

# Polar angle distributions on generalized circles

## Representations and generalizations of the von Mises distribution

Wolf-Dieter Richter

University of Rostock, Institute of Mathematics

Advances in Directional Statistics  
International workshop  
Brussels, May 22, 2014

# Content

- 1 Introduction
  - The conditional offset approach
  - Generalized circles / generalized conditioning radius variables
  - Directional vector component
- 2 Representations of the von Mises distribution: Joint work with T.Dietrich
- 3 Generalizations of the von Mises distribution
  - General representation
  - Classes of generalized vMds: Jointly with T. Dietrich
- 4 Summary and outlook

# From physical problems to circular distributions.

Some of the basic references

or... bringing Atomium to Brussels.

Langevin(1905)

von Mises(1918)

Gumbel, Greenwood and Durand(1953)

Fisher(1959)

Mardia(1972)

Batschelet(1981)

Fisher, Lewis and Embleton(1987)

Mardia and Jupp(2000)

Jammalamadaka and SenGupta(2001)

Pewsey, Neuhäuser and Ruxton(2013)

# The conditional offset approach

Mardia(1972)

by a construction following arguments in Fisher(1959)

Jammalamadaka and SenGupta(2001)

Mardia and Jupp(2000)

Shimizu and Iida(2002)

Jones and Pewsey(2005)

Gatto and Jammalamadaka(2007)

## ... the conditional offset approach

Conditional distribution of the directional component of a random vector given its fixed distance from the origin.

Distance from the origin

= radius of the circle where the vector belongs to

= value of the vector's radius variable.

# Generalized circles / radius variables, 1

Power exponential distribution:

$$f_{(X,Y)}(x,y) = C_p \exp\left\{-\frac{|x|^p}{p} - \frac{|y|^p}{p}\right\} = C_p \exp\left\{-\frac{\|(x,y)\|_p^p}{p}\right\}$$

$p = 1$  ... Laplace distribution,  $p = 2$  ... Gaussian distribution

$$\|(x,y)\|_p = (|x|^p + |y|^p)^{1/p} \text{ p-norm if } p \geq 1$$

$$\frac{(X,Y)}{\|(X,Y)\|_p} : \Omega \rightarrow \{(x,y) \in \mathbb{R}^2 : |x|^p + |y|^p = 1\}$$

**p-circle**

**Generalized radius variable** of  $(X, Y)$ :  $R = \|(X, Y)\|_p$ .

# Generalized circles, 2

Norm contoured distribution:

$$f_{(X,Y)}(x,y) = C(g, \|\cdot\|)g(\|(x,y)\|), (x,y) \in R^2$$

$$\|\cdot\| : R^2 \rightarrow [0, \infty) \text{ arbitrary norm}$$

$$\frac{(X,Y)}{\|(X,Y)\|_\rho} : \Omega \rightarrow \{(x,y) \in R^2 : \|(x,y)\| = 1\}$$

**norm-circle**

**Generalized radius variable** of  $(X, Y)$ :  $R = \|(X, Y)\|$ .

# Generalized radius variables, 3

$K \subset \mathbb{R}^2$  star body having the origin as an inner point

$$h_K((x, y)) = \inf\{\lambda > 0 : (x, y) \in \lambda K\}$$

$r \cdot K = \{(x, y) \in \mathbb{R}^2 : h_K((x, y)) \leq r\} =: K(r)$  star ball of star radius  $r$

Topological boundary of  $K(r)$

$$= \{(x, y) \in \mathbb{R}^2 : h_K((x, y)) = r\} = S(r)$$

**star circle of star radius  $r$ ,  $S = S(1)$ .**

**Generalized radius variable of  $(X, Y)$ :  $R = h_K((X, Y))$ .**



# Generalized radius variable, 4

**Example**  $K$  is a convex body, symmetric w.r.t. the origin  
 iff  
 $h_K$  is a norm.

**Example**  $K$  elliptical disc ...

... circumscribed by  $E_{(a,b)}$ ... axes-aligned ellipse:

$$E_{(a,b)} = \{(x, y) \in \mathbb{R}^2 : (\frac{x}{a})^2 + (\frac{y}{b})^2 = 1\}, 0 < b \leq a$$

heteroscedastic Gaussian or elliptically contoured distribution

$$R = h_K((X, Y)) = \left( \left( \frac{X}{a} \right)^2 + \left( \frac{Y}{b} \right)^2 \right)^{1/2}$$

