

# Information Geometry of Complex Autoregressive Models and its Positive Superharmonic Priors

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It is difficult to decide what are the good features to be equipped on statistical manifolds. The Kähler structure on a statistical manifold is considered to be one of the good properties that the statistical model should have. Complex-valued stochastic processes are useful models for parametrizing complex or bivariate signals. Choi and Mullhaupt showed that the information geometry of complex signals is Kähler [2]. By making use of this structure, they showed that, for the complex autoregressive models of order 2, there is a Bayesian predictive distribution which asymptotically dominates the Bayesian predictive distribution based on the Jeffreys prior. We generalized this theorem to the complex autoregressive models of any order  $p$ .

Let us consider a stable discrete-time linear system with a complex scalar input and a complex scalar output. Let  $h = \{h_t\}$  be a complex linear operator that maps an input  $x = \{x_t\}$  to an output  $y = \{y_t\}$  defined by  $y_t = \sum_i h_i x_{t-i}$  where all  $x_t$ ,  $y_t$  and  $h_t$  are complex numbers. The output  $y$  is the convolution  $y = h * x$  of the input  $x$  and the operator  $h$ . We assume throughout this presentation that the linear system is invertible and that both the original linear system and the inversed linear system are stable and causal. We call the system real if all  $h_t$  are real numbers. The function  $H(z) := \sum_t h_t z^{-t}$  of a formal variable  $z$  is called the transfer function of the linear system, and the function  $S(\omega) := |H(e^{i\omega})|^2$  is called the power spectral density of the linear system. If the linear system is real, then  $S$  is an even function on  $[-\pi, \pi]$ . The  $\alpha$ -geometry of the statistical manifold of real autoregressive models was investigated in [1].

Let  $\text{AR}(\sigma^2; p; \mathbb{C}) := \left\{ S(\omega) = \sigma^2 \left| \frac{1}{\prod_{j=1}^p (1 - \xi^j e^{-i\omega})} \right|^2 \mid |\xi_j| < 1 \right\}$  be the set of all the complex autoregressive models of order  $p$ . Then, the statistical manifold  $\text{AR}^\circ(\sigma^2; p; \mathbb{C}) := \{S \in \text{AR}(\sigma^2; p; \mathbb{C}) \mid |g| \neq 0\}$  is a Kähler manifold, where  $g_{i\bar{j}} := \frac{1}{4\pi} \int_{-\pi}^{\pi} (\partial_i \log S(\omega)) (\partial_{\bar{j}} \log S(\omega)) d\omega$  is the Fisher information matrix of the complex-valued stochastic process. We have investigated the Bayesian predictive distribution on the Kähler manifold  $\text{AR}^\circ(\sigma^2; p; \mathbb{C})$ . We have proved that the function  $\phi = \prod_{k,l} (1 - \xi^k \bar{\xi}^l)$  is positive and superharmonic  $\Delta\phi = -2p(p+1)\phi$  on  $\text{AR}^\circ(\sigma^2; p; \mathbb{C})$ . The  $p = 2$  case has already been proved in [2]. For the real case  $\text{AR}^\circ(\sigma^2; p; \mathbb{R})$ , the similar result has also been given for  $p = 2$  in [3], and for general  $p$  in [4]. We have proved that the prior  $\pi_H := \phi \pi_J$  asymptotically dominates the Bayesian predictive distribution based on the Jeffreys prior  $\pi_J$ ,

$$E_S \left[ D_{\text{KL}} \left( S \parallel \hat{S}_{\pi_J} \right) \mid \xi \right] - E_S \left[ D_{\text{KL}} \left( S \parallel \hat{S}_{\pi_H} \right) \mid \xi \right] = \frac{1}{N^2} (2\|\xi\| + 2p(p+1) + o(1)) \quad ,$$

where  $\|\xi\| := 2g_{i\bar{j}} \xi^i \bar{\xi}^j = \sum_{i=1}^p \sum_{j=1}^p \frac{\xi^i \bar{\xi}^j}{1 - \xi^i \bar{\xi}^j}$ .

## References

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