
On Linear Estimator Formula

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August 03, 2017

1 PROBLEM STATEMENT

Suppose X, Y_1, Y_2, \dots, Y_n are random scalars and Y_1, Y_2, \dots, Y_n have finite second moments. According to [1], the linear estimation of X based on Y_1, Y_2, \dots, Y_n is defined as

$$\hat{E}[X|Y] = E[X] + \text{Cov}[X, Y]\text{Cov}[Y]^{-1}(Y - E[Y]). \quad (1.1)$$

This requires that $\text{Cov}[Y]$ is not singular, which means $\text{Cov}[Y]^{-1}$ exists. Prove that $\text{Cov}[Y]$ is not singular only if none of Y_i can be written as linear combination of the other random scalars.

2 ELABORATION

Without loss of generality, we suppose that

$$Y_1 = kZ. \quad (2.1)$$

where $Z^T = [Y_2, Y_3, \dots, Y_n]$ and $\text{Cov}[Z]$ is not singular. Then $\text{Cov}[Y]$ is calculated as

$$\text{Cov}[Y] = \begin{bmatrix} \text{Cov}[Y_1] & \text{Cov}[Y_1, Z] \\ \text{Cov}[Z, Y_1] & \text{Cov}[Z] \end{bmatrix} = \begin{bmatrix} k\text{Cov}[Z]k^T & k\text{Cov}[Z] \\ \text{Cov}[Z]k^T & \text{Cov}[Z] \end{bmatrix}$$

Since $\text{Cov}[Z]$ is not singular, according to **Note 11**, the *Schur Diagonalization* of $\text{Cov}[Y]$ is calculated as

$$\text{Cov}[Y] = \begin{bmatrix} k\text{Cov}[Z]k^T & k\text{Cov}[Z] \\ \text{Cov}[Z]k^T & \text{Cov}[Z] \end{bmatrix} = \begin{bmatrix} I & k \\ 0 & I \end{bmatrix} \begin{bmatrix} M/\text{Cov}[Z] & 0 \\ 0 & \text{Cov}[Z] \end{bmatrix} \begin{bmatrix} I & 0 \\ k^T & I \end{bmatrix}$$

Since

$$M/\text{Cov}[Z] = k\text{Cov}[Z]k^T - k\text{Cov}[Z]\text{Cov}[Z]^{-1}\text{Cov}[Z]k^T = 0,$$

so

$$\text{Cov}[Y] = \begin{bmatrix} I & k \\ 0 & I \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & \text{Cov}[Z] \end{bmatrix} \begin{bmatrix} I & 0 \\ k^T & I \end{bmatrix}$$

is singular in this case.

Therefore the linear estimator formula indicates that none of Y_i can be linear defined by the rest.

REFERENCES

- [1] Bruce Hajek, *Random processes for engineers*, Cambridge University Press, 2015.