Bayesian Inference over the Stiefel Manifold via the Givens Representation: Appendix

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Appendix A: Deriving the Change-of-Measure Term

We derive the simplified form of the differential form showing that

$$\bigwedge_{i=1}^{p} \bigwedge_{j=i+1}^{n} G_{j}^{T} dY_{i} = \prod_{i=1}^{p} \prod_{j=i+1}^{n} \cos^{j-i-1} \theta_{ij}.$$
(A.1)

We point out that Khatri and Mardia (1977) provide a similar expression for a slightly different representation, but do not offer a derivation. We start by considering the determinant of the matrix form of the change-of-measure term:

$$\bigwedge_{i=1}^{p} \bigwedge_{j=i+1}^{n} G_{j}^{T} J_{Y_{i}(\Theta)}(\Theta) d\Theta = \begin{pmatrix} G_{2:n}^{T} J_{Y_{1}(\Theta)}(\Theta) \\ G_{3:n}^{T} J_{Y_{2}(\Theta)}(\Theta) \\ \vdots \\ G_{p:n}^{T} J_{Y_{p}(\Theta)}(\Theta) \end{pmatrix}$$
(A.2)

For $l = 1, \dots, n$, we define the following shorthand notation

$$\partial_{i,i+l}Y_k := \frac{\partial}{\partial \theta_{i,i+l}}Y_k \tag{A.3}$$

and

$$\partial_i Y_k := \begin{pmatrix} \partial_{i,i+1} Y_k & \partial_{i,i+2} Y_k & \cdots & \partial_{in} Y_k. \end{pmatrix} \tag{A.4}$$

In the new notation Equation can be written in the following block matrix form:

$$\begin{pmatrix} G_{2:n}^{T} \partial_{1} Y_{1} & G_{2:n}^{T} \partial_{2} Y_{1} & \cdots & G_{2:n}^{T} \partial_{p} Y_{1} \\ G_{3:n}^{T} \partial_{1} Y_{2} & G_{3:n}^{T} \partial_{2} Y_{2} & \cdots & G_{3:n}^{T} \partial_{p} Y_{2} \\ \vdots & \vdots & \ddots & \vdots \\ G_{p:n}^{T} \partial_{1} Y_{p} & G_{p:n}^{T} \partial_{2} Y_{p} & \cdots & G_{p:n}^{T} \partial_{p} Y_{p} \end{pmatrix}$$
(A.5)

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Note that the block matrices above the diagonal are all zero. This can be seen by observing that the rotations in the Givens representation involving elements greater than i will not affect e_i , i.e. letting $R_i := R_{i,i+1} \cdots R_{in}$,

$$Y_i = R_1 R_2 \cdots R_p e_i = R_1 \cdots R_i e_i. \tag{A.6}$$

Thus for j > i, $\partial_j Y_i = 0$ and the determinant of Expression A.5 simplifies to the product of the determinant of the matrices on the diagonal i.e. the following expression:

$$\prod_{i=1}^{p} \det \left(G_{i+1:n}^{T} \partial_{i} Y_{i} \right). \tag{A.7}$$

A.1 Simplifying Diagonal Block Terms

Let I_i denote the first i columns of the $n \times n$ identity matrix and let I_{-i} represent the last n-i columns. The term $G_{i+1:n}^T$ in expression A.7 can be written as

$$G_{i+1:n}^{T} = I_{-i}^{T} G^{T} = I_{-i}^{T} R_{v}^{T} \cdots R_{1}^{T}. \tag{A.8}$$

To simplify the diagonal block determinant terms in Expression A.7 we take advantage of the following fact

$$\det \left(G_{i+1:n}^T \partial_i Y_i \right) = \det \left(I_{-i}^T R_p^T \cdots R_1^T \right) = \det \left(I_{-i}^T R_i^T \cdots R_1^T \partial_i Y_i \right). \tag{A.9}$$

