第一章 矩阵代数

- 1.1 分块矩阵
- 1.2 广义逆
- 1.3 拉直运算和Kronecker积
- 1.4 矩阵的微商

§ 1.1 分块矩阵

定义1.1.1

$$A_{11}: k \times l, A_{12}: k \times (q-l),$$

$$A_{21}:(p-k)\times l, \quad A_{22}:(p-k)\times (q-l),$$

则
$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}$$
称为矩阵 A 的分块表示形式.

若A和B有相同的分块,则

$$A + B = \begin{pmatrix} A_{11} + B_{11} & A_{12} + B_{12} \\ A_{21} + B_{21} & A_{22} + B_{22} \end{pmatrix}$$

若
$$C$$
为 $q imes r$ 矩阵,它分为 $C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$,其中
$$C_{11} : l imes m , C_{12} : l imes (r-m), C_{21} : (q-l) imes m,$$

$$C_{22} : (q-l) imes (r-m), 则$$

$$AC = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

$$= \begin{pmatrix} A_{11}C_{11} + A_{12}C_{21} & A_{11}C_{12} + A_{12}C_{22} \\ A_{21}C_{11} + A_{22}C_{21} & A_{21}C_{12} + A_{22}C_{22} \end{pmatrix}$$

若A为方阵, A_1 也为方阵,

(1) 若
$$|A_{11}| \neq 0$$
, 则 $|A| = |A_{11}| A_{22\cdot 1}|$ 其中 $A_{22\cdot 1} = A_{22} - A_{21}A_{11}^{-1}A_{12}$

(2) 若
$$|A_{22}| \neq 0$$
, 则 $|A| = |A_{11\cdot 2}||A_{22}|$ 其中 $A_{11\cdot 2} = A_{11} - A_{12}A_{22}^{-1}A_{21}$

证明:(1) $|A_{11}| \neq 0$, 利用

$$\begin{pmatrix}
I & 0 \\
-A_{21}A_{11}^{-1} & I
\end{pmatrix}
\begin{pmatrix}
A_{11} & A_{12} \\
A_{21} & A_{22}
\end{pmatrix}
\begin{pmatrix}
I & -A_{11}^{-1}A_{12} \\
0 & I
\end{pmatrix}$$

$$= \begin{pmatrix}
A_{11} & 0 \\
0 & A_{22\cdot 1}
\end{pmatrix}$$
(1.1)

两边同时取行列式即可.

(2) $|A_{22}| \neq 0$,利用

$$\begin{pmatrix} I & -A_{12}A_{22}^{-1} \end{pmatrix} \begin{pmatrix} A_{11} & A_{12} \end{pmatrix} \begin{pmatrix} I & 0 \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} -A_{22}^{-1}A_{21} & I \end{pmatrix}$$

$$= \begin{pmatrix} A_{11\cdot 2} & 0 \\ 0 & A_{22} \end{pmatrix}$$
 (1.2)

两边同时取行列式即可.

若A为可逆方阵, A_{11} 和 A_{22} 均为方阵

(1) 若 $A_{11} \neq 0$, 则

$$A^{-1} = \begin{pmatrix} A_{11}^{-1} + A_{11}^{-1} A_{12} A_{22\cdot 1}^{-1} A_{21} A_{11}^{-1} & -A_{11}^{-1} A_{12} A_{22\cdot 1}^{-1} \\ -A_{22\cdot 1}^{-1} A_{21} A_{11}^{-1} & A_{22\cdot 1}^{-1} \end{pmatrix}$$

(2) 若 $|A_{22}| \neq 0$,则

$$A^{-1} = \begin{pmatrix} A_{11\cdot2}^{-1} & -A_{11\cdot2}^{-1} A_{12} A_{22}^{-1} \\ -A_{22}^{-1} A_{21} A_{11\cdot2}^{-1} & A_{22}^{-1} + A_{22}^{-1} A_{21} A_{11\cdot2}^{-1} A_{12} A_{22}^{-1} \end{pmatrix}$$