# Adapted directional vector variable, 1

$E_{(a,b)}$ -generalized trigonometric functions:

$$\cos_{(a,b)}(\phi) = \frac{\cos \phi}{aN_{(a,b)}(\phi)}, \quad \sin_{(a,b)}(\phi) = \frac{\sin \phi}{bN_{(a,b)}(\phi)},$$

$$N_{(a,b)}(\phi) = \left( \left( \frac{\cos \phi}{a} \right)^2 + \left( \frac{\sin \phi}{b} \right)^2 \right)^{1/2}.$$

$$E_{(a,b)} = \left\{ \begin{pmatrix} a \cos_{(a,b)}(\phi) \\ b \sin_{(a,b)}(\phi) \end{pmatrix}, 0 \leq \phi < 2\pi \right\}$$

## Adapted directional vector variable, 2

$E_{(a,b)}$ -generalized elliptical coordinates  $R, \Phi$ :

$$X = Ra \cos_{(a,b)}(\Phi), Y = Rb \sin_{(a,b)}(\Phi)$$

Inverse transformation:

$R = h_{K_{(a,b)}}((X, Y)) = ((X/a)^2 + (Y/b)^2)^{1/2}$ , elliptical radius coordinate

$$K_{(a,b)} = \{(x, y) \in \mathbb{R}^2 : (\frac{x}{a})^2 + (\frac{y}{b})^2 \leq 1\}, 0 < b \leq a$$

axes-aligned elliptical disc

$$\tan \Phi = \frac{Y}{X}, \quad \Phi = \text{polar angle coordinate}$$

# Using heteroscedastic, uncorrelated Gaussian vector

$$(X, Y) \sim \Phi \left( \begin{array}{c} a \cdot \cos_{(a,b)}(\mu) \\ b \cdot \sin_{(a,b)}(\mu) \end{array} \right), \frac{1}{\delta} \left( \begin{array}{cc} a^2 & 0 \\ 0 & b^2 \end{array} \right) \text{ for some } \delta > 0, \lambda > 0$$

$\Phi$ ... polar angle coordinate of  $(X, Y)$ .

$$\text{Elliptical polar angle: } T(\varphi) = \begin{cases} \arccos[\cos_{(a,b)}(\varphi)] & \text{if } 0 \leq \varphi < \pi \\ 2\pi - \arccos[\cos_{(a,b)}(\varphi)] & \pi \leq \varphi < 2\pi \end{cases}$$

**Representation 1:**  $f_{T(\Phi)|R}(\psi|r) = vM d_{\kappa, T(\mu)}(\psi), \quad \kappa = \lambda\delta r$

# Using a regular Gaussian vector, 1

$$(X, Y) \sim \Phi_{\nu, \frac{1}{\delta}\Sigma}, \Sigma = \begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix}$$

$$\alpha = \alpha(\Sigma) = [1 - I_{\{0\}}(\rho)]$$

$$\times \left[ \frac{\pi}{4} I_{\{\sigma_2\}}(\sigma_1) + (1 - I_{\{\sigma_2\}}(\sigma_1)) \left( \frac{1}{2} \arctan \frac{2\sigma_1\sigma_2\rho}{\sigma_1^2 - \sigma_2^2} + \frac{\pi}{2} I_{(-\infty, 0)}(\rho(\sigma_1 - \sigma_2)) \right) \right]$$

$$a = a(\Sigma) = \sqrt{\sigma_1^2 \cos^2 \alpha + \sigma_2^2 \sin^2 \alpha + 2\rho\sigma_1\sigma_2 \sin \alpha \cos \alpha},$$

$$b = b(\Sigma) = \sqrt{\sigma_2^2 \cos^2 \alpha + \sigma_1^2 \sin^2 \alpha - 2\rho\sigma_1\sigma_2 \sin \alpha \cos \alpha}.$$