In other words, the terms $R_p^T \cdots R_{i+1}^T$ have no effect on the determinant. This can be seen by first separating terms so that

$$\det \left(G_{i+1:n}^T \partial_i Y_i \right) = \det \left(\underbrace{I_{-i}^T}_{(n-i)\times n} R_p^T \cdots R_1^T \underbrace{\partial_i Y_i}_{n\times (n-i)} \right) \tag{A.10}$$

$$= \det \left(I_{-i}^T \left[R_p^T \cdots R_{i+1}^T \right] \left[R_i^T \cdots R_1^T \partial_i Y_i \right] \right), \tag{A.11}$$

and then noticing that $R_{i+1} \cdots R_p$ affects only the first i columns of the identity matrix so

$$I_{-i}^{T} \left[R_{p}^{T} \cdots R_{i+1}^{T} \right] = \left(R_{i+1} \cdots R_{p} I_{-i} \right)^{T} = \left(I_{-i} \right)^{T}.$$
 (A.12)

Thus Expression A.7 is equivalent to

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$$\prod_{i=1}^{p} \det \left(I_{-i}^{T} R_{i}^{T} \cdots R_{1}^{T} \partial_{i} Y_{i} \right). \tag{A.13}$$

Now consider the k, l element of the $(n-i) \times (n-i)$ block matrix $I_{-i}^T R_i^T \cdots R_1^T \partial_i Y_i$. This can be written as

$$e_{i+k}^{T} R_{i}^{T} \cdots R_{1}^{T} \partial_{i,i+l} Y_{i} = e_{i+k}^{T} R_{i}^{T} \cdots R_{1}^{T} \partial_{i,i+l} (R_{1} \cdots R_{i} e_{i})$$

$$= e_{i+k}^{T} R_{i}^{T} \cdots R_{1}^{T} R_{1} \cdots R_{i-1} (\partial_{i,i+l} R_{i} e_{i})$$

$$= e_{i+k}^{T} R_{i}^{T} (\partial_{i,i+l} R_{i} e_{i}). \tag{A.14}$$

Since $e_{i+k}^T R_i^T R_i e_i = 0$, taking the derivatives of both sides and applying the product rule yields

$$\partial_{i,i+l}(e_{i+k}^T R_i^T R_i e_i) = \partial_{i,i+l} 0$$

$$\Rightarrow (\partial_{i,i+l} e_{i+k}^T R_i^T) R_i e_i + e_{i+k}^T R_i^T (\partial_{i,i+l} R_i e_i) = 0$$

$$\Rightarrow e_{i+k}^T R_i^T (\partial_{i,i+l} R_i e_i) = -(\partial_{i,i+l} e_{i+k}^T R_i^T) R_i e_i. \tag{A.15}$$

Combining expression A.15 this fact with expression A.14, the expression for the k, l element of $I_{-i}^T R_i^T \cdots R_1^T \partial_i Y_i$ becomes $-(\partial_{i,i+l} e_{i+k}^T R_i^T) R_i e_i$.

However, note that

$$e_{i+k}^T R_i^T = e_{i+k}^T R_{in}^T \cdots R_{i,i+1}^T = e_{i+k}^T R_{i,i+k}^T \cdots R_{i,i+1}^T, \tag{A.16}$$

and the partial derivative of this expression with respect to i, i+l is zero when k>l. Thus it is apparent that $I_{-i}^T R_i^T \cdots R_1^T \partial_i Y_i$ contains zeros above the diagonal and that $(I_{-i}^T R_i^T \cdots R_1^T \partial_i Y_i)$ is simply the product of the diagonal elements of the matrix.

A.2 Diagonal Elements of the Block Matrices

To obtain the diagonal terms of the block matrices, we directly compute $-\partial_{i,i+l}e_{i+k}^TR_i^T$ for $l=k, R_ie_i$, and their inner-product. Defining $D_{ij} := \partial_{ij}R_{ij}$,

$$-\partial_{i,i+k}R_ie_{i+k} = -\partial_{i,i+k}(R_{i,i+1}\cdots R_{i,i+k}e_{i+k})$$
(A.17)

$$= -R_{i,i+1} \cdots R_{i,i+k-1} D_{i,i+k} e_{i+k} \tag{A.18}$$

(A.19)