性质4(续)

(3) 若
$$|A_{11}| \neq 0$$
, $|A_{22}| \neq 0$, 则

$$A^{-1} = \begin{pmatrix} A_{11 \cdot 2}^{-1} & -A_{11}^{-1} A_{12} A_{22 \cdot 1}^{-1} \\ -A_{22}^{-1} A_{21} A_{11 \cdot 2}^{-1} & A_{22 \cdot 1}^{-1} \end{pmatrix}$$

证明: (1)对(1.1)式两边求逆

$$\begin{pmatrix} I & -A_{11}^{-1}A_{12} \\ 0 & I \end{pmatrix}^{-1} \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}^{-1} \begin{pmatrix} I & 0 \\ -A_{21}A_{11}^{-1} & I \end{pmatrix}^{-1} = \begin{pmatrix} A_{11}^{-1} & 0 \\ 0 & A_{22\cdot 1}^{-1} \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}^{-1} = \begin{pmatrix} I & -A_{11}^{-1}A_{12} \\ 0 & I \end{pmatrix} \begin{pmatrix} A_{11}^{-1} & 0 \\ 0 & A_{22\cdot 1}^{-1} \end{pmatrix} \begin{pmatrix} I & 0 \\ -A_{21}A_{11}^{-1} & I \end{pmatrix}$$

$$= \begin{pmatrix} A_{11}^{-1} + A_{11}^{-1}A_{12}A_{22\cdot 1}^{-1}A_{21}A_{11}^{-1} & -A_{11}^{-1}A_{12}A_{22\cdot 1}^{-1} \\ A_{22\cdot 1}^{-1}A_{21}A_{11}^{-1} & A_{22\cdot 1}^{-1} \end{pmatrix}$$

§ 1.2 矩阵的广义逆

一、减号逆

定义1.2.1 设A为 $n \times p$ 阶矩阵,若存在一个 $p \times n$ 阶矩阵X,使得 AXA = A,则称X为A的广义逆或A的减号逆,记作 $X = A^-$.

性质1 任何矩阵的广义逆一定存在,但可能不唯一。

证明:设rank(A) = r, A 可分解为

$$A = P \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} Q$$

其中P和Q分别为n和p阶非奇异方阵,于是有

$$AXA = A \Leftrightarrow$$

$$P\begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} QXP\begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} Q = P\begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} Q$$

$$\begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} QXP \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix}$$

记

$$QXP = \begin{pmatrix} T_{11} & T_{12} \\ & & \\ T_{21} & T_{22} \end{pmatrix} \quad \sharp P T_{11} : r \times r$$

于是

$$\begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} T_{11} & T_{12} \\ T_{21} & T_{22} \end{pmatrix} \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix}$$

$$\Leftrightarrow T_{11} = I_r$$

所以
$$QXP = \begin{pmatrix} I_r & T_{12} \\ T_{21} & T_{22} \end{pmatrix}$$

即
$$X = Q^{-1} \begin{pmatrix} I_r & T_{12} \\ T_{21} & T_{22} \end{pmatrix} P^{-1}$$

其中 T_{12} , T_{21} , T_{22} 可任意选择,X就是A的广义逆 A^- , 这表明 A^- 一定存在,但可能不唯一.

性质**2** 若A非退化,则 A^- 唯一,且 $A^- = A^{-1}$.

