

Today

- 1) Covariance between two random variables
- 2) Correlation coefficient \rightarrow operational meaning, properties.
- 3) Uncorrelated-ness vs Independence
- 4) Iterated Expectation Theorem

Covariance between X & Y $\rightarrow \underline{f_{X,Y}(x,y)}$

$$\begin{aligned} C_{XY} &= E[(X - \mu_X)(Y - \mu_Y)] \\ (C) \quad &= E[X Y - \mu_Y X - \mu_X Y + \mu_X \mu_Y] \\ &= E[XY] - E[\cancel{(\mu_Y \cdot X)}] - E[\cancel{(\mu_X \cdot Y)}] + E[\cancel{\mu_X \mu_Y}] \\ &= E[XY] - \mu_Y \cdot \mu_X - \mu_X \cdot \mu_Y + \mu_X \cdot \mu_Y \end{aligned}$$

$$C_{XY} = E[XY] - \mu_X \cdot \mu_Y$$

\downarrow $\left\{ \begin{array}{l} Y = \text{const} \\ X = \text{random var.} \end{array} \right.$

Correlation coefficient ρ_{XY} (or ρ).

$$\rho_{XY} = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \times \text{Var}(Y)}} = \frac{\text{Cov}(X, Y)}{\sigma_X \cdot \sigma_Y}.$$

$$\left. \begin{array}{l} (1) \quad -1 \leq \rho_{XY} \leq 1. \\ (2) \quad -\sigma_X \sigma_Y \leq \text{Cov}(X, Y) \leq \sigma_X \sigma_Y \end{array} \right\} \text{Basic property of } \rho_{XY}.$$

Interpretation of ρ_{XY}

$\left\{ \text{If } \rho_{XY} > 0 \Rightarrow \text{if } X \text{ is high then } Y \text{ will tend to be high.} \right.$

If $\rho_{xy} < 0 \Rightarrow$ — "opposite" —.

If $\rho_{xy} = 0 \Rightarrow X, Y$ are uncorrelated.

$X \rightarrow$ randomly chosen digits
for Telephone #.

$Y \rightarrow \underline{\text{SSN}}$.

$$\rho_{xy} = ?$$

$X \rightarrow \text{height}$, $Y \rightarrow \text{weight}$.

$$0 < \rho_{xy} \leq 1$$

$X \rightarrow$ distance of a
cell phone tower
from a user.

$Y \rightarrow$ Power of the
received signal.

$$-1 \leq \rho_{xy} \leq 0$$

(ρ is \searrow)

When is $\rho_{xy} = 1$ or $\rho_{xy} = -1$?

\Rightarrow When X and Y have an exact linear
relationship.

$$\begin{cases} X = a_1 Y + b \\ \rho_{xy} = \pm 1 \text{ or } \mp 1 \end{cases}$$

$X \rightarrow$ temp in
Kelvin.

$Y \rightarrow$ temp in
Celsius.

$$\rho_{xy} = +1.$$

$$\boxed{\rho_{xy} = \rho_{yx}}.$$

$$\boxed{\rho_{xy} = 0.3}$$

$$\left\{ \begin{array}{l} X = Y^3 + 3Y^2 + \cos(Y), \\ P_{XY} = 0.3. \end{array} \right. \Rightarrow$$

Uncorrelated R.V.'s		Independent R.V.'s.
X, Y are uncorrelated if.		X, Y are independent if
$\text{Cov}(X, Y) = 0.$		$f_{X,Y}(x,y) = f_X(x) \cdot f_Y(y)$
(or) $P_{XY} = 0$		Joint PDF Product of marginals.
(or) $E[XY] = E[X]E[Y].$		

If X & Y are independent $\Rightarrow X$ & Y are uncorrelated. (or $P_{XY} = 0$)

How to prove this?

$$E[\underline{XY}] = \iint x \cdot y \cdot \underline{f_{X,Y}(x,y)} dx dy.$$

$$E[g(x,y)] = \iint g(x,y) \cdot f_{X,Y}(x,y) dx dy,$$

since X, Y are
indep.

