

Statistics: Math

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Chapter 1 Review of statistics

1.1 Random Vectors

Suppose a random vector $\vec{Z} \in \mathbb{R}^{m \times 1}$ has mean $\vec{\mu} = \mathbb{E}(\mathbf{Z}) = \left[\mathbb{E}(Z_1), \mathbb{E}(Z_2), \cdots, \mathbb{E}(Z_m)\right]^T$ and variance-covariance matrix: $\Sigma_{m \times m} = Cov(\mathbf{Z}) = \mathbb{E}((\mathbf{Z} - \mu)(\mathbf{Z} - \mu)^T) = \begin{bmatrix} Var(Z_1) & \cdots & Cov(Z_1, Z_m) \\ & \cdots & & \cdots \\ Cov(Z_m, Z_1) & \cdots & Var(Z_m) \end{bmatrix}$.

1.2 Affine Transformation

(1) The affine transformation $\mathbf{W} = \mathbf{a}_{n \times 1} + \mathbf{B}_{n \times m} \mathbf{Z}_{m \times 1}$ should have following properties:

$$\mathbb{E}(\mathbf{W}) = \mathbf{a} + \mathbf{B}\mu, \ Cov(\mathbf{W}) = \mathbf{B}\Sigma\mathbf{B}^T$$

(2) The affine transformation $\mathbf{W} = \mathbf{v}^T \mathbf{Z} = v_1 Z_1 + ... + v_m Z_m$ should have following properties:

$$\mathbb{E}(\mathbf{W}) = \mathbf{v}^T \mu = \sum_{i=1}^m v_i \mu_i$$

$$Var(\mathbf{W}) = \mathbf{v}^T \Sigma \mathbf{v} = \sum_{i=1}^m v_i^2 Var(Z_i) + 2 \sum_{i < j} v_i v_j Cov(Z_i, Z_j)$$
 i.e.
$$\mathbb{E}(\mathbf{AZ}) = \mathbf{A}\mathbb{E}(Z); \ Var(\mathbf{AZ}) = \mathbf{A}Var(\mathbf{Z})\mathbf{A}^T$$

(3)
$$Cov(\mathbf{AX}, \mathbf{BY}) = \mathbb{E}[(\mathbf{AX} - \mathbf{A}\mathbb{E}(X))(\mathbf{BY} - \mathbf{B}\mathbb{E}(Y))^{T}]$$
$$= \mathbf{A}\mathbb{E}[(\mathbf{X} - \mathbb{E}(X))(\mathbf{Y} - \mathbb{E}(Y))^{T}]\mathbf{B}^{T}$$
$$= \mathbf{A}Cov(\mathbf{X}, \mathbf{Y})\mathbf{B}^{T}$$