

Chapter 42

Handbook of Jacobians

Function	Sizes	Jacobian
$Y = BXA^T$ $= (A \otimes B)X$ “Kronecker Product”	$X \quad m \times n$ $A \quad n \times n$ $B \quad m \times m$	$(dY) = (\det A)^n (\det B)^m$ 8.3.1
$Y = X^{-1}$ “inverse”	$X \quad n \times n$	$(dY) = (\det X)^{-2n} (dX)$
$Y = X^2$	$X \quad n \times n$	$(dY) = \prod_{i,j} \lambda_i + \lambda_j (dX)$
$Y = X^k$	$X \quad n \times n$	$(dY) = \prod_{i \geq j} \left \sum_{l=0}^{k-1} \lambda_i^l \lambda_j^{n-1-l} \right (dX)$ 8.3.2
$Y = \frac{1}{2}(AXB + B^T X A)$ “Symmetric Kronecker”	$X, A, B \quad n \times n$ $A, B \quad \text{sym}$ $B = A^T$	$(dY) = \prod_{i \leq j} \lambda_i M_j + \lambda_j M_i (dX)$ $(dY) = (\det A)^{n+1} (dX)$

42.1 Real Case; General Matrices

In factorizations that involve a square $m \times m$ orthogonal matrix Q , and start from a thin, tall $m \times n$ matrix A ($m \geq n$ always), $H^T dQ$ stands for the wedge product

$$H^T dQ = \bigwedge_{i=1}^n \bigwedge_{j=i+1}^n q_i^T dq_j.$$

Here H is a stand-in for the first n columns of Q . If the differential matrix and the matrix whose transpose we multiply by are one and the same, the wedging is done over all of the columns. For example,

$$V^T dV = \bigwedge_{i=1}^n \bigwedge_{j=i+1}^n v_i^T dv_j,$$

for V an $n \times n$ orthogonal matrix.

The Jacobian of the last factorization (non-symmetric eigenvalue) is computed only for the positive-measure set of matrices with real eigenvalues and with a full set of eigenvectors.

Factorization	Matrix sizes & properties	Parameter count	Jacobian
$A = LU$ lu	A, L, U $n \times n$ L, U^T lower triangular $l_{ii} = 1, \forall i = 1, n$	$n^2 =$ $\frac{n(n-1)}{2} + \frac{n(n+1)}{2}$	$(dA) =$ $\prod_{i=1}^n u_{ii}^{n-i} (dL)(dU)$
$A = QR$ qr	A, R $m \times n$ Q $m \times m$ $Q^T Q = I_n$ R upper triangular	$mn =$ $(mn - \frac{n(n+1)}{2}) + \frac{n(n+1)}{2}$	$(dA) =$ $\prod_{i=1}^n r_{ii}^{m-i} (dR)(H^T dQ)$
$A = U \Sigma V^T$ svd	A, Σ $m \times n$ U $m \times m, V$ $n \times n$ $U^T U = I_m$ $V^T V = I_n$	$mn =$ $(mn - \frac{n(n+1)}{2}) + n + \frac{n(n-1)}{2}$	$(dA) =$ $\prod_{i < j} (\sigma_i^2 - \sigma_j^2) \prod_{i=1}^n \sigma_i^{m-n} (d\Sigma)(H^T dU)(V^T dV)$
$A = QS$ polar	A, Q $m \times n$ S $n \times n$ $Q^T Q = I_n$ S positive definite	$mn =$ $(mn - \frac{n(n+1)}{2}) + \frac{n(n+1)}{2}$	$(dA) =$ $\prod_{i < j} (\sigma_i + \sigma_j) (dS)(Q^T dQ)$
$A = X \Lambda X^{-1}$ nonsymm eig	A, X, Λ $n \times n$ Λ diagonal	$n^2 =$ $(n^2 - n) + n$	$(dA) =$ $\prod_{i < j} (\lambda_i - \lambda_j)^2 (d\Lambda)(X^{-1} dX)$

42.2 Real Matrices; Symmetric “+” Cases

Here we gather factorizations that are symmetric or, in the case when this is required, positive definite. All matrices are square $m \times m$.

Factorization	“+”	Parameter count	Matrix properties	Jacobian
$A = LL'$ Choleski	Positive definite	$\frac{n(n+1)}{2} =$ $\frac{n(n+1)}{2}$	L lower triangular	$(dA) =$ $2^n \prod_{i=1}^n l_{ii}^{n+1-i} (dL)$
$A = LDL'$ ldl	—	$\frac{n(n+1)}{2} =$ $\frac{n(n-1)}{2} + n$	L lower triangular $l_{ii} = 1, \forall i = 1, n$ D diagonal	$(dA) =$ $\prod_{i=1}^n d_i^{n-i} (dL)(dD)$
$A = Q \Lambda Q^T$ eig	—	$\frac{n(n+1)}{2} =$ $\frac{n(n-1)}{2} + n$	$Q^T Q = I_n$ Λ diagonal	$(dA) =$ $\prod_{i < j} \lambda_i - \lambda_j (d\Lambda)(Q^T dQ)$

42.3 Real Matrices; Orthogonal Case

This is the case of the CS decomposition. Note that the matrices U_1 and V_1 share $\frac{p(p-1)}{2}$ parameters; this is due to the fact that the product of the first p columns of U_1 and the first p rows of V_1^T is invariant and determined. This is equivalent to saying that we introduce an equivalence relation of the set of pairs (U_1, V_1) of orthogonal matrices of size k , by having

$$(U_1, V_1) \sim \left(U_1 \begin{bmatrix} Q & 0 \\ 0 & I_j \end{bmatrix}, V_1 \begin{bmatrix} Q & 0 \\ 0 & I_j \end{bmatrix} \right),$$

for any $p \times p$ orthogonal matrix Q .