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$$= R_{i,i+1} \cdots R_{i,i+k-1} \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \cos \theta_{i,i+k} \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$= R_{i,i+1} \cdots R_{i,i+k-2} \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \cos \theta_{i,i+k-1} \cos \theta_{i,i+k} \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$= \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \cos \theta_{i,i+k-1} \cos \theta_{i,i+k} \\ \sin \theta_{i,i+k} \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$= \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \sin \theta_{i,i+k-1} \cos \theta_{i,i+k} \\ \sin \theta_{i,i+k} \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

$$= \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \cos \theta_{i,i+1} \cos \theta_{i,i+2} \cdots \cos \theta_{i,i+k-1} \cos \theta_{i,i+k} \\ \sin \theta_{i,i+k} \cos \theta_{i,i+k-1} \cos \theta_{i,i+k} \\ \sin \theta_{i,i+k-1} \cos \theta_{i,i+k} \\ \sin \theta_{i,i+k} \cos \theta_{i,i+k} \\ \sin \theta_{i,i+k} \cos \theta_{i,i+k} \\ \sin \theta_{i,i+k} \cos \theta_{i,i+k} \\ \cos \theta_{i,i+k} \cos \theta_{i,i+k} \\$$

which is zero up to the *i*th spot. After the i + kth spot,

$$R_i e_i = R_{i,i+1} \cdots R_{in} e_i \tag{A.24}$$

(A.23)

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$$= \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \cos \theta_{i,i+1} \cos \theta_{i,i+2} \cdots \cos \theta_{i,n-1} \cos \theta_{in} \\ \sin \theta_{i,i+1} \cos \theta_{i,i+2} \cdots \cos \theta_{i,n-1} \cos \theta_{in} \\ \vdots \\ \sin \theta_{i,n-1} \cos \theta_{in} \\ \sin \theta_{in} \end{pmatrix}. \tag{A.26}$$

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Finally, directly computing the inner-product of $-\partial_{i,i+l}e_{i+k}^TR_i^T$ and R_ie_i yields

$$-(\partial_{i,i+l}e_{i+k}^TR_i^T)(R_ie_i) = \cos^2\theta_{i,i+1}\cos^2\theta_{i,i+2}\cdots\cos^2\theta_{i,i+k}\cos\theta_{i,i+k+1}\cdots\cos\theta_{in} + \sin^2\theta_{i,i+1}\cos^2\theta_{i,i+2}\cdots\cos^2\theta_{i,i+k}\cos\theta_{i,i+k+1}\cdots\cos\theta_{in} + \sin^2\theta_{i,i+2}\cos^2\theta_{i,i+3}\cdots\cos^2\theta_{i,i+k}\cos\theta_{i,i+k+1}\cdots\cos\theta_{in} \vdots + \sin^2\theta_{i,i+k}\cos\theta_{i,i+k+1}\cdots\cos\theta_{in} = \cos^2\theta_{i,i+2}\cos^2\theta_{i,i+3}\cdots\cos^2\theta_{i,i+k}\cos\theta_{i,i+k+1}\cdots\cos\theta_{in} + \sin^2\theta_{i,i+2}\cos^2\theta_{i,i+3}\cdots\cos^2\theta_{i,i+k}\cos\theta_{i,i+k+1}\cdots\cos\theta_{in} \vdots + \sin^2\theta_{i,i+k}\cos\theta_{i,i+k+1}\cdots\cos\theta_{in} = \cdots = \cos\theta_{i,i+k+1}\cdots\cos\theta_{in} = \prod_{k=i+1}^{n}\cos\theta_{ik}.$$

$$(A.27)$$

Thus the determinant of the entire block matrix $I_{-i}^T R_i^T \cdots R_1^T \partial_i Y_i$ simplifies to

$$\prod_{k=i+1}^{n} \left(\prod_{j=k+1}^{n} \cos \theta_{ik} \right) = \prod_{j=i+1}^{n} \cos^{j-i-1} \theta_{ij}.$$
 (A.28)

Combining this with Expression A.13 yields

$$\prod_{i=1}^{p} \det \left(I_{-i}^{T} R_{i}^{T} \cdots R_{1}^{T} \partial_{i} Y_{i} \right) = \prod_{i=1}^{p} \prod_{j=i+1}^{n} \cos^{j-i-1} \theta_{ij}. \tag{A.29}$$

References

Khatri, C. and Mardia, K. (1977). "The von Mises-Fisher matrix distribution in orientation statistics." *Journal of the Royal Statistical Society. Series B (Methodological)*, 95–106.