

Lecture (30)

P ①

Recap:

The Probabilistic Method

Covariance

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

independence $\Rightarrow \text{Cov}(X, Y) = 0$

\nLeftarrow

Variance of Binomial r.v.

Correlation

$$\begin{aligned} \rho(X, Y) &= \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X) \text{Var}(Y)}} \\ &= \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y} \end{aligned}$$

if X & Y are independent (2)

$$\rho(X, Y) = 0.$$

$$\rho(X, Y) = \frac{\text{cov}(X, Y)}{\sigma_X \sigma_Y}$$



it is defined when

$$\sigma_X \neq 0, \sigma_Y \neq 0$$

$$\text{Var}(X) \neq 0 \text{ AND } \text{Var}(Y) \neq 0$$

When is $\text{Var}(X) = 0$?
|| X is a constant.

$$D: \sum p_i (x_i - \mu)^2 = 0$$

$$C: \int f(x) (x - \mu)^2 dx$$

$\rho(x, y)$ is defined (3)
if x is NOT a constant
AND
 y is NOT a constant

$$-1 \leq \rho(x, y) \leq +1$$

$$Z = \frac{x}{\sigma_x} + \frac{y}{\sigma_y}$$

$$\text{Var}(Z) \geq 0$$

$$\text{Var}\left(\frac{x}{\sigma_x} + \frac{y}{\sigma_y}\right) =$$

$$\text{Var}(\sum x_i) = \sum \text{Var}(x_i) + 2 \sum_{i < j} \text{cov}(x_i, x_j)$$

$$\text{Var}\left(\frac{X}{\sigma_x} + \frac{Y}{\sigma_y}\right)$$

④

$$= \text{Var}\left(\frac{X}{\sigma_x}\right) + \text{Var}\left(\frac{Y}{\sigma_y}\right)$$

$$+ 2 \text{Cov}\left(\frac{X}{\sigma_x}, \frac{Y}{\sigma_y}\right)$$

$$= \text{Var}\left(\frac{1}{\sigma_x} \cdot X\right) + \text{Var}\left(\frac{1}{\sigma_y} \cdot Y\right)$$

$$+ 2 \text{Cov}\left(\frac{1}{\sigma_x} \cdot X, \frac{1}{\sigma_y} \cdot Y\right)$$

$$= \frac{1}{\sigma_x^2} \text{Var}(X) + \frac{1}{\sigma_y^2} \text{Var}(Y) + 2 \cdot \frac{1}{\sigma_x} \cdot \frac{1}{\sigma_y} \text{Cov}(X, Y)$$

⑤

$$1 + 1 + 2 \rho(x, y) \geq 0$$

$$2 \rho(x, y) \geq -2$$

$$\rho(x, y) \geq -1$$

$$\rho(x, y) \leq 1$$

$$w = \frac{x}{\sigma_x} - \frac{y}{\sigma_y}$$

$$\text{Var}(w) \geq 0$$

f.w.

⑥

What happens

when $\rho = -1$?

When is $\rho = -1$?