Since it is rather hard to integrate over a manifold of pairs of orthogonal matrices with the equivalence relationship mentioned above, we will use a different way of thinking about this splitting

of parameters. We can assume that U_1 contains $\frac{k(k-1)}{2}$ parameters (a full set) while V_1 only contains $\frac{k(k-1)}{2} - \frac{p(p-1)}{2}$ (having the first p columns already determined from the relationship with U_1). Alternatively, we could assign a full set of parameters to V_1 and think of U_1 as having the first r columns predetermined. In the following we choose to make the former assumption, and think of the undetermined part H of V_1^T as a point in the Stiefel manifold $V_{j,k}$.

Factorization	Matrix sizes & properties	Parameter count	Jacobian
$Q =$ $\begin{bmatrix} U_1 & 0 \\ 0 & U_2 \end{bmatrix} \begin{bmatrix} I_p & 0 & 0 \\ 0 & C & S \\ 0 & S & -C \end{bmatrix} \begin{bmatrix} V_1 & 0 \\ 0 & V_2 \end{bmatrix}^T$ $n = k + j, p = k - j$ CS decomposition	$Q \ n \times n$ $U_1, V_1 \ k \times k$ $U_2, V_2 \ j \times j$ $Q^T Q = I_n$ U_1, V_1, U_2, V_2 orthogonal $C = \text{diag}(\cos(\theta_i)), i = 1 \dots n$ $S = \text{diag}(\sin(\theta_i)), i = 1 \dots n$	$U_1 V_1^T$ has $k(k-1) - \frac{p(p-1)}{2}$ U_2, V_2 have $\frac{j(j-1)}{2}$ each C and S share j	$(dA) =$ $\prod_{i < j} \sin(\theta_j - \theta_i) \sin(\theta_i + \theta_j) (d\theta)$ $(U_1^T dU_1)(U_2^T dU_2)(H^T dH)(V_2^T dV_2)$

42.4 Real Matrices; Symmetric Tridiagonal

The important factorization of this section is the eigenvalue decomposition of the symmetric tridiagonal. The tridiagonal is defined by its $2n - 1$ parameters, n on the diagonal and $n - 1$ on the upper diagonal:

$$T = \begin{pmatrix} a_n & b_{n-1} & & & & & \\ b_{n-1} & a_{n-1} & b_{n-2} & & & & \\ & & \ddots & \ddots & \ddots & & \\ & & & b_2 & a_2 & b_1 & \\ & & & & b_1 & a_1 & \end{pmatrix}.$$

The matrix of eigenvectors Q can be completely determined from knowledge of its first row $q = (q_1, \dots, q_n)$ ($n - 1$ parameters) and the eigenvalues on the diagonal of Λ (another n parameters). Here dq represents integration over the $(1, n)$ Stiefel manifold.

Factorization	Matrix sizes	Parameter count	Matrix properties	Jacobian
$T = Q \Lambda Q^T$ eig	$T, Q, \Lambda \ n \times n$	$2n - 1 =$ $(n - 1) + n$	$Q^T Q = I_n$ Λ diagonal	$(dT) = \frac{\prod_{i=1}^{n-1} b_i}{\prod_{i=1}^n q_i} (d\Lambda)(dq)$

Chapter 43

Joint Densities

Table of Joint Element Densities

General Matrices	MATLAB	Joint element density
$G(m, n)$	<code>randn(m, n)</code>	$(2\pi)^{-mn/2} \exp(-\frac{1}{2}A^T A)$
$G(n, n)$	<code>randn(n)</code>	$(2\pi)^{-n^2/2} \exp(-\frac{1}{2}A^T A)$
$G(m, n; A, B \otimes A)$	$B^{1/2} \text{randn}(m, n) \cdot A^{-1/2T} + M$	
$G(m, n; B, M)$	$B^{1/2} \text{randn}(m, n) + M$	
$G(m, n; B)$	$B^{1/2} \text{randn}(m, n)$	
Symmetric Matrices	MATLAB	Joint element density
GUE	$A = \text{randn}(n); S = (A + A')/2$	$2^{-n/2} \pi^{-\frac{n(n+1)}{4}} \exp(-\frac{1}{2}S^2)$
$W(m, n)$	$A = \text{randn}(m, n); S = A^T A$	$2^{-mn/2} \Gamma_n^{-1}(\frac{m}{2}) S ^{(m-n-1)/2} \exp(-\frac{1}{2}S)$
$W(m, n; \Sigma)$	$A = \text{randn}(m, n); S = A^T \Sigma A$	$2^{-mn/2} \Gamma_m^{-1}(\frac{m}{2}) \Sigma ^{-m/2} S ^{(m-n-1)/2} \exp(-\frac{1}{2}\Sigma^{-1}S)$
$W(m, n; \Sigma; M)$	$A = \Sigma^{1/2} \text{randn}(m, n) \cdot M; S = A^T A$	