证明: 因为 $AA^{-1}A = A$,所以 A^{-1} 是A的一个 广义逆. 若X也是A的广义逆,则 AXA = A $\Leftrightarrow X = A^{-1}$,唯一性得证. 性质3 $rk(A^-) \ge rk(A)$

证明 由性质1中知,若rank(A) = r,A 可分解为

$$A = P \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} Q$$

其中P和Q分别为n和p阶非奇异方阵,于是有

$$A^{-} = Q^{-1} \begin{pmatrix} I_r & T_{12} \\ T_{21} & T_{22} \end{pmatrix} P^{-1}$$

所以 $rk(A^-) \ge rk(A)$.

$$rk(A) = rk(AA^{-}) = rk(A^{-}A)$$
$$= tr(AA^{-}) = tr(A^{-}A)$$

证明 若rank(A) = r, A 可分解为

$$A = P \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} Q$$

其中P和Q分别为n和p阶非奇异方阵,于是有

$$A^{-} = Q^{-1} \begin{pmatrix} I_r & T_{12} \\ T_{21} & T_{22} \end{pmatrix} P^{-1}$$

$$\Rightarrow AA^{-} = P \begin{pmatrix} I_{r} & 0 \\ 0 & 0 \end{pmatrix} QQ^{-1} \begin{pmatrix} I_{r} & T_{12} \\ T_{21} & T_{22} \end{pmatrix} P^{-1}$$
$$= P \begin{pmatrix} I_{r} & T_{12} \\ 0 & 0 \end{pmatrix} P^{-1}$$

$$\Rightarrow (AA^{-})^{2} = AA^{-}$$

所以AA⁻为幂等阵

同理

$$A^{-}A = Q^{-1} \begin{pmatrix} I_r & T_{12} \\ T_{21} & T_{22} \end{pmatrix} P^{-1}P \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} Q$$

$$=Q^{-1}\begin{pmatrix}I_r&0\\T_{21}&0\end{pmatrix}Q$$

$$\Rightarrow (A^-A)^2 = A^-A$$

所以 AA^{-} 和 $A^{-}A均为幂等阵$

特别,

若
$$rk(A) = p$$
,则 $A^-A = I_p$

若
$$rk(A) = n$$
,则 $AA^- = I_n$

对任意矩阵A,有

$$A'A(A'A)^{-}A' = A' \quad A(A'A)^{-}A'A = A$$

$$A'A(A'A)^{-}A'A = A'A$$

证明

$$(1)Ax = 0 \Leftrightarrow A'Ax = 0$$

事实上,

$$Ax = 0 \Rightarrow A'Ax = 0$$

$$\Rightarrow x'A'Ax = 0 \Rightarrow Ax = 0$$

$$(2)Ax = Ay \Leftrightarrow A'Ax = A'Ay$$

$$(3)$$
对 $\forall x \in \mathbb{R}^p$,有

$$A'A(A'A)^{-}A'Ax = A'Ax$$
,利用(2)

⇒ 対
$$\forall x \in \mathbb{R}^p$$
, 有 $A(A'A)^-A'Ax = Ax$,

$$\Rightarrow A(A'A)^{-}A'A = A$$

同理可证
$$A'A(A'A)^-A'=A'$$

等号两边同时消去一个相同矩阵的一般结论: 定理

- (1)ABC = 0和BC = 0等价的充分必要 条件是rk(AB) = rk(B).
- (2)CAB = 0和CA = 0等价的充分必要 条件是rk(AB) = rk(A).

 $A(A'A)^{-}A'$ 是一个投影矩阵,

且与(A'A)⁻的取法无关.

证明

 $(1) A(A'A)^{-} A' 与 (A'A)^{-}$ 的取法无关.

设
$$rk(A) = r$$
, $(A'A)_1^- 和 (A'A)_2^- 是 A'A$ 的两个广义

逆,则有
$$A'A(A'A)_1^-A'A = A'A(A'A)_2^-A'A = A'A$$

利用性质
$$5 \Rightarrow A'A(A'A)_1^-A' = A'A(A'A)_2^-A' = A'$$

$$\Rightarrow A(A'A)_1^- A' = A(A'A)_2^- A'$$

$(2) A(A'A)^{-}A'$ 一定是对称阵.

