13. DISCRIMINATION OF TWO POPULATIONS WITH COMMON UNKNOWN COVARIANCE MATRIX. G_{13} -ANDERSON-FISHER STATISTICS ESTIMATOR

Let $\vec{x}_i, \ \vec{y}_j; \ i = 1, ..., n_1; \ j = 1, ..., n_2$ be independent observations of *m*-dimensional random vectors $\vec{\xi} = \vec{a}_1 + \sqrt{R}\vec{\mu}, \ \vec{\mu}^T = \{\mu_i, \ i = 1, ..., m\}$ and $\vec{\eta} = \vec{a}_2 + \sqrt{R}\vec{\nu}, \ \vec{\nu}^T = \{\nu_i, \ i = 1, ..., m\}$ respectively and suppose that random variables $\mu_i, \ \nu_i, \ i = 1, ..., m$ are independent for every *n*.

For the observation classification in the case of two Normal populations, the so-called discriminant function is used:

$$U(\vec{x}) = \left\{ \vec{x} - \frac{1}{2} \left(\vec{\alpha}_1 + \vec{\alpha}_2 \right) \right\}^T R^{-1} \left(\vec{\alpha}_1 - \vec{\alpha}_2 \right), \vec{x}^T = (x_1, \cdots, x_m).$$

We use empirical mean value vectors and the covariance matrix R,

$$\hat{\vec{a}}_1 = n_1^{-1} \sum_{i=1}^{n_1} \vec{x}_i, \quad \hat{\vec{a}}_2 = n_2^{-1} \sum_{i=1}^{n_2} \vec{y}_i,$$

$$\hat{R} = \frac{1}{n_1 + n_2 - 2} \left\{ \sum_{i=1}^{n_1} \left(\vec{x}_i - \hat{\vec{a}}_1 \right) \left(\vec{x}_i - \hat{\vec{a}}_1 \right)^T + \sum_{i=1}^{n_2} \left(\vec{y}_i - \hat{\vec{a}}_2 \right) \left(\vec{y}_i - \hat{\vec{a}}_2 \right)^T \right\}.$$

We shall refer to the expression

$$G_{13}\left(\vec{x}\right) = \left\{\vec{x} - \frac{1}{2}\left(\hat{\vec{a}}_{1} + \hat{\vec{a}}_{2}\right)^{T}\hat{R}^{-1}\left(\hat{\vec{a}}_{1} - \hat{\vec{a}}_{2}\right)\right\}\frac{n_{1} + n_{2} - 2 - m_{1}}{n_{1} + n_{2} - 2}$$

as the $G_{13}(\vec{x})$ -estimator of the discriminant function.

THEOREM 13.1. [Gir54, p.611] If in addition to the conditions of Theorem 11.1

$$\lim_{n_1, n_2 \to \infty} \left(\vec{a}_1 + \vec{a}_2 \right)^T R^{-1} \left(\vec{a}_1 + \vec{a}_2 \right) \left[n_1^{-1} + n_2^{-1} \right] = 0,$$

then

$$p \lim_{n \to \infty} \{ G_{13} (\nu_i) - U (\nu_i) \} = 0,$$

where $\vec{\nu}_i$ is an observation of vector $\vec{\xi}$ or $\vec{\eta}$ which does not depend on $\vec{x}_i, \ \vec{y}_j; \ i = 1, ..., n_1; \ j = 1, ..., n_2$ and distributed as $N\{\vec{a}_1, R\}$ or $N\{\vec{a}_2, R\}$.

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