APPENDIX

## Directional Statistics with the Spherical Normal Distribution — Supplementary Material —

Søren Hauberg Department of Mathematics and Computer Science Technical University of Denmark Kgs. Lyngby, Denmark sohau@dtu.dk

The isotropic spherical normal distribution has density

$$\mathsf{SN}(\mathbf{x} \mid \boldsymbol{\mu}, \lambda) = \frac{1}{\mathcal{Z}_1(\lambda)} \exp\left(-\frac{\lambda}{2} \mathrm{Log}_{\boldsymbol{\mu}}(\mathbf{x})^{\mathsf{T}} \mathrm{Log}_{\boldsymbol{\mu}}(\mathbf{x})\right), \qquad \|\boldsymbol{\mu}\| = 1, \quad \lambda > 0.$$
(1)

In the main paper, we claim that the normalization constant is

$$\mathcal{Z}_{1}^{(\text{even})}(\lambda) = \frac{A_{D-2}}{2^{D-2}} {D-2 \choose D/2-1} \sqrt{\frac{\pi}{2\lambda}} \operatorname{erf}\left(\pi\sqrt{\frac{\lambda}{2}}\right) + \frac{A_{D-2}}{2} \sqrt{\frac{2\pi}{\lambda}} \frac{(-1)^{D/2-1}}{2^{D-3}}$$

$$\cdot \sum_{k=0}^{D/2-2} (-1)^{k} {D-2 \choose k} \exp\left(-\frac{(D-2-2k)^{2}}{2\lambda}\right) \operatorname{Re}\left[\operatorname{erf}\left(\frac{\pi\lambda - (D-2-2k)i}{\sqrt{2\lambda}}\right)\right]$$
(2a)

$$\mathcal{Z}_{1}^{(\text{odd})}(\lambda) = A_{D-2} \frac{(-1)^{(D-3)/2}}{2^{D-3}} \sqrt{\frac{\pi}{2\lambda}} \sum_{k=0}^{(D-3)/2} (-1)^{k} {D-2 \choose k} \exp\left(-\frac{(D-2-2k)^{2}}{2\lambda}\right) \\ \cdot \left\{ \operatorname{Im}\left[ \operatorname{erf}\left(\frac{(D-2-2k)i}{\sqrt{2\lambda}}\right) \right] + \operatorname{Im}\left[ \operatorname{erf}\left(\frac{\pi\lambda - (D-2-2k)i}{\sqrt{2\lambda}}\right) \right] \right\}$$
(2b)

when D is even and odd. Here erf is the imaginary error function, while  $\operatorname{Re}[\cdot]$  and  $\operatorname{Im}[\cdot]$  takes the real and imaginary parts of a complex number, respectively. In this section, we provide the derivation of this constant.

By definition, we have

$$\mathcal{Z}_{1}(\lambda) = \int_{\mathcal{S}^{D-1}} \exp\left(-\frac{\lambda}{2} \mathrm{Log}_{\mu}(\mathbf{x})^{\mathsf{T}} \mathrm{Log}_{\mu}(\mathbf{x})\right) \mathrm{d}\mathbf{x}.$$
 (3)

We express this integral in the tangent space of the sphere at  $\mu$ , i.e. we perform the substitution

$$\mathbf{v} = \mathrm{Log}_{\boldsymbol{\mu}}(\mathbf{x}). \tag{4}$$

The appropriate Jacobian is

$$\det(\mathbf{J}) = \left(\frac{\sin(\|\mathbf{v}\|)}{\|\mathbf{v}\|}\right)^{D-2}$$
(5)

and the integral becomes

$$\mathcal{Z}_{1}(\lambda) = \int_{\|\mathbf{v}\| < \pi} \exp\left(-\frac{\lambda \|\mathbf{v}\|^{2}}{2}\right) \left(\frac{\sin(\|\mathbf{v}\|)}{\|\mathbf{v}\|}\right)^{D-2} \mathrm{d}\mathbf{v}.$$
 (6)

This project has received funding from the European Research Council (ERC) under the European Union's Horizon 2020 research and innovation programme (grant agreement n° 757360). SH was supported by a research grant (15334) from VILLUM FONDEN.

We then write this in hyper-spherical coordinates<sup>1</sup>

$$\mathcal{Z}_{1}(\lambda) = \int_{r=0}^{\pi} \int_{\phi_{1}=0}^{\pi} \int_{\phi_{2}=0}^{\pi} \dots \int_{\phi_{D-3}=0}^{\pi} \int_{\phi_{D-2}=0}^{2\pi} \exp\left(-\frac{\lambda r^{2}}{2}\right) \left(\frac{\sin(r)}{r}\right)^{D-2} r^{D-2} \\ \sin(\phi_{1})^{D-3} \sin(\phi_{2})^{D-4} \dots \sin(\phi_{D-3}) \mathrm{d}\phi_{D-2} \mathrm{d}\phi_{D-3} \dots \mathrm{d}\phi_{2} \mathrm{d}\phi_{1} \mathrm{d}r.$$
(7)