由于A'A为对称阵,存在正交阵P,使得

$$A'A = P'diag(\lambda_1^2, \lambda_2^2, \dots, \lambda_r^2, 0, \dots, 0)P$$

$$= P'diag(\lambda_1, \lambda_2, \dots, \lambda_r, 1, \dots, 1) \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} diag(\lambda_1, \lambda_2, \dots, \lambda_r, 1, \dots, 1) P$$

它是对称阵. 所以 $A(A'A)^-A'$ 一定是对称阵.

 $(3)A(A'A)^{-}A'$ 是幂等阵.

$$(A(A'A)^{-}A')^{2} = \underline{A(A'A)^{-}A'A(A'A)^{-}A'}$$

$$= A(A'A)^{-}A'$$

故 $A(A'A)^-A'$ 是一个投影矩阵

 $A(A'A)^{-}A'$ 是到R(A)空间的投影矩阵,

记作
$$P_A = A(A'A)^- A'$$

 $(1) \forall x \in R(A)$,有 $P_A x = x$.

$$(2)$$
对 $\forall x \in \mathbb{R}^n, P_A x \in \mathbb{R}(A).$

证明

$$(1) \forall x \in R(A)$$
,有 $P_A x = x$.

事实上

 $\forall x \in R(A)$, 存在 $y \in R^p$, 使得x = Ay.

$$P_A x = A(A'A)^- A'Ay = Ay = x$$

 $(2) \forall \forall x \in R^n, P_A x \in R(A).$

故 $P_A = A(A'A)^- A'$ 是到R(A)空间的投影矩阵

二、加号逆

定义1.2.2 设A是一个 $n \times p$ 阶矩阵,若存在一

 $^{p\times n}$ 阶矩阵X,使得

$$\begin{cases} AXA = A & XAX = X \\ (AX)' = AX & (XA)' = XA \end{cases}$$

则称X是A的加号逆,也称Moore - Penrose逆,记作 $X = A^{+}$.

对任何矩阵A,A⁺存在且唯一.

- (1) 存在性
- (2) 唯一性

证明 (存在性)
$$\diamondsuit rk(A) = r$$
,

- (1)当r = 0时,必有A = 0,这时取 $A^+ = 0$ 即可.
- (2)当r > 0时,则存在可逆阵 P_1 和 Q_1 使得

$$A = P_1 \begin{pmatrix} I_r & 0 \\ 0 & 0 \end{pmatrix} Q_1 = P_1 \begin{pmatrix} I_r \\ 0 \end{pmatrix} (I_r 0) Q_1 \stackrel{\triangle}{=} PQ'$$

其中
$$P = P_1 \begin{pmatrix} I_r \\ 0 \end{pmatrix}$$
, $Q' = (I_r \ 0)Q_1$, 且 P 和 Q 均为

列满秩矩阵. 所以P'P和Q'Q均可逆.

令 $X = Q(Q'Q)^{-1}(P'P)^{-1}P'$,我们来验证X满足加号逆的四个条件.

$$(1)AXA = PQ'Q(Q'Q)^{-1}(P'P)^{-1}P'PQ' = PQ' = A$$

$$(2)XAX = Q(Q'Q)^{-1}(P'P)^{-1}P'PQ'Q(Q'Q)^{-1}(P'P)^{-1}P'$$

$$= Q(Q'Q)^{-1}(P'P)^{-1}P' = X$$

$$(3)AX = PQ'Q(Q'Q)^{-1}(P'P)^{-1}P' = P(P'P)^{-1}P'$$
 对称

$$(4)XA = Q(Q'Q)^{-1}(P'P)^{-1}P'PQ' = Q(Q'Q)^{-1}Q'$$
 对称 故 X 是 A 的加号逆,存在性得证.

