15. G_{15} -estimator of the nonlinear discriminant function, obtained by observation of random vectors with different covariance matrices

In the case of classifying into two populations based on the Normal distribution, the nonlinear discriminant function is equal to

$$V(\vec{x}) = \frac{1}{2} \left\{ -\left(\vec{x} - \vec{a}_{1}\right)^{T} R_{1}^{-1} \left(\vec{x} - \vec{a}_{1}\right) + \left(\vec{x} - \vec{a}_{2}\right)^{T} R_{2}^{-1} \left(\vec{x} - \vec{a}_{2}\right) - \ln \det R_{1} R_{2}^{-1} \right\}.$$

Let $\vec{x}_i, \ \vec{y}_j; \ i = 1, \ldots, n_1; \ j = 1, \ldots, n_2$ be independent observations of *m*-dimensional random vectors $\vec{\xi}$ and $\vec{\eta}$ respectively; these vectors $\vec{\xi}$ and $\vec{\eta}$ are independent and distributed as $N\{\vec{a}_1, R\}, \ N\{\vec{a}_2, R\}$. As the empirical mean vectors and the covariance matrices R_i , we take:

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

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

We shall refer to the expression

$$\frac{1}{2} \left[-\left(\vec{x} - \hat{\vec{a}}_{1}\right)^{T} \hat{R}_{1}^{-1} \left(\vec{x} - \hat{\vec{a}}_{1}\right) \frac{n_{1} - 1 - m}{n_{1} - 1} + \frac{m}{n_{1}} + \left(\vec{x} - \hat{\vec{a}}_{2}\right)^{T} \hat{R}_{2}^{-1} \left(\vec{x} - \hat{\vec{a}}_{2}\right) \frac{n_{2} - 1 - m}{n_{2} - 1} + \frac{m}{n_{2}} \right] - \frac{1}{2} \ln \frac{\det \hat{R}_{1} \hat{R}_{2}^{-1}}{\left(1 - mn_{1}^{-1}\right) \left(1 - mn_{2}^{-1}\right)}$$

as the $G_{15}(\vec{x})$ -estimator of the nonlinear discriminant function.

THEOREM 15.1. [Gir54, p.615] Let random variables μ_i , ν_i , i = 1, ..., m be independent for every n, $\mathbf{E} \mu_i = \mathbf{E} \nu_i = 0$, $\mathbf{E} \mu_i^2 = \mathbf{E} \nu_i^2 = 1$, i = 1, ..., m, for a certain $\beta > 0$

$$\begin{split} \sup_{n} \max_{i=1,\dots,m} \mathbf{E} \left[\left| \mu_{i} \right|^{4+\beta} + \left| \nu_{i} \right|^{4+\beta} \right] < \infty, \\ \inf_{n} \min_{i=1,\dots,m} \lambda_{i} \left(R \right) > 0, \quad \sup_{n} \max_{i=1,\dots,m} \lambda_{i} \left(R \right) < \infty, \end{split}$$

and the G-condition be satisfied:

$$\lim_{m \to \infty} m n_1^{-1} < 1, \quad \lim_{m \to \infty} m n_2^{-1} < 1,$$
$$\sup_{m} (\vec{a}_1 - \vec{a}_2)^T (\vec{a}_1 - \vec{a}_2) < \infty.$$

Then,

 $p \lim_{n \to \infty} \{ G_{15}(\nu) - V(\nu) \} = 0,$

where ν is an observation of vector $\vec{\xi}$ or $\vec{\eta}$, which does not depend on \vec{x}_i, \vec{y}_i .

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