when $\text{Var}(Z) = 0$

$$\text{Var}\left(\frac{X}{\sigma_X} + \frac{Y}{\sigma_Y}\right) = 0$$

\implies

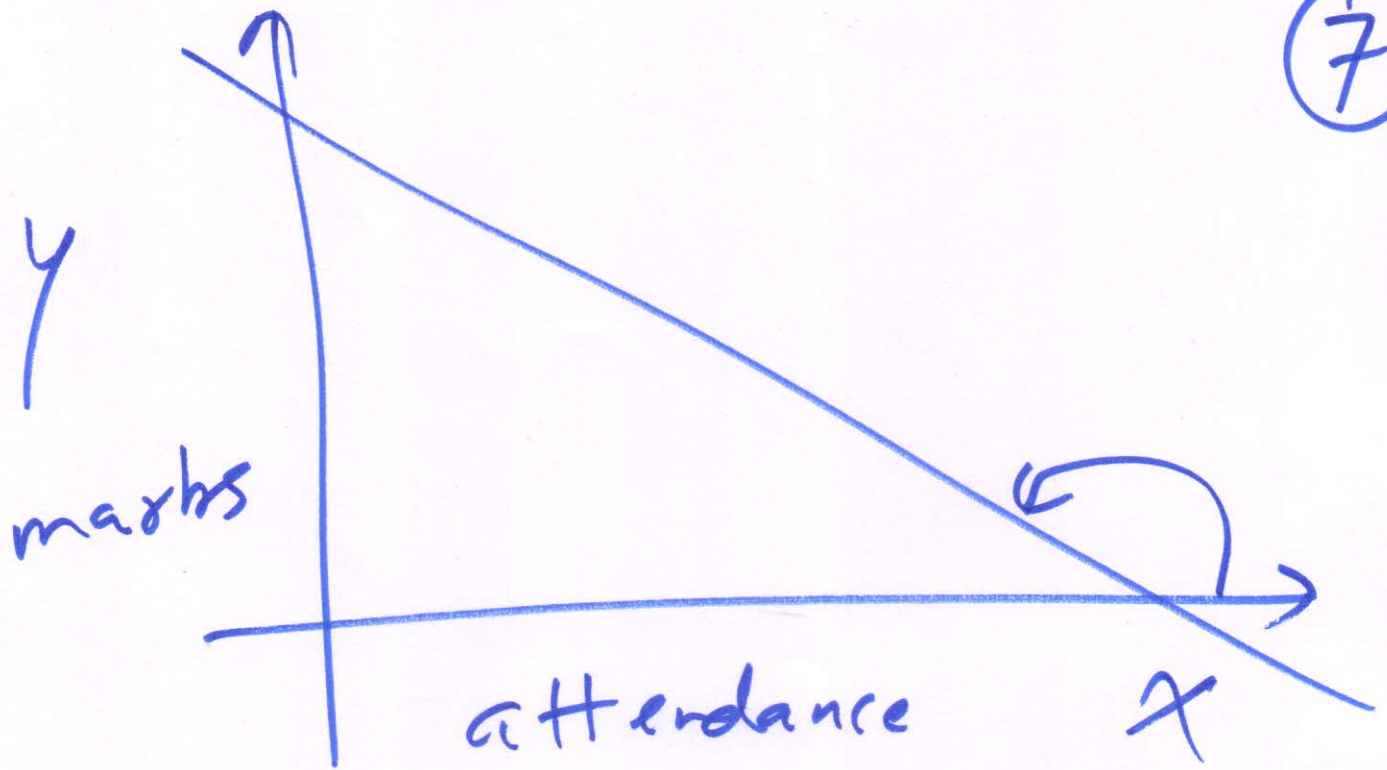
$$\frac{X}{\sigma_X} + \frac{Y}{\sigma_Y} = C$$

$$\frac{Y}{\sigma_Y} = C - \frac{X}{\sigma_X}$$

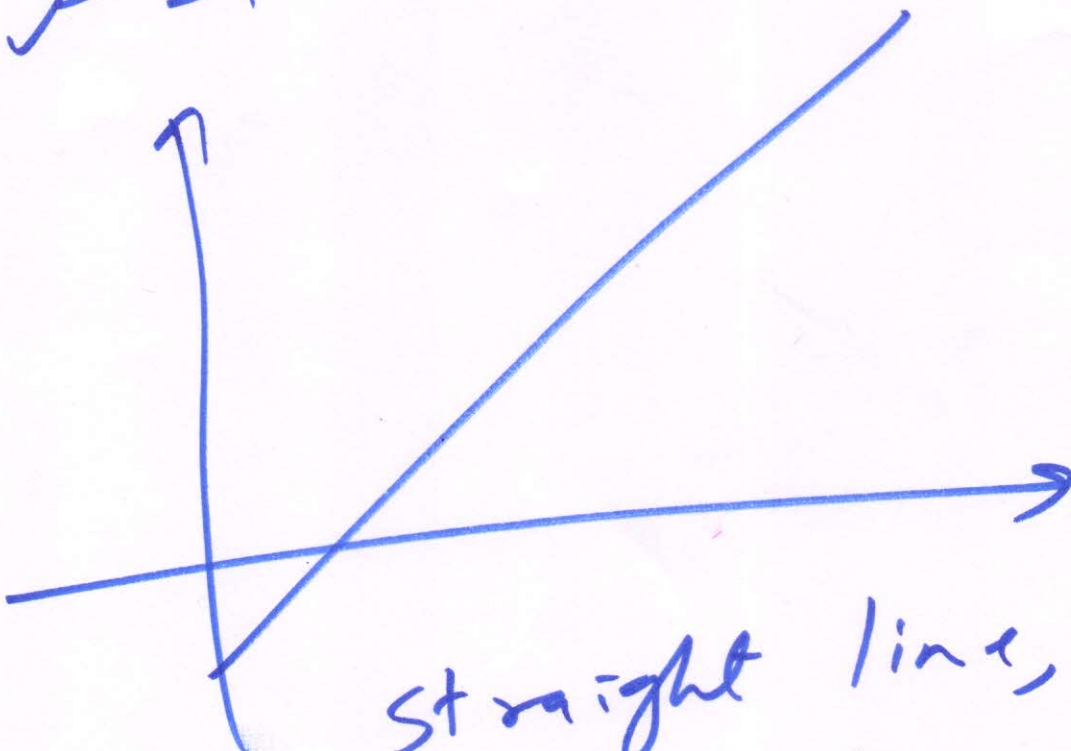
$$Y = C \cdot \sigma_Y - \left(\frac{\sigma_Y}{\sigma_X}\right) X$$

Y is linear in X . Slope is negative.

