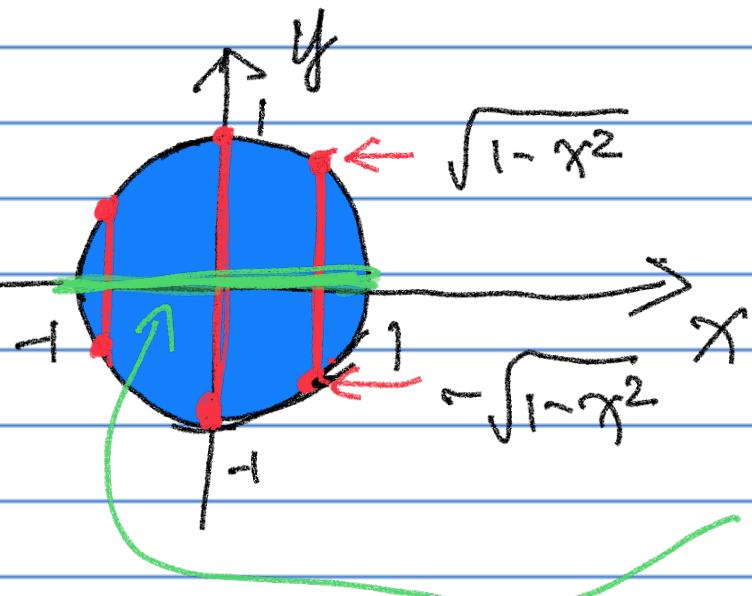
$$= 0$$

$$\rho_{X,Y} = 0$$

Examp(1):

$$(X, Y) \sim \text{Unif}(S^2)$$

$$S^2 = \{(x, y) / y^2 + x^2 \leq 1\}$$



$$Y|X \sim \text{Unif}(-\sqrt{1-x^2}, \sqrt{1-x^2})$$

$$f(y|x) \neq f_x(x)$$

X and Y are not indep.

$$\underline{E(Y|X) = 0 = \theta + 0 \cdot X}$$

$$\begin{aligned} \cdot E(XY) &= E(X \cdot E(Y|X)) \\ &= E(X \cdot 0) = 0 \end{aligned}$$

$$\begin{aligned} \text{Cov}(X, Y) &= E(X \cdot Y) - E(X) \cdot E(Y) \\ &= 0 - 0 \cdot 0 = 0 \end{aligned}$$

Alternativif, $\rho = \rho \cdot \frac{\sigma_Y}{\sigma_X} = 0 \Rightarrow \rho = 0$

$$\rho = 0$$

Thm: If $X \perp Y$

then $E(g(x) \cdot h(y)) = E(g(x)) \underset{\delta}{\circ} E(h(y))$

Thm: If $X \perp Y$

then $F(x, y) = F_x(x) \cdot F_y(y)$.

Thm: If $X \perp Y$

then $P(X \in A, Y \in B)$
 $= P(X \in A) \cdot P(Y \in B)$

Thm: If $X \perp Y$

$$\text{then } M_{X,Y}(t_1, t_2)$$

$$= E(e^{t_1 X + t_2 Y})$$

$$= E(e^{t_1 X} \cdot e^{t_2 Y})$$

$$= E(e^{t_1 X}) \cdot E(e^{t_2 Y})$$

$$= M_X(t_1) \cdot M_Y(t_2)$$