Since the radius r and the angles  $\phi$  are never mixed in the integrand, we can split this into the product of two integrals

$$\mathcal{Z}_1(\lambda) = \int_{r=0}^{\pi} \exp\left(-\frac{\lambda r^2}{2}\right) \sin^{D-2}(r) \mathrm{d}r$$
(8)

$$\int_{\phi_1=0}^{\pi} \int_{\phi_2=0}^{\pi} \cdots \int_{\phi_{D-3}=0}^{\pi} \int_{\phi_{D-2}=0}^{2\pi} \sin(\phi_1)^{D-3} \sin(\phi_2)^{D-4} \cdots \sin(\phi_{D-3}) \mathrm{d}\phi_{D-2} \mathrm{d}\phi_{D-3} \cdots \mathrm{d}\phi_2 \mathrm{d}\phi_1.$$

The second term is merely the surface area of the D-2 dimensional unit sphere

$$A_{D-2} = \frac{2\pi^{(D-1)/2}}{\Gamma\left(\frac{(D-1)/2}{2}\right)},\tag{9}$$

where  $\Gamma$  is the usual Gamma function. The integral (9) then reduces to

$$\mathcal{Z}_{1}(\lambda) = A_{D-2} \int_{r=0}^{\pi} \exp\left(-\frac{\lambda r^{2}}{2}\right) \sin^{D-2}(r) \mathrm{d}r.$$
 (10)

To evaluate this expression we need the trigonometric power formulas [2]

$$\sin^{2n}(x) = \frac{1}{2^{2n}} \binom{2n}{n} + \frac{(-1)^n}{2^{2n-1}} \sum_{k=0}^{n-1} (-1)^k \binom{2n}{k} \cos(2(n-k)x), \tag{11}$$

$$\sin^{2n+1}(x) = \frac{(-1)^n}{4^n} \sum_{k=0}^n (-1)^k \binom{2n+1}{k} \sin((2n+1-2k)x).$$
(12)

We are now ready evaluate the normalization constant. First we consider the case were D-2 is even.

$$\mathcal{Z}_{1}^{(\text{even})}(\lambda) = A_{D-2} \int_{r=0}^{\pi} \exp\left(-\frac{\lambda r^{2}}{2}\right) \sin^{D-2}(r) \mathrm{d}r$$

$$A_{D-2} \left(D-2\right) \int_{r=0}^{\pi} \left(-\frac{\lambda r^{2}}{2}\right) \mathrm{d}r$$
(13)

$$= \frac{A_{D-2}}{2^{D-2}} {D-2 \choose D/2-1} \int_{r=0}^{D/2-2} \exp\left(-\frac{\lambda r}{2}\right) dr + A_{D-2} \frac{(-1)^{D/2-1}}{2^{D-3}} \sum_{k=0}^{D/2-2} (-1)^k {D-2 \choose k} \int_{r=0}^{\pi} \exp\left(-\frac{\lambda r^2}{2}\right) \cos((D-2-2k)r) dr.$$
(14)

To evaluate this we need two simple integrals, which we evaluate using Maple,

$$\int_{r=0}^{\pi} \exp\left(-\frac{\lambda r^2}{2}\right) \mathrm{d}r = \frac{\sqrt{\pi}}{\sqrt{2\lambda}} \mathrm{erf}\left(\frac{\pi\sqrt{\lambda}}{\sqrt{2}}\right) \tag{15}$$

$$\int_{r=0}^{\pi} \exp\left(-\frac{\lambda r^2}{2}\right) \cos(a \cdot r) \mathrm{d}r = \frac{\sqrt{2\pi}}{4\sqrt{\lambda}} \exp\left(-\frac{a^2}{2\lambda}\right) \left(\operatorname{erf}\left(\frac{\pi\lambda - ai}{\sqrt{2\lambda}}\right) + \operatorname{erf}\left(\frac{\pi\lambda + ai}{\sqrt{2\lambda}}\right)\right), \quad (16)$$

where *i* denotes the complex unit. Inserting these expressions into Eq. 14 and simplifying expressions gives the desired result (2a). Note that in the simple special case D = 2, the spherical normal is a distribution over the unit circle. Here, the normalization constant reduce to

$$\mathcal{Z}_{1}^{(D=2)}(\lambda) = \frac{\sqrt{2\pi}}{\sqrt{\lambda}} \operatorname{erf}\left(\frac{\pi\sqrt{\lambda}}{\sqrt{2}}\right).$$
(17)

<sup>1</sup>Using the convention of https://en.wikipedia.org/wiki/N-sphere#Spherical\_coordinates

We now consider the case where D - 2 is odd. Akin to the previous derivation, we insert Eq. 12 into Eq. 10 and get

$$\mathcal{Z}_{1}^{(\text{odd})}(\lambda) = A_{D-2} \int_{r=0}^{\pi} \exp\left(-\frac{\lambda r^{2}}{2}\right) \sin^{D-2}(r) \mathrm{d}r$$
(18)

$$=A_{D-2}\frac{(-1)^{(D-3)/2}}{2^{D-3}}\sum_{k=0}^{(D-3)/2}(-1)^k\binom{D-2}{k}\int_{r=0}^{\pi}\exp\left(-\frac{\lambda r^2}{2}\right)\sin((D-2-2k)r)\mathrm{d}r.$$
 (19)

