

Generate Random Samples from von Mises-Fisher and Watson Distributions

Yu-Hui Chen @ University of Michigan

August 5, 2015

1 Formula

The following derivation is based on the normal-tangent decomposition property of the distribution on sphere. Let $f(\mathbf{x}; \boldsymbol{\mu})$ to be the p.d.f. of the distribution where $\boldsymbol{\mu}$ is the mean direction. The random variable x can be decomposed as:

$$\begin{aligned}\mathbf{x} &= (\mathbf{x}^T \boldsymbol{\mu}) \boldsymbol{\mu} + (I_p - \boldsymbol{\mu} \boldsymbol{\mu}^T) \mathbf{x} \\ &= (\mathbf{x}^T \boldsymbol{\mu}) \boldsymbol{\mu} + \|I_p - \boldsymbol{\mu} \boldsymbol{\mu}^T\| S_{\boldsymbol{\mu}}(\mathbf{x}),\end{aligned}\tag{1}$$

where $S_{\boldsymbol{\mu}}(\mathbf{x}) = (I_p - \boldsymbol{\mu} \boldsymbol{\mu}^T) \mathbf{x} / \|(I_p - \boldsymbol{\mu} \boldsymbol{\mu}^T) \mathbf{x}\|$. Under any rotationally symmetric distribution, $S_{\boldsymbol{\mu}}(\mathbf{x})$ is uniformly distributed on $S_{\boldsymbol{\mu}^\perp}^{p-2}$ and the density of $t = \mathbf{x}^T \boldsymbol{\mu}$ is given by:

$$t \mapsto c f(t) (1 - t^2)^{(p-3)/2}\tag{2}$$

Von Mises-Fisher Distribution

The p.d.f. of VMF distribution has the following form:

$$f(\mathbf{x}; \boldsymbol{\mu}, \kappa) = \left(\frac{\kappa}{2}\right)^{p/2-1} \frac{1}{\Gamma(p/2) I_{p/2-1}(\kappa)} \exp\{\kappa \boldsymbol{\mu}^T \mathbf{x}\}\tag{3}$$

Let $\mathbf{x} = t \boldsymbol{\mu} + (1 - t^2)^{1/2} \boldsymbol{\xi}$ and substitute in (3) we have:

$$f(t, \boldsymbol{\xi}; \boldsymbol{\mu}, \kappa) = \left(\frac{\kappa}{2}\right)^{p/2-1} \frac{1}{\Gamma(p/2) I_{p/2-1}(\kappa)} \exp\{\kappa t\}\tag{4}$$

According to (2), we know that the density of t is proportional to $\exp\{\kappa t\}(1-t^2)^{(p-3)/2}$ and the integration from -1 to 1 should be equal to 1:

$$\begin{aligned} \int_{-1}^1 f(t)dt &= C \int_{-1}^1 \exp\{\kappa t\}(1-t^2)^{(p-3)/2}dt = 1 \\ \Rightarrow C &= \left(\int_{-1}^1 \exp\{\kappa t\}(1-t^2)^{(p-3)/2}dt \right)^{-1} \end{aligned} \quad (5)$$

From the equation 9.6.18 of [1], we have

$$I_\nu(\kappa) = \frac{(\kappa/2)^\nu}{\Gamma(\nu + \frac{1}{2})\Gamma(\frac{1}{2})} \int_{-1}^1 \exp\{\kappa t\}(1-t^2)^{\nu-\frac{1}{2}}dt \quad (6)$$

Let $\nu = p/2 - 1$ in (6)

$$\begin{aligned} \int_{-1}^1 \exp\{\kappa t\}(1-t^2)^{(p-3)/2}dt &= (\kappa/2)^{-(p/2-1)} I_{p/2-1}(\kappa) \Gamma(\frac{p-1}{2}) \Gamma(\frac{1}{2}) \\ \Rightarrow C &= (\frac{\kappa}{2})^{(p/2-1)} \left(I_{p/2-1}(\kappa) \Gamma(\frac{p-1}{2}) \Gamma(\frac{1}{2}) \right)^{-1} \\ \Rightarrow f(t) &= (\frac{\kappa}{2})^{(p/2-1)} \left(I_{p/2-1}(\kappa) \Gamma(\frac{p-1}{2}) \Gamma(\frac{1}{2}) \right)^{-1} \exp\{\kappa t\}(1-t^2)^{(p-3)/2} \end{aligned} \quad (7)$$

Notice that the subscript $p/2 - 1$ of Bessel function I is different and should be corrected from (9.3.12) in Mardia's book [2], which is $(p-1)/2$.