## Using a regular Gaussian vector, 2

$$X = \frac{R \cos \Phi}{N_{(a,b)}(\Phi - \alpha)}, \quad Y = \frac{R \sin \Phi}{N_{(a,b)}(\Phi - \alpha)}$$

$$\Phi = \tan \frac{Y}{X} \dots \text{polar angle}$$

$$R = h_{K_{a,b}}((X, Y)) = \|(X, Y)\|_2 N_{(a,b)}(\Phi - \alpha)$$

**Representation 2:**  $f_{T(\Phi-\alpha)|R}(\psi|r) = \nu M d_{\kappa, T(\mu-\alpha)}$ ,  $\kappa = \lambda \delta r$

where  $\mu \in [-\pi, \pi)$ ,  $\lambda > 0$  are uniquely determined by

$$\nu = \frac{\lambda}{N_{(a,b)}(\mu - \alpha)} \begin{pmatrix} \cos \mu \\ \sin \mu \end{pmatrix}$$

# Using an elliptically contoured distributed vector

$g : [0, \infty) \rightarrow [0, \infty)$  ... density generating function

$$0 < \int_0^{\infty} rg(r)dr < \infty \quad g\text{-generalized von}$$

Mises density (Jones and Pewsey (2005)):

$$vMd_{g,\lambda,r,\theta}(\psi) = Cg((\lambda^2 + r^2 - 2\lambda r \cos(\psi - \theta))^{1/2}), \lambda \geq r$$

$$(X, Y) \sim \Phi_{g,\nu,\Theta}$$

$$\phi_{g,\nu,\Theta}(x, y) = D(g) \det \Theta^{-1/2} g(((x-\nu_1, y-\nu_2)\Theta^{-1}(x-\nu_1, y-\nu_2)^T)^{1/2})$$

**Representation 3:**  $f_{T(\Phi-\alpha)|R}(\varphi|r) = vMd_{g,\lambda,r,T(\mu-\alpha)}(\psi)$

$\alpha, \mu, \lambda$  as before

# Common angular and $K$ -generalized radius variable

Star-shaped vector distribution:

$$(X, Y) \sim \varphi_{g,K,\nu}(x, y) = C(g, K)g\left(h_K\left(\begin{pmatrix} x \\ y \end{pmatrix} - \begin{pmatrix} \nu_1 \\ \nu_2 \end{pmatrix}\right)\right)$$

General representation:  $f_{\Phi|R}(\phi|r) = \text{vMd}_{g,K,r,\lambda,\mu}(\phi, r)$

$\lambda, \mu$  are uniquely determined by  $\nu = \lambda \begin{pmatrix} \cos_K(\mu) \\ \sin_K(\mu) \end{pmatrix}$

$$\text{vMd}_{g,K,r,\lambda,\mu}(\varphi) = D(g, K)R_S^2(\varphi)g\left(h_K\left(\begin{pmatrix} r \cos_K(\varphi) - \lambda \cos_K(\mu) \\ r \sin_K(\varphi) - \lambda \sin_K(\mu) \end{pmatrix}\right)\right)$$

$\{R_S(\varphi) \begin{pmatrix} \cos_K(\varphi) \\ \sin_K(\varphi) \end{pmatrix}\} = S \dots \text{unit star circle}$