To evaluate this, we need to evaluate the integral

$$\int_{r=0}^{\pi} \exp\left(-\frac{\lambda r^2}{2}\right) \sin(a \cdot r) dr = -\frac{i\sqrt{2\pi}}{4\sqrt{\lambda}} \exp\left(-\frac{a^2}{2\lambda}\right) \left(2\operatorname{erf}\left(\frac{ia}{\sqrt{2\lambda}}\right) + \operatorname{erf}\left(\frac{\pi\lambda - ia}{\sqrt{2\lambda}}\right) - \operatorname{erf}\left(\frac{\pi\lambda + ia}{\sqrt{2\lambda}}\right)\right),$$
(20)

which we have evaluated using Maple. Inserting this expression into Eq. 19 and simplifying gives the result in Eq. 2b. In the important special-case D = 3, the normalization constant reduce to

$$\mathcal{Z}_{1}^{(D=3)}(\lambda) = \frac{-i\pi^{3/2}}{\sqrt{2\lambda}} \exp\left(-\frac{1}{2\lambda}\right) \left\{ 2\operatorname{erf}\left(\frac{i}{\sqrt{2\lambda}}\right) + \operatorname{erf}\left(\frac{\pi\lambda - i}{\sqrt{2\lambda}}\right) - \operatorname{erf}\left(\frac{\pi\lambda + i}{\sqrt{2\lambda}}\right) \right\}.$$
 (21)

This concludes the derivation.

## A. Quality of the approximation

We approximate the inverse normalization constant  $1/z_1$  with a straight line in order to derive an expression for the anisotropic normalization constant. Figure 1 (center) show the inverse normalization constant for varying values of  $\lambda$ . The right panel of the figure show the difference between the inverse normalization and a fitted straight line. From this we draw two conclusions: 1) the inverse normalization is indeed *not* a straight line; 2) a straight line is, however, a good approximation. While using one globally fitted straight line gives a fairly accurate estimate of the normalization constant, we find that accuracy can be slightly improved by fitting the line locally. We make this local fit through  $1/z_1(\lambda_2)$  and  $1/z_1(\lambda_2 + \alpha(\lambda_1 - \lambda_2))$ . By extensive numerical optimization we have found that  $\alpha^{-1} = 0.46 \lambda_2 + 1.55$  minimizes the worst-case approximation error of the integral.

At times it may be easier to interpret a variance parameter rather than a concentration parameter. The variance of the spherical normal distribution is defined as [1]

$$\mathsf{Var}[\mathbf{x}] = \int_{\mathcal{S}^{D-1}} \arccos^2(\mathbf{x}^{\mathsf{T}} \boldsymbol{\mu}) \; \mathsf{SN}(\mathbf{x} \mid \boldsymbol{\mu}, \boldsymbol{\Lambda}) \mathrm{d}\mathbf{x}. \tag{22}$$

When the distribution is isotropic, this expression can be evaluated for  $S^2$  similarly to the proof of proposition 1 to give

$$\operatorname{Var}[\mathbf{x}] = -\frac{\pi}{\lambda^{5/2}} \left( -\frac{i\sqrt{\pi}(\lambda-1)}{\sqrt{2}} \exp(-\frac{1}{2\lambda}) \operatorname{erf}\left(\frac{i-\pi\lambda}{\sqrt{2\lambda}}\right) -\frac{i\sqrt{\pi}(\lambda-1)}{\sqrt{2}} \exp(-\frac{1}{2\lambda}) \operatorname{erf}\left(\frac{i+\pi\lambda}{\sqrt{2\lambda}}\right) +i(\lambda-1)\sqrt{2\pi} \exp(-\frac{1}{2\lambda}) \operatorname{erf}\left(\frac{i}{\sqrt{2\lambda}}\right) -2\sqrt{\lambda} \exp(-\frac{\pi^{2}\lambda}{2}) - 2\sqrt{\lambda} \right).$$

$$(23)$$

The left panel of Fig. 1 show how the variance change as a function of  $\lambda$ . Notice that the curve is roughly shaped as  $1/\lambda$ .

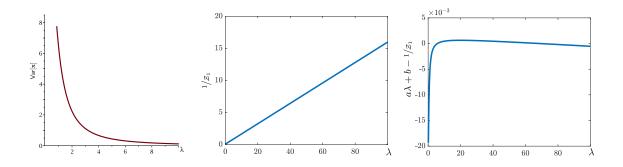


Fig. 1. Left: the variance as a function of the concentration. Center: the inverse normalization constant for the isotropic distribution. Right: The deviation between the inverse normalization constant and a single linear approximation.

## REFERENCES

- [1] X. Pennec. Intrinsic Statistics on Riemannian Manifolds: Basic Tools for Geometric Measurements. *Journal of Mathematical Imaging and Vision (JMIV)*, 25(1):127–154, 2006.
- [2] E. W. Weisstein. Trigonometric power formulas. From MathWorld A Wolfram Web Resource. http://mathworld.wolfram.com/TrigonometricPowerFormulas.html.