$$\begin{aligned}
 &= \iint x \cdot y \cdot f_X(x) f_Y(y) dx dy \\
 &= (\underbrace{\int x f_X(x) dx}_{= E[X]}) (\underbrace{\int y f_Y(y) dy}_{= E[Y]}). \\
 &= E[X] E[Y].
 \end{aligned}$$

If X and Y are uncorrelated. $\cancel{\xrightarrow{}} \xrightarrow{(not\ true\ in\ general)}$ X and Y are independent.

$$\rho_{XY} = 0$$

$$f_{XY} = f_X f_Y$$

Example: $\{X \sim \text{unif}[0,1], Y \sim \text{unif}[0,1]\}$, and $Z = X + Y$, $W = X - Y$. X, Y are independent.

Prove: (Z, W) are uncorrelated, however they are NOT independent.

$$\rho_{Z,W} = 0.$$

$$\underline{f_{ZW}^{(z,w)} \neq f_Z^{(z)} f_W^{(w)}}$$

$$E[ZW] = ? \quad E(Z) E(W)$$

$$\begin{aligned} E[(X+Y)(X-Y)] &\rightarrow \text{RHS.} \\ = E[X^2 - Y^2] &= E(Z) \cdot E[X-Y] \\ = E[X^2] - E[Y^2] &= 0. \\ = 0. & \quad E[X-Y] = \overline{E(X)} \\ &\quad - \overline{E(Y)} \\ &= 0. \end{aligned}$$

$$\underline{E[Z] \times E[W]}$$

$$= (E[X+Y]) \times E[X-Y].$$

$$= [\underbrace{\{ \}}_{X \sim \text{unif}[0,1]} \times \{ E[X] - E[Y] \}_{Y \sim \text{unif}[0,1]}]$$

$$\underbrace{\text{same mean}}_{\Rightarrow}$$

$$Y = \text{const.} \Rightarrow \text{Var}(Y) = 0.$$

$\uparrow X \rightarrow r.v.$

$$\begin{aligned}\text{cov}(X, Y) & \stackrel{?}{=} E[(X - \mu_X)(Y - \mu_Y)] \\ & = E[(X - \mu_X)(C - C)]\end{aligned}$$

$$= 0.$$

$$\rho_{XY} = 0$$

$$\text{cor}(X, Y) = 0.$$

Iterated Expectation Theorem

(X, Y)

$$E[E[X|Y]] = E[X]$$

$E[X] \rightarrow$ mean of X . $E[X] = \underline{\int x f_X(x) dx}$.

$E[X|Y=y] \rightarrow$ conditional mean of X given $Y=y$.

$$\begin{aligned}& \text{Function of } y \\& \downarrow \\& \int_{x=-\infty}^{\infty} x \cdot f_{X|Y}(x|y) dx\end{aligned}$$

$$X|Y=y \Rightarrow \underline{f_{X|Y}(x|y)}$$

conditional PDF of
 X given $Y=y$.

as y changes, $E[X|Y=y]$ changes.

$$E[X|Y]$$

is a random variable.

* What values does it take?

it is a function of

Suppose $Y \in \{0, 1, 2\}$.

$$\boxed{Y}$$

$$Y=0 \longrightarrow E[X|Y=0].$$

$$Y=1 \longrightarrow E[X|Y=1]$$

$$Y=2 \longrightarrow E[X|Y=2].$$

$$E[E[X|Y]] = \underline{\underline{E[X]}}.$$

is a R.V. $\Rightarrow g(Y)$.

$$LHS = E[g(Y)].$$

$$= \int_{-\infty}^{\infty} \underbrace{g(y)}_{y=-\infty} \cdot f_Y(y) dy. \quad \text{OK?}$$

$$= \int_{-\infty}^{\infty} [E[X|Y=y]] f_Y(y) dy.$$

$$= \int_{-\infty}^{\infty} f_Y(y) \left[\int_{x=-\infty}^{\infty} \underbrace{f_{X|Y}(x|y)}_{x=-\infty} dx \right] dy.$$

$$= \int_{x=-\infty}^{\infty} dx \times \left[\int_{y=-\infty}^{\infty} f_Y(y) f_{X|Y}(x|y) dy \right] dx.$$

$$= \int_{x=-\infty}^{\infty} dx f_X(x). dx = E[X].$$

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