

Maximum likelihood estimation of autoregressive models with a near unit root and Cauchy errors

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Received: 20 December 2016 / Revised: 29 April 2018 / Published online: 14 June 2018 © The Institute of Statistical Mathematics, Tokyo 2018

Abstract This paper studies maximum likelihood estimation of autoregressive models of order 1 with a near unit root and Cauchy errors. Autoregressive models with an intercept and with an intercept and a linear time trend are also considered. The maximum likelihood estimator (MLE) for the autoregressive coefficient is $n^{3/2}$ -consistent with *n* denoting the sample size and has a mixture-normal distribution in the limit. The MLE for the scale parameter of Cauchy distribution is $n^{1/2}$ -consistent, and its limiting distribution is normal. The MLEs of the intercept and the linear time trend are $n^{1/2}$ - and $n^{3/2}$ -consistent, respectively. It is also shown that the *t* statistic for the null hypothesis of a unit root based on the MLE has a standard normal distribution in the limit. In addition, finite-sample properties of the MLE is more efficient than the LSE when the errors have a Cauchy distribution or a distribution which is a mixture of Cauchy and normal distributions. It is also shown that empirical power of the MLE-based *t* test for a unit root is much higher than that of the Dickey–Fuller *t* test.

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This paper stemmed from the MA thesis of the first author which was supervised by the second author.

Electronic supplementary material The online version of this article (https://doi.org/10.1007/s10463-018-0671-z) contains supplementary material, which is available to authorized users.

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Keywords Autoregressive model \cdot Near unit root \cdot Cauchy distribution \cdot Maximum likelihood estimator \cdot Infinite variance