

# Introduction of general distributions on sphere and torus in view of time series spectra

BY MASANOBU TANIGUCHI

Waseda University  
taniguchi@waseda.jp

5

AND YUJIE XUE

The Institute of Statistical Mathematics  
xue.yujie@ism.ac.jp

## SUMMARY

There are various fields where observations are taken on directions in three dimensions, e.g., sphere and torus. Here we will introduce a very general family of distributions on sphere and torus by use of time series spectra, which includes a lot of proposed classical one as special cases. Because time series spectra can be described by a lot of famous parametric models, e.g., AR, ARMA etc., we can develop the systematic model selection in this field by use of AIC, BIC, etc. Applications are very wide.

10

15

*Some key words:* Distributions on sphere and torus; Time series spectra; AR, ARMA models; Model selection.

## 1. INTRODUCTION

There are various fields where observations are taken on directions in three dimensions, e.g., molecular biology and physics. The von Mises-Fisher distribution is very fundamental with probability density function

20

$$f(\mathbf{x}) = c(\kappa, \boldsymbol{\mu})^{-1} \exp[\kappa \boldsymbol{\mu}^T \mathbf{x}], \quad \mathbf{x} \in \mathbb{S}^2, \quad (1.1)$$

where  $\mathbb{S}^2 = \{\mathbf{x} \in \mathbb{R}^3 : \|\mathbf{x}\| = 1\}$ ,  $\boldsymbol{\mu} \in \mathbb{S}^2$  and  $c(\cdot)$  is the normalizing constant (e.g., Mardia and Jupp (2000)). The Fisher-Bingham distribution is given by density

$$f(\mathbf{x}) = c(\kappa, \boldsymbol{\mu}, \mathbf{A})^{-1} \exp[\kappa \boldsymbol{\mu}^T \mathbf{x} + \mathbf{x}^T \mathbf{A} \mathbf{x}], \quad (1.2)$$

where  $\mathbf{A}$  is a symmetric  $3 \times 3$  matrix (Mardia (1975)). For a special choice of  $\boldsymbol{\mu}$  and  $\mathbf{A}$ , Kent (1982) derived a polar co-ordinates form of (1.2) by

$$f(\theta, \phi) = c(\kappa, \beta)^{-1} \exp[\kappa \cos \theta + \beta \sin^2 \theta \cos 2\phi]. \quad (1.3)$$

Kato and McCullagh (2020) introduced a Cauchy family of distribution by density

25

$$f(\mathbf{x}) = c(\rho, \boldsymbol{\mu})^{-1} \left[ \frac{1 - \rho^2}{1 + \rho^2 - 2\rho \boldsymbol{\mu}^T \mathbf{x}} \right], \quad \mathbf{x} \in \mathbb{S}^2. \quad (1.4)$$

For modelling of torsional angles of molecules, Singh et al. (2002) introduced the following distribution on torus by

$$f(\theta, \phi) = c \exp[\kappa_1 \cos(\theta - \mu_1) + \kappa_2 \cos(\phi - \mu_2) + \lambda \sin(\theta - \mu_1) \sin(\phi - \mu_2)], \quad (1.5)$$

where  $-\pi \leq \theta, \phi \leq \pi$ ,  $\kappa_1, \kappa_2 \geq 0$ ,  $-\infty < \lambda < \infty$ ,  $-\pi \leq \mu_1, \mu_2 \leq \pi$  and  $c$  is a normalization constant. Also, Kato and Pewsey (2015) introduced the following wrapped Cauchy type distribution by density

$$f(\theta, \phi) = c[c_0 - c_1 \cos(\theta - \mu_1) - c_2 \cos(\phi - \mu_2) - c_3 \cos(\theta - \mu_1) \cos(\phi - \mu_2) - c_4 \sin(\theta - \mu_1) \sin(\phi - \mu_2)]^{-1}, \quad (-\pi \leq \theta, \phi \leq \pi), \quad (1.6)$$

where  $c_i$  and  $\mu_i$  are constants.

