

Asymptotic expansion for the distribution of the discriminant function in the first order autoregressive processes

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1. Introduction

This paper is concerned with the problem of classifying a series of T observations coming from one of two first order autoregressive Gaussian processes. Let Π_j ($j=1, 2$) denote the first order autoregressive Gaussian process which satisfies the stochastic equation

$$(1.1) \quad y_t = \alpha_j y_{t-1} + u_t \quad (t = \dots - 1, 0, 1, \dots)$$

where α_j ($|\alpha_j| < 1$) is known, and u_t 's are independent identically distributed as $N(0, \sigma_j^2)$ with known variance σ_j^2 . Suppose that $y = (y_1, \dots, y_T)'$ is a series of T observations coming from Π_1 or Π_2 . It is natural to consider a classification method based on the density of y . The density of y is given by

$$(1.2) \quad f(y; \alpha_j, \sigma_j^2) = (2\pi\sigma_j^2)^{-T/2} (1 - \alpha_j^2)^{1/2} \exp\left(-\frac{1}{2} y' \Sigma_j^{-1} y\right)$$

when y comes from Π_j , where

$$\Sigma_j^{-1} = \frac{1}{\sigma_j^2} \begin{pmatrix} 1 & -\alpha_j & & & \\ -\alpha_j & 1 + \alpha_j^2 & & & 0 \\ & \ddots & \ddots & \ddots & \\ 0 & & \ddots & 1 + \alpha_j^2 & -\alpha_j \\ & & & -\alpha_j & 1 \end{pmatrix}.$$

Since all the parameters are known, an optimum classification rule is based on the statistic (cf. Anderson [1])

$$(1.3) \quad V = \log \frac{f(y; \alpha_1, \sigma_1^2)}{f(y; \alpha_2, \sigma_2^2)} = Z - \frac{1}{2} \left\{ T \log \frac{\sigma_1^2}{\sigma_2^2} - \log \frac{1 - \alpha_1^2}{1 - \alpha_2^2} \right\},$$

where

$$(1.4) \quad Z = \frac{1}{2} y' (\Sigma_2^{-1} - \Sigma_1^{-1}) y.$$