Watson Distribution

The p.d.f. of Watson distribution has the following form:

$$f(\mathbf{x}; \boldsymbol{\mu}, \kappa) = \mathbb{M}(\frac{1}{2}, \frac{p}{2}, \kappa)^{-1} \exp\{\kappa(\boldsymbol{\mu}^T \mathbf{x})^2\} \quad (8)$$

Follow the similar flow, we substitute x by $t\boldsymbol{\mu} + (1-t^2)^{1/2}\boldsymbol{\xi}$ and obtain

$$f(t, \boldsymbol{\xi}; \boldsymbol{\mu}, \kappa) = \mathbb{M}(\frac{1}{2}, \frac{p}{2}, \kappa)^{-1} \exp\{\kappa t^2\} \quad (9)$$

The density of t is proportional to $\exp\{\kappa t^2\}(1-t^2)^{(p-3)/2}$:

$$\begin{aligned} \int_{-1}^1 f(t)dt &= C \int_{-1}^1 \exp\{\kappa t^2\}(1-t^2)^{(p-3)/2}dt = 1 \\ \Rightarrow C &= \left(\int_{-1}^1 \exp\{\kappa t^2\}(1-t^2)^{(p-3)/2}dt \right)^{-1} \end{aligned} \quad (10)$$

According to (13.2.1) in [1], we have

$$\begin{aligned} \frac{\Gamma(b-a)\Gamma(a)}{\Gamma(b)}\mathbb{M}(a, b, \kappa) &= \int_0^1 \exp\{\kappa x\}x^{a-1}(1-x)^{b-a-1}dx \\ &= \int_0^1 \exp\{\kappa t^2\}t^{2a-2}(1-t^2)^{b-a-1}2tdt \end{aligned} \quad (11)$$

Let $a = 1/2, b = p/2$ in (11)

$$\begin{aligned} \frac{\Gamma(\frac{p-1}{2})\Gamma(\frac{1}{2})}{\Gamma(\frac{p}{2})}\mathbb{M}(\frac{1}{2}, \frac{p}{2}, \kappa) &= 2 \int_0^1 \exp\{\kappa t^2\}(1-t^2)^{(p-3)/2}dt \\ &= \int_{-1}^1 \exp\{\kappa t^2\}(1-t^2)^{(p-3)/2}dt \end{aligned} \quad (12)$$

Substitute (12) into (10)

$$\begin{aligned} C &= \left(\int_{-1}^1 \exp\{\kappa t^2\}(1-t^2)^{(p-3)/2}dt \right)^{-1} \\ &= \frac{\Gamma(\frac{p}{2})}{\Gamma(\frac{p-1}{2})\Gamma(\frac{1}{2})} \frac{1}{\mathbb{M}(\frac{1}{2}, \frac{p}{2}, \kappa)} \\ \Rightarrow f(t) &= \frac{\Gamma(\frac{p}{2})}{\Gamma(\frac{p-1}{2})\Gamma(\frac{1}{2})} \frac{1}{\mathbb{M}(\frac{1}{2}, \frac{p}{2}, \kappa)} \exp\{\kappa t^2\}(1-t^2)^{(p-3)/2} \end{aligned} \quad (13)$$

2 Algorithm

Given the target $\boldsymbol{\mu}$ and k for the von Mises-Fisher or Watson distribution, we can generate the random samples from the following steps:

1. Draw the vector \mathbf{r} from the uniform distribution in $(p-1)$ -dimension unit sphere and concatenate a 0 as the first element to \mathbf{r} to get the p -dimension vector $\mathbf{u} = (0, \mathbf{r}) \in \mathcal{R}^p$.
2. Draw the scalar t from Eq.7 (VMF) or Eq.13. This step requires rejection sampling.

Since the density function has support as $[0, 1]$ and is bounded, rejection sampling is feasible.

3. Combine t and \mathbf{u} according to Eq.2 to get the random variable \mathbf{v} .
4. Notice that \mathbf{v} generated from the last step are from the VMF distribution with $\boldsymbol{\mu} = \mathbf{e}_1$. We have to rotate the generated samples to the target mean direction.

References

- [1] M. Abramowitz, I. A. Stegun, and others, *Handbook of mathematical functions*. Dover New York, 1972, vol. 1, no. 5.
- [2] K. V. Mardia and P. E. Jupp, “Directional statistics,” 1999. [Online]. Available: <http://cds.cern.ch/record/1254260>