For circular data, the wrapped Cauchy density is often used, and is defined by

$$f_W(\omega) = c \frac{1}{1 + \rho^2 + 2\rho \cos \omega}, \quad \omega \in [-\pi, \pi]. \quad (1.7)$$

Time series people understand that this is exactly the AR(1) spectral density

$$f_S(\omega) = c \frac{1}{|1 + \rho e^{i\omega}|^2}, \quad \omega \in [-\pi, \pi]. \quad (1.8)$$

Motivated by this, Taniguchi et al. (2020) introduced a very general family of joint circular distributions by a higher order spectral density

$$f_S(\omega_1, \omega_2, \dots, \omega_n), \quad \omega_k \in [-\pi, \pi], \quad (1.9)$$

which can be decomposed to

$$\prod_{k=1}^n f_S(\omega_k), \quad (1.10)$$

if  $\omega_1, \omega_2, \dots, \omega_n$  are independent, where  $f_S(\omega_k)$  is the spectral density of frequency  $\omega_k$ . Advantage of this approach is that we can introduce time series models for  $f_S(\omega_k)$ , i.e., AR( $p$ ), ARMA( $p, q$ ) models etc., then the systematic model selection in this field can be carried out.

In this paper, for distributions on sphere and torus, in view of above, we will introduce a very general family of the distributions by time series spectra, whose forms will be

$$f(\theta, \phi) = \sum_{k=-\infty}^{\infty} \{A_k e^{ik(\theta+\phi)} + B_k e^{ik(\theta-\phi)} + C_k e^{ik\theta}\}$$

for  $(\theta, \phi) \in \mathbb{S}^2$ .  $f(\theta, \phi)$  is the sum of time series spectra, hence, we may use ARMA( $p, q$ ) modelling etc. We can develop the model selection by use of AIC and BIC etc. Discussion on torus is similar. The applications are in various fields.

## 2. DISTRIBUTIONS ON SPHERE

In this section we introduce a very general distribution on sphere in view of time series spectra. Let the 3-dim polar representation be given by

$$\begin{cases} x &= \sin \theta \cos \phi; \\ y &= \sin \theta \sin \phi; \\ z &= \cos \theta. \end{cases} \quad (2.1)$$

Substituting  $\begin{cases} \cos \theta &= \frac{e^{i\theta} + e^{-i\theta}}{2} \\ \sin \theta &= \frac{e^{i\theta} - e^{-i\theta}}{2i} \end{cases}$  to (2.1), we obtain

$$\begin{cases} x &= \frac{e^{i\theta} - e^{-i\theta}}{2i} \times \frac{e^{i\phi} + e^{-i\phi}}{2} = \frac{e^{i(\theta+\phi)} - e^{-i(\theta+\phi)} + e^{i(\theta-\phi)} - e^{-i(\theta-\phi)}}{4i}; \\ y &= \frac{e^{i\theta} - e^{-i\theta}}{2i} \times \frac{e^{i\phi} - e^{-i\phi}}{2i} = \frac{e^{i(\theta+\phi)} + e^{-i(\theta+\phi)} - e^{i(\theta-\phi)} - e^{-i(\theta-\phi)}}{-4}; \\ z &= \frac{e^{i\theta} + e^{-i\theta}}{2}. \end{cases} \quad (2.2)$$

$$\quad (2.3)$$

$$\quad (2.4)$$

Let  $\delta_k = \begin{cases} \frac{1}{4i}, & k = 1 \\ -\frac{1}{4i}, & k = -1 \end{cases}$ ,  $\gamma_k = \begin{cases} -\frac{1}{4}, & k = 1 \\ -\frac{1}{4}, & k = -1 \end{cases}$ , and  $\eta_k = \begin{cases} \frac{1}{2}, & k = 1 \\ \frac{1}{2}, & k = -1 \end{cases}$ . Then we can see that

$$x = \sum_{k=\pm 1} \delta_k \{e^{ik(\phi+\theta)} + e^{ik(\phi-\theta)}\}; \quad (2.5)$$

$$y = \sum_{k=\pm 1} \gamma_k \{e^{ik(\phi+\theta)} - e^{ik(\phi-\theta)}\}; \quad (2.6)$$

$$z = \sum_{k=\pm 1} \eta_k e^{ik\theta}. \quad (2.7)$$

Next we introduce general Fourier coefficients:

- (i) pure imaginary coefficient  $a_k = -\bar{a}_k$  satisfying  $a_{-k} = -a_k$ ;
- (ii) real coefficients  $b_k, k \in \mathbb{Z}, b_{-k} = b_k$ ;
- (iii) real coefficients  $c_k, k \in \mathbb{Z}, c_{-k} = c_k$ .

