

Consider n independent tosses of coin with heads probability p .

Let X and Y be numbers of heads/tails and look at correlation coefficient of X and Y . Note that $X+Y=n$ so $E[X]+E[Y]=n$.

Thus, $X - E[X] = -(Y - E[Y])$.

now to do computation,

$$\begin{aligned} \text{cov}(X, Y) &= E[(X - E[X])(Y - E[Y])] \\ &= -E[(X - E[X])^2] = -\text{var}(X). \end{aligned}$$

so
$$\rho_{XY} = \frac{\text{cov}(X, Y)}{\sqrt{\text{var}(X)\text{var}(Y)}} = \frac{-\text{var}(X)}{\sqrt{\text{var}(X)\text{var}(X)}} = -1.$$

Suppose we define matrix of covariances, so (i, j) entry is $\text{cov}(x_i, x_j)$.

Let $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$ have covariance matrix $\begin{bmatrix} 5 & 2 & 0 \\ 2 & 5 & 2 \\ 0 & 2 & 5 \end{bmatrix}$

(a) find $\text{cov}(x_1+x_2, x_1+x_3)$

$$\begin{aligned} &= \text{cov}(x_1, x_1) + \text{cov}(x_1, x_3) + \text{cov}(x_2, x_1) + \text{cov}(x_2, x_3) \\ &= 5 + 0 + 2 + 2 = 9. \end{aligned}$$

(b) find a so $x_2 - ax_1$ is uncorrelated with x_1 .

$$\begin{aligned} \text{cov}(x_2 - ax_1, x_1) &= \text{cov}(x_2, x_1) - a\text{cov}(x_1, x_1) = 2 - 5a \\ &\text{which is zero for } a = \frac{2}{5}. \end{aligned}$$

MMSE estimation

Given data X , we want to infer a random variable Y that cannot be observed directly.

- In communication, from received signal X , infer transmitted signal Y .
- From noisy measurements, infer value Y of true quantity.
→ wastewater treatment.
- From noisy measurements of past positions of vehicle X , infer present position in inertial guidance system Y .
- From observations X on temperature, pressure at several locations, infer future temperature Y at a given site.
- From measurements from array of sensors, infer location of callers in cellular telephony environment.

Want to find an estimator $\hat{Y}(X)$ or $\hat{Y} = g(X)$ that minimizes mean square error (MSE) among all possible functions of data:

$$E[(Y - g(X))^2].$$

best estimator, MMSE estimator, is called $g^*(X)$.

Derive the estimator:

$$\begin{aligned} E[(Y - \hat{Y})^2] &= E[(Y - E[Y|X] + E[Y|X] - \hat{Y})^2] \\ &= E[(Y - E[Y|X])]^2 + 1 \end{aligned}$$

Intuitively it seems $E[Y|X]$, the conditional is a good thing to use, i.e. the mean conditioned on the given data.