(唯一性)

 $若A_1^+ 和 A_2^+ 是A$ 的两个加号逆,则

$$A_{1}^{+} = A_{1}^{+} A A_{1}^{+} = A_{1}^{+} (A A_{1}^{+})' = A_{1}^{+} (A_{1}^{+})' A' = A_{1}^{+} (A_{1}^{+})' (A A_{2}^{+} A)'$$

$$= A_{1}^{+} (A_{1}^{+})' A' (A A_{2}^{+})' = A_{1}^{+} (A A_{1}^{+})' (A A_{2}^{+})' = A_{1}^{+} A A_{1}^{+} A A_{2}^{+}$$

$$= A_{1}^{+} A A_{2}^{+}$$

$$A_{2}^{+} = A_{2}^{+} A A_{2}^{+} = (A_{2}^{+} A)' A_{2}^{+} = A' (A_{2}^{+})' A_{2}^{+} = (A A_{1}^{+} A)' (A_{2}^{+})' A_{2}^{+}$$

$$= A' (A_{1}^{+})' A' (A_{2}^{+})' A_{2}^{+} = (A_{1}^{+} A)' (A_{2}^{+} A)' A_{2}^{+} = A_{1}^{+} A A_{2}^{+} A A_{2}^{+}$$

$$= A_{1}^{+} A A_{2}^{+}$$
唯一性得证.

若A行满秩,则 $A^+ = A'(AA')^{-1}$.

若A列满秩,则 $A^+ = (A'A)^{-1}A^T$.

若A可逆,则 $A^+ = A^{-1}$.

性质
$$3 (A^+)^+ = A$$

性质4
$$A^+ = A'(AA')^+ = (A'A)^+ A'$$

性质5
$$(A'A)^+ = A^+(A^+)'$$

性质
$$\mathbf{6} (A')^{\dagger} = (A^{\dagger})'$$

由此,当A为对称阵时,A⁺也是对称矩阵

记
$$A = PQ'$$
,其中

$$rk(A) = rk(P) = rk(Q) = r$$
 $_{n \times p}$

则
$$A^+ = (Q^+)'P^+$$
.

注意:

一般,对任意两个矩阵

$$(AB)^{^{+}}=B^{^{+}}A^{^{+}}$$

未必成立.

若A是投影矩阵,则 $A^+ = A$

若A为n阶对称方阵,有分解 A = H'AH,

H为正交矩阵, $\Lambda = diag(\lambda_1, \lambda_2, \dots, \lambda_n)$,令

$$\lambda^{+} = \begin{cases} \lambda^{-1} & \ddot{\Xi}\lambda \neq \mathbf{0} \\ \mathbf{0} & \ddot{\Xi}\lambda = \mathbf{0} \end{cases}$$

则 $A^+ = H'diag(\lambda_1^+, \lambda_2^+, \cdots \lambda_n^+)H$.

 $AA^{+}与A^{+}A均为投影矩阵,$

记
$$P_A = AA^+$$

$$P_{A'} = A^+ A$$

§ 1.3 矩阵的拉直运算和Kronecker积

$$\underset{n\times p}{A}=(a_1,a_2,\cdots,a_p)$$

$$\overrightarrow{A} = vec(A) = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_p \end{pmatrix}$$

定义 1.3.1 设A =
$$(a_{ij})$$
 = (a_1, a_2, \dots, a_p) =
$$\begin{pmatrix} a'_{(1)} \\ a'_{(2)} \\ \vdots \\ a'_{(n)} \end{pmatrix}$$

为
$$n \times p$$
阶矩阵,则 $vec(A) = \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_p \end{pmatrix}$ 称为矩阵 A 的

按列拉直运算,记作 vec(A)或 $\stackrel{\rightarrow}{A}$.