It is seen that

$$\left[ \sum_{k=\pm 1} a_k \{e^{ik(\phi+\theta)} + e^{ik(\phi-\theta)}\} + \sum_{k=\pm 1} b_k \{e^{ik(\phi+\theta)} - e^{ik(\phi-\theta)}\} + \sum_{k=\pm 1} c_k e^{ik\theta} \right] \sin \theta \quad (2.8)$$

corresponds to the part  $\boldsymbol{\mu}^T \mathbf{x}$  in the von Mises-Fisher distribution (1.1) and Kato-McCullagh distribution (1.4). Here  $\sin \theta$  is the Jacobian. Because  $e^{ik(\phi+\theta)}$ ,  $e^{ik(\phi-\theta)}$  and  $e^{ik\theta}$  are Fourier basis at frequency  $\phi + \theta$ ,  $\phi - \theta$  and  $\theta$ , it is natural to think of the strength of these frequencies by the spectral densities

$$f_S(\theta, \phi) = \left[ \sum_{k=-\infty}^{\infty} a_k \{e^{ik(\phi+\theta)} + e^{ik(\phi-\theta)}\} + \sum_{k=-\infty}^{\infty} b_k \{e^{ik(\phi+\theta)} - e^{ik(\phi-\theta)}\} + \sum_{k=-\infty}^{\infty} c_k e^{ik\theta} \right] \sin \theta. \quad (2.9)$$

Mardia (1975) introduced the Fisher-Bingham distribution (1.2) which includes  $\kappa \boldsymbol{\mu}^T \mathbf{x} + \mathbf{x}^T \mathbf{A} \mathbf{x}$ . Hence, in (2.5) and (2.6), if we consider  $x^2 - y^2$ , it is not difficult to see that (2.9) becomes

$$[f_S(\theta, \phi) + \sum_{k=-\infty}^{\infty} d_k e^{ik\phi}] \sin \theta. \quad (2.10)$$

Therefore we can introduce our very general family of distributions on sphere as follows. For a smooth function  $H[\cdot]$ , we propose

$$F_S(\theta, \phi) = H \left[ \sum_{k=-\infty}^{\infty} (A_k e^{ik(\theta+\phi)} + B_k e^{ik(\theta-\phi)} + C_k e^{ik\theta} + D_k e^{ik\phi}) \right] \sin \theta, \quad (2.11)$$

as a probability density on sphere, where  $A_k, B_k, C_k$  and  $D_k$  are complex-valued, and  $F_S(\theta, \phi) \geq 0$  so that  $\int F_S(\theta, \phi) d\mathbb{P} = 1$ .

### 3. DISTRIBUTIONS ON TORUS

In this section we introduce a very general distribution on torus  $\mathbb{T}$  in view of time series  
75 spectra. Let the 3-dim polar representation be given by

$$\begin{cases} x &= R \cos \theta + r \cos \phi \cos \theta \\ y &= R \sin \theta + r \cos \phi \sin \theta, \\ z &= r \sin \phi \end{cases} \quad (0 \leq \theta, \phi \leq 2\pi). \quad (3.1)$$

Write  $\begin{cases} \cos \theta &= \frac{1}{2}(e^{i\theta} + e^{-i\theta}) \\ \sin \theta &= \frac{1}{2i}(e^{i\theta} - e^{-i\theta}) \end{cases}$ , then we have

$$\begin{aligned} x &= R \frac{1}{2}(e^{i\theta} + e^{-i\theta}) + r \frac{1}{4}(e^{i\phi} + e^{-i\phi})(e^{i\theta} + e^{-i\theta}) \\ &= R \frac{1}{2}(e^{i\theta} + e^{-i\theta}) + r \frac{1}{4}(e^{i(\theta+\phi)} + e^{-i(\theta+\phi)} + e^{-i(\phi-\theta)} + e^{i(\phi-\theta)}); \end{aligned} \quad (3.2)$$

$$\begin{aligned} y &= R \frac{1}{2i}(e^{i\theta} - e^{-i\theta}) + r \frac{1}{4i}(e^{i\phi} + e^{-i\phi})(e^{i\theta} - e^{-i\theta}) \\ &= R \frac{1}{2i}(e^{i\theta} - e^{-i\theta}) + r \frac{1}{4i}(e^{i(\theta+\phi)} - e^{-i(\theta+\phi)} + e^{-i(\phi-\theta)} - e^{i(\phi-\theta)}); \end{aligned} \quad (3.3)$$

$$z = r \frac{1}{2i}(e^{i\phi} - e^{-i\phi}). \quad (3.4)$$

Similarly as in Section 2, we propose a very general family of distributions on  $\mathbb{T}$  by

$$g(\theta, \phi) \propto \left[ \sum_{k=-\infty}^{\infty} a_k e^{ik\theta} + \sum_{k=-\infty}^{\infty} b_k e^{ik\phi} + \sum_{k=-\infty}^{\infty} c_k e^{ik(\theta+\phi)} + \sum_{k=-\infty}^{\infty} d_k e^{ik(\theta-\phi)} \right] |J|. \quad (3.5)$$

