

# HW 2

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Let  $X$  and  $Y$  be bivariate Gaussian with mean  $(\mu_X, \mu_Y)$ , common variance  $\sigma^2$ , and correlation coefficient  $\rho > 0$ . The bivariate mean is random and determined by  $Z$ , where

$$(\mu_X, \mu_Y) = \begin{cases} (-\mu, \mu) & Z = -1, \\ (\mu, -\mu) & Z = 1 \end{cases}$$

with equal probability for  $Z = -1$  and  $Z = 1$ .

## 1

(a) The conditional expectation  $E[Y|X, Z]$  can be calculated as:

$$E[Y|X, Z] = \mu_Y + \frac{Cov(X, Y)}{Var(X)}(X - \mu_X)$$

Since  $\rho = \frac{Cov(X, Y)}{\sqrt{Var(X)Var(Y)}}$ , we have:

$$\begin{aligned} \rho &= \frac{Cov(X, Y)}{\sqrt{\sigma^2 \dot{\sigma}^2}} \\ Cov(X, Y) &= \rho \sigma^2 \end{aligned}$$

Substituting this back into our equation:

$$\begin{aligned} E[Y|X, Z] &= \mu_Y + \frac{\rho \sigma^2}{\sigma^2}(X - \mu_X) \\ E[Y|X, Z] &= \mu_Y + \rho(X - \mu_X) \end{aligned}$$

$\therefore$

$$E[Y|X, Z] = \begin{cases} \mu - \rho(X + \mu) & Z = -1, \\ -\mu + \rho(X - \mu) & Z = 1 \end{cases}$$

The CEF consequently is linear in  $X$  for each fixed  $Z$ .  
The conditional expectation  $E[Y|Z]$  is:

$$E[Y|Z] = \mu_Y + \frac{Cov(X, Y)}{Var(X)}(E[X|Z] - \mu_X)$$

$$E[Y|Z] = \mu_Y + \rho(\mu_X - \mu_X)$$

$$E[Y|Z] = \mu_Y$$

$$E[Y|Z] = \begin{cases} \mu & Z = -1, \\ -\mu & Z = 1 \end{cases}$$

The CEF consequently is constant for each fixed  $Z$ .  
The conditional expectation  $E[Y|X]$  is:

$$E[Y|X] = E[E[Y|X, Z]|X] = \frac{1}{2}(\mu - \rho(X + \mu)) + \frac{1}{2}(-\mu + \rho(X - \mu))$$

$$E[Y|X] = \rho X - (1 - \rho)\mu$$

The CEF consequently is linear in  $X$ .