# Elliptically contoured Gaussian generalized vMd, 1

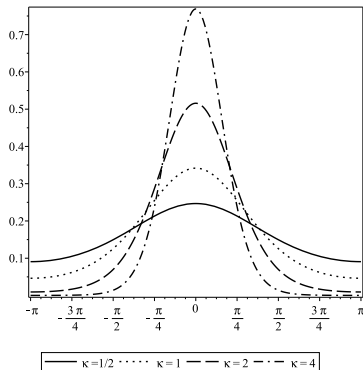


Fig. : unimodal, symmetric

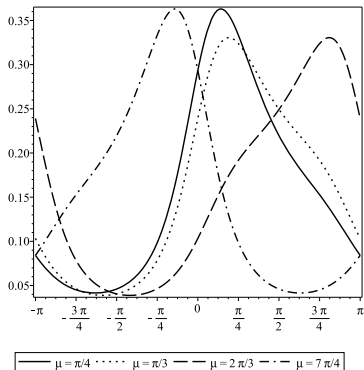


Fig. : unimodal, asymmetric

# Elliptically contoured Gaussian generalized vMd, 2

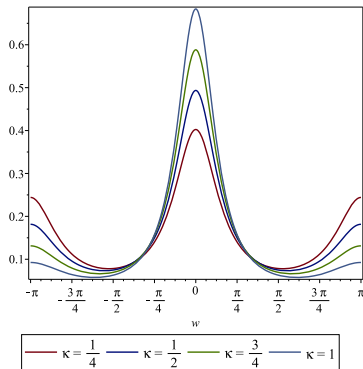


Fig. : bimodal, symmetric

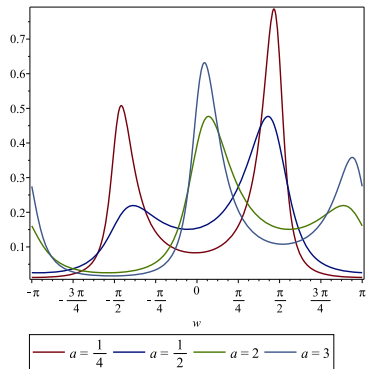


Fig. : bimodal, asymmetric

# Elliptically contoured generalized vMd, Kotz type dgf

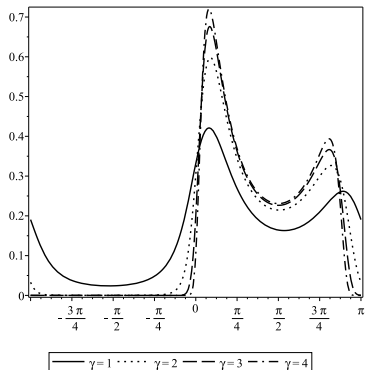


Fig. : bimodal

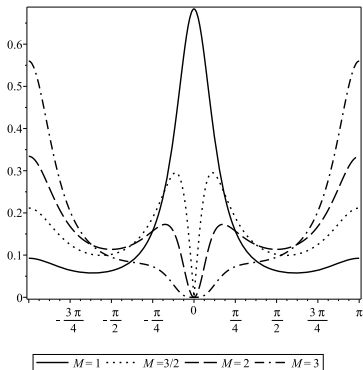


Fig. : multimodal

# Elliptically contoured gen. vMd, Pearson type VII dgf, 1

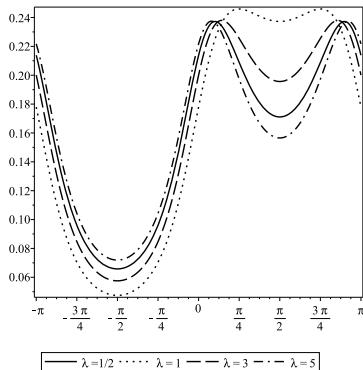


Fig. : bimodal, shift-symmetric

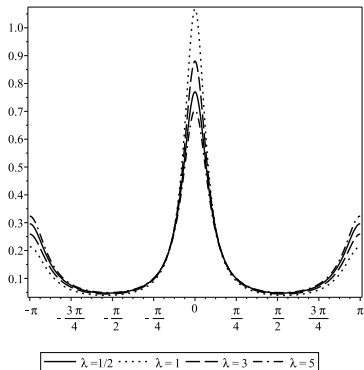


Fig. : bimodal, symmetric

# Elliptically contoured gen. vMd, Pearson type VII dgf, 2

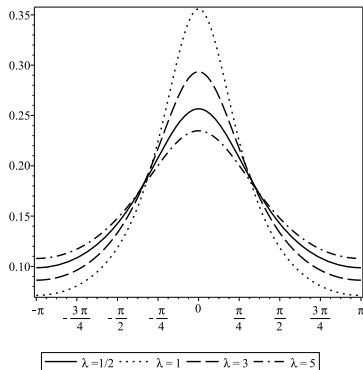


Fig. : unimodal

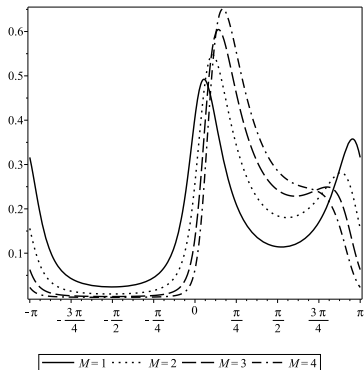


Fig. : bimodal, asymmetric

# Polygonally contoured generalized vMd

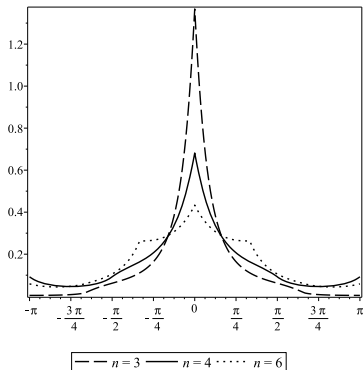


Fig. :  $g = g_G$

