

Matrix Function¹[1]

Definition 3

Let $F : S \rightarrow \mathbf{R}^{m \times p}$ be a matrix function defined on a set S in $\mathbf{R}^{n \times q}$. Let C be an interior point of S , and let $B(C; r) \subset S$ be a ball with centre C and radius r (also called a neighbourhood of C and denoted $N(C)$). Let U be a point in $\mathbf{R}^{n \times q}$ with $\|U\| < r$, so that $C + U \in B(C; r)$. If there exists a real $mp \times nq$ matrix A , depending on C but not on U , such that

$$\text{vec } F(C + U) = \text{vec } F(C) + A(C) \text{vec } U + \text{vec } R_C(U) \quad (2)$$

for all $U \in \mathbf{R}^{n \times q}$ with $\|U\| < r$ and

$$\lim_{U \rightarrow 0} \frac{R_C(U)}{\|U\|} = 0, \quad (3)$$

then the function F is said to be *differentiable at C* . The $m \times p$ matrix $dF(C; U)$ defined by

$$\text{vec } dF(C; U) = A(C) \text{vec } U \quad (4)$$

is then called the *(first) differential of F at C with increment U* and the $mp \times nq$ matrix $A(C)$ is called the *(first) derivative of F at C* .

¹Magnus, Jan R.; Neudecker, Heinz (1999). Matrix differential calculus with applications in statistics and econometrics

Matrix Function²

In view of Definition 3, all calculus properties of matrix functions follow immediately from the corresponding properties of vector functions because, instead of the matrix function F , we can consider the vector function $f : \text{vec } S \rightarrow \mathbf{R}^{mp}$ defined by

$$f(\text{vec } X) = \text{vec } F(X). \quad (7)$$

It is easy to see from (2) and (3) that the differentials of F and f are related by

$$\text{vec } dF(C; U) = df(\text{vec } C; \text{vec } U). \quad (8)$$

We then define the *Jacobian matrix of F at C* as

$$DF(C) = Df(\text{vec } C). \quad (9)$$

This is an $mp \times nq$ matrix, whose ij -th element is the partial derivative of the i -th component of $\text{vec } F(X)$ with respect to the j -th element of $\text{vec } X$, evaluated at $X = C$.

²Magnus, Jan R.; Neudecker, Heinz (1999). Matrix differential calculus with applications in statistics and econometrics

$$f(\xi) = \mathbf{x}(\xi)^T \mathbf{C}(\xi)^{-1} \mathbf{x}(\xi)$$

$$\xi \in \mathbb{R}^6, \mathbf{x} \in \mathbb{R}^3, \mathbf{C} \in \mathbb{R}^{3 \times 3}$$

$$\frac{\partial f(\xi)}{\partial \xi} = (\mathbf{D}\mathbf{C}^{-1})^T \cdot \text{vec } \mathbf{x}\mathbf{x}^T + \frac{\partial \mathbf{x}^T}{\partial \xi} (\mathbf{C} + \mathbf{C}^T) \mathbf{x}$$

边缘点协方差 ${}^c\mathbf{C}_{pk}$ 的估计

取 ${}^c\mathbf{p}_k$ 邻域内的边缘点，拟合协方差 ${}^c\mathbf{C}_{pk}$ ，假设其特征值 $\lambda_{p1}, \lambda_{p2}, \lambda_{p2}$ ，及对应特征向量 $\mathbf{u}_{p1}, \mathbf{u}_{p2}, \mathbf{u}_{p2}$ 。

$${}^c\mathbf{C}_{pk}^{-1} = \frac{1}{\lambda_{p1}} \mathbf{u}_{p1} \mathbf{u}_{p1}^T + \frac{1}{\lambda_{p2}} \mathbf{u}_{p2} \mathbf{u}_{p2}^T + \frac{1}{\lambda_{p3}} \mathbf{u}_{p3} \mathbf{u}_{p3}^T \quad (1)$$

对于边缘上的点，有 $\lambda_{p1} \gg \lambda_{p2} \geq \lambda_{p2}$ ，即 $\frac{1}{\lambda_{p3}} \geq \frac{1}{\lambda_{p2}} \gg \frac{1}{\lambda_{p1}}$ 。则

$$\begin{aligned} J_{pk}(\xi) &= \frac{1}{2} ({}^c\mathbf{p}_k - T_{cr}({}^r\mathbf{p}_k, \xi))^T {}^c\mathbf{C}_{pk}^{-1} ({}^c\mathbf{p}_k - T_{cr}({}^r\mathbf{p}_k, \xi)) \\ &\approx \frac{1}{2} ({}^c\mathbf{p}_k - T_{cr}({}^r\mathbf{p}_k, \xi))^T \left(\frac{1}{\lambda_{p2}} \mathbf{u}_{p2} \mathbf{u}_{p2}^T + \frac{1}{\lambda_{p3}} \mathbf{u}_{p3} \mathbf{u}_{p3}^T \right) ({}^c\mathbf{p}_k - T_{cr}({}^r\mathbf{p}_k, \xi)) \end{aligned} \quad (2)$$

只有沿 ${}^c\mathbf{p}_k$ 所在边缘的垂直方向上的运动才会导致 $J_{pk}(\xi)$ 变化。

$$J_{pk}(\xi) = \frac{1}{2} ({}^c\mathbf{p}_k - {}^r\mathbf{p}_k)^T \mathbf{C}_{vpk}^{-1} ({}^c\mathbf{p}_k - {}^r\mathbf{p}_k)$$

$$\begin{aligned} \mathbf{C}_{vpk} &= {}^c\mathbf{C}_{pk} + {}^r\mathbf{C}_{pk} \\ &= \frac{1}{N-1} \sum_{i=1}^N ({}^c\mathbf{p}_{pi} - {}^c\mu_p) ({}^c\mathbf{p}_{pi} - {}^c\mu_p)^T + \frac{1}{N-1} \sum_{j=1}^N ({}^r\mathbf{p}_{pj} - {}^r\mu_p) ({}^r\mathbf{p}_{pj} - {}^r\mu_p)^T \\ &= \frac{1}{N-1} \left[\sum_{i=1}^N {}^c\bar{\mathbf{p}}_{pi} {}^c\bar{\mathbf{p}}_{pi}^T + \sum_{j=1}^N {}^r\bar{\mathbf{p}}_{pj} {}^r\bar{\mathbf{p}}_{pj}^T \right] \\ &= \frac{1}{N-1} \sum_{i=1}^{2N} \bar{\mathbf{p}}_{pi} \bar{\mathbf{p}}_{pi}^T \end{aligned}$$

[2]



J. R. Magnus and H. Neudecker, *Matrix differential calculus with applications in statistics and econometrics* /.
Wiley,, 1988.



A. Knutson and T. Tao, “Honeycombs and sums of hermitian matrices,” *Notices of the American Mathematical Society*, vol. 48, no. 2, pp. 175–186, 2000.