⑦



$$\rho = 1$$



straight line,
with +ive slope.

If $\rho = 0$, we say $\textcircled{8}$
that X & Y are
uncorrelated.

Properties of
Conditional Expectation.

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

e.g. $f(x, y) = \frac{e^{-x/y} e^{-y}}{y}$

Compute $0 < x < \infty$
 $0 < y < \infty$

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

Continuous case

⑨

$$E[X|Y=z] = \int_{-\infty}^{\infty} x f_{X|Y}(x|z) dx$$

?

Discrete case

$$E[X|Y=z] = \sum_x x p_{X|Y}(x|z)$$

$$f_{X|Y}(x|z) = \frac{f_{X,Y}(x,z)}{f_Y(z)} \rightarrow \text{given}$$

?

$$f_Y(z) = \int_{-\infty}^{\infty} f_{X,Y}(x,z) dx$$

(10)

$$f_{X|Y}(x|y)$$

$$= \frac{1}{y} e^{-x/y}$$

exponential

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

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

$$D: E[X] = \sum_y \underbrace{E[X|Y=z]} P(Y=z)$$

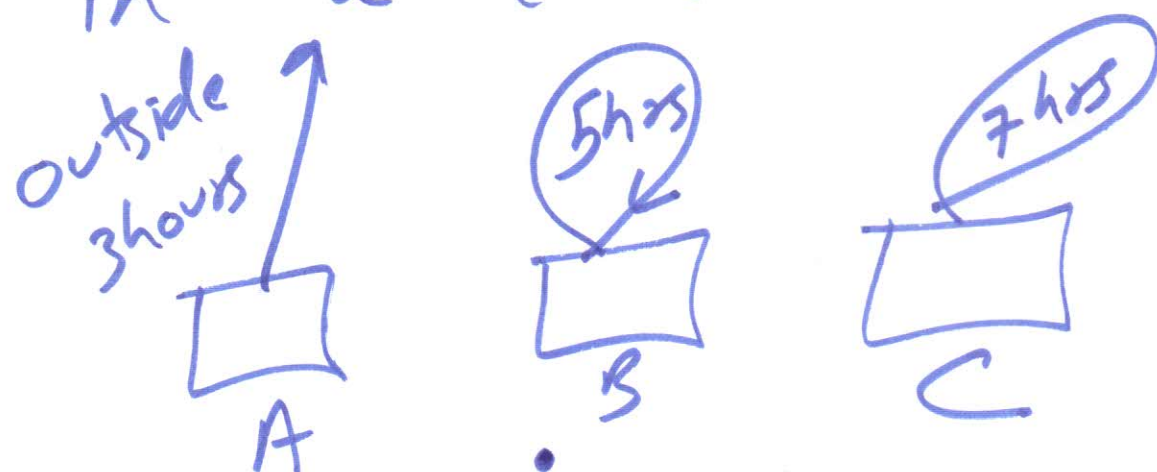
$$C: E[X] = \int_{-\infty}^{\infty} \underbrace{E[X|Y=z]} f_Y(y) dy$$

(considering $E[X|Y]$ as a random variable, which is a function of Y .)

ex:

11

A miner is trapped
in a coal mine.



X = no. of hours before
he gets out

$$Y = \left\{ \begin{array}{ccc} 1, & 2, & 3 \\ A, & B, & C \end{array} \right\}$$

$$\begin{aligned} E[X] &= E[E[X|Y]] \\ &= E[X|Y=1]P(Y=1) + \\ &\quad E[X|Y=2]P(Y=2) + \\ &\quad E[X|Y=3]P(Y=3) \end{aligned}$$

$$P(Y=1) = P(Y=2) = P(Y=3) = \frac{1}{3} \quad (12)$$

$$E[X|Y=1] = 3$$

$$E[X|Y=2] = 5 + E[X]$$

$$E[X|Y=3] = 7 + E[X]$$