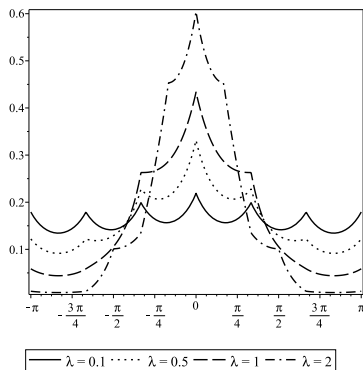


Fig. :  $g = g_G, n = 6$

# Non-concentric elliptically contoured generalized vMd

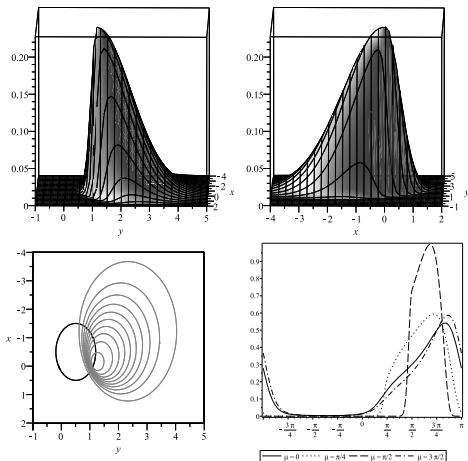


Fig. :  $e = 0.5, f = -0.5, \mu = \pi/2, g = g_G$

# The $p$ -generalized elliptically contoured vMd, 1

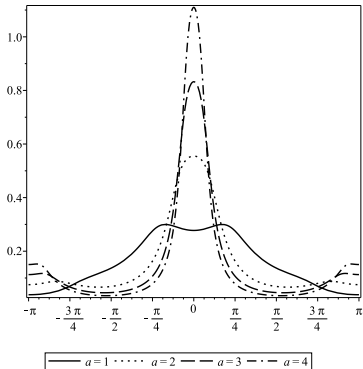


Fig. :

$b = 1, p = 4, \lambda = r = 1, \mu = 0$

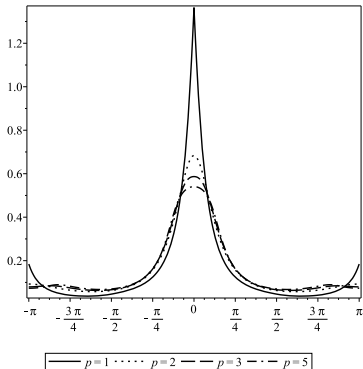


Fig. :

$a = 2, b = 1 = \lambda = r, \mu = 0$



# The $\rho$ -generalized elliptically contoured vMd, 2

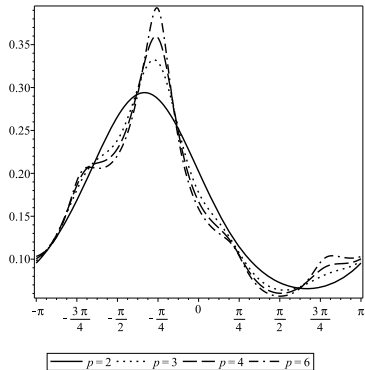


Fig. :  $a = b = 1, \lambda = 3/4, r = 1, \mu = 5\pi/3$

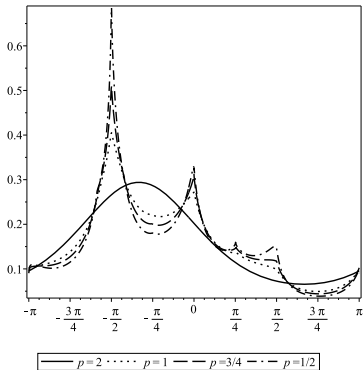


Fig. :  $a = b = 1, \lambda = 3/4, r = 1, \mu = 5\pi/3$

# Summary

- New **representations** of the von Mises distribution. The interpretation of the angle, however, is involved.
- New **generalizations** of the von Mises distribution. Asymmetry and multimodality are possible.
- The conditioning **radius variable** is constant on norm level sets or star-shaped circles.
- Outlook
  - Multivariate generalizations.
  - Statistics.