$\{a_k\}, \{b_k\}, \{c_k\}, \{d_k\}$  are complex-valued coefficients, and are chosen so that  $g(\theta, \phi)$  is real-  
85 valued. Here  $|J| = r(r \cos \phi + R)$ , the Jacobian.

Singh et al. (2002) introduced the distribution on  $\mathbb{T}$  by (1.5). We can see that the exponent is of our form (3.5). Also Kato and Pewsey (2015) introduced the distribution on  $\mathbb{T}$  by (1.6). It is seen that the inside of the reverse function is of our form (3.5).

Hence we introduce our very general family of distributions on  $\mathbb{T}$  as follows.

90 For a smooth function  $G[\cdot]$ , we propose

$$F_T(\theta, \phi) \equiv G \left[ \sum_{k=-\infty}^{\infty} (A'_k e^{ik(\theta+\phi)} + B'_k e^{ik(\theta-\phi)} + C'_k e^{ik\theta} + D'_k e^{ik\phi}) \right] |J|, \quad (3.6)$$

as a probability density on  $\mathbb{T}$ , where  $A'_k, B'_k, C'_k$  and  $D'_k$  are complex-valued, and  $F_T(\theta, \phi) \geq 0$  so that  $\int F_T(\theta, \phi) d\mathbb{P} = 1$ .

### 4. SUMMARY AND CONCLUDING REMARKS

We could introduce a very general distributions on  $\mathbb{S}^2$  and  $\mathbb{T}$  by time series spectra (2.11)  
95 and (3.6) respectively. The advantage is that we can develop the problem of model selection

systematically because the time series spectra have a lot of famous finite parametric models, e.g., AR, MA, and ARMA models, i.e., for modelling, we can use

$$F(\theta, \phi) = H[s_1 f_{\text{ARMA}}^{(1)}(\theta + \phi) + s_2 f_{\text{ARMA}}^{(2)}(\theta - \phi) + s_3 f_{\text{ARMA}}^{(3)}(\theta) + s_4 f_{\text{ARMA}}^{(4)}(\phi)],$$

where  $s_1, \dots, s_4$  are real constants and  $f_{\text{ARMA}}^{(j)}$  are ARMA spectral densities (e.g., Taniguchi and Kakizawa (2000)). Hence we can use AIC, BIC etc to select the model, which enriches applications for data from  $\mathbb{S}^2$  and  $\mathbb{T}$ . Also, the systematic asymptotic estimation theory will be possible (e.g., Taniguchi and Kakizawa (2000)).

#### ACKNOWLEDGEMENT

The authors thank Professor Shogo Kato for his instructive comments.

#### REFERENCES

- KATO, S. & PEWSEY, A. (2015). A Möbius transformation–introduced distribution on the torus. *Biometrika* **102**(2), 359–370. 105
- KATO, S. & MCCULLAGH, P. (2020). Some properties of a Cauchy family on the sphere derived from the Möbius transformations. *Bernoulli* **26**(4), 3224–3248.
- KENT, J. T. (1982). The Fisher-Bingham distribution on the sphere. *J. R. Statist. Soc. B* **44**(1), 71–80.
- MARDIA, K. V. (1975). Statistics of directional data (with discussion) *J. R. Statist. Soc. B* **37**, 349–393. 110
- MARDIA, K. V. & JUPP, P. E. (2000). *Directional Statistics*. New York: Wiley.
- SINGH, H., HINZDO, V. & DEMCHUK, E. (2002). Probablistic model for two dependent circular variables. *Biometrika* **89**(3), 719–723.
- TANIGUCHI, M. & KAKIZAWA, Y.(2000). *Asymptotic Theory of Statistical Inference for Time Series*. New York: Springer. 115
- TANIGUCHI, M., KATO, S., OGATA, H. & PEWSEY, A. (2020). Models for circular data from time series spectra. *J. Time Ser. Anal.* **41**, 808–829.