若c,d是实数,A和B是两个大小相同的

矩阵,则

$$vec(c \cdot A + d \cdot B) = c \cdot vec(A) + d \cdot vec(B)$$

性质2

$$A_{n \times p} = \sum_{i,j} a_{ij} E_{ij} = \sum_{i,j} a_{ij} e_i e'_j = \sum_j a_j e'_j = \sum_i e_i a'_{(i)}$$

(矩阵<math>A的不同表示形式)

$$a_j = Ae_j = \sum_i a_{ij}e_i$$

$$a_{(i)} = A'e_i = \sum_j a_{ij}e_j$$

$$a_{ij} = e'_i A e_j$$

性质5
$$tr(E'_{rs}A) = a_{rs}$$

事实上,

$$tr(E'_{rs}A) = tr(e_se'_rA)$$

$$= tr(e'_r A e_s) = e'_r A e_s = a_{rs}$$

$$tr(AB) = \sum_{i,j} a_{ij}b_{ji} = [vec(A')]'[vec(B)]$$

定义1.3.2 设A = A'为p阶方阵,令

$$svec(A) = (a_{11}, a_{21}, \dots, a_{p1}, a_{22}, a_{32}, \dots, a_{p2}, \dots, a_{pp})'$$

为 $\frac{p(p+1)}{2}$ 维向量,称svec(A)为对称矩阵的拉

直运算

定义1.3.3 设 $A = (a_{ij})$ 为 $n \times p$ 阶矩阵,B为 $m \times q$ 阶矩阵,则A和B的kronecker积定义为

$$A \otimes B = (a_{ij}B) = \begin{pmatrix} a_{11}B & a_{12}B \cdots a_{1p}B \\ a_{21}B & a_{22}B \cdots a_{2p}B \\ \cdots & \cdots & \cdots \\ a_{n1}B & a_{n2}B \cdots a_{np}B \end{pmatrix}$$

它是 $nm \times pq$ 阶矩阵。

若 α 是任一实数,则

$$(\alpha A) \otimes B = A \otimes (\alpha B) = \alpha (A \otimes B)$$

性质2(分配律)

$$A \otimes (B + C) = A \otimes B + A \otimes C$$

$$(B+C)\otimes A=B\otimes A+C\otimes A$$

性质3(结合律)

$$(A \otimes B) \otimes C = A \otimes (B \otimes C)$$

$$I_{mn} = I_m \otimes I_n = I_n \otimes I_m$$

$$(A \otimes B)' = A' \otimes B'$$

$$(A \otimes B)(C \otimes D) = AC \otimes BD$$

证明:我们用 $[A \otimes B]_{\alpha\beta}$ 表示 $A \otimes B$ 的 (α, β) 块

$$[(A \otimes B)(C \otimes D)]_{\alpha\beta} = \sum_{\gamma=1}^{p} [A \otimes B]_{\alpha\gamma} [C \otimes D]_{\gamma\beta}$$

$$=\sum_{\gamma=1}^{p}a_{\alpha\gamma}B\cdot c_{\gamma\beta}D=\sum_{\gamma=1}^{p}a_{\alpha\gamma}c_{\gamma\beta}BD$$

$$= (AC)_{\alpha\beta} BD = [AC \otimes BD]_{\alpha\beta}$$

所以
$$(A \otimes B)(C \otimes D) = AC \otimes BD$$

若A和B为非退化方阵,则

$$(A \otimes B)^{-1} = A^{-1} \otimes B^{-1}$$

设
$$A: n \times m, B: p \times q, X: m \times p, 则$$

$$vec(AXB) = (B' \otimes A)vec(X)$$

注:这个性质很重要,它建立了拉直运算和 Kronecker积之间的联系。 证明: $\diamondsuit(A)_i$ 和 $(A)_{(i)}$ 分别表示A的第i列和第j行,则

$$(AXB)_k = AXBe_k = A(\sum_{j=1}^p (X)_j e_j')Be_k$$

$$= \sum_{j=1}^{p} A(X)_{j} e'_{j} B e_{k} = \sum_{j=1}^{p} b_{jk} A(X)_{j}$$

$$= (b_{1k}A \ b_{2k}A \cdots b_{pk}A) \begin{pmatrix} (X)_1 \\ (X)_2 \\ \vdots \\ (X)_p \end{pmatrix} = (B)'_k \otimes A) vec(X).$$

$$vec(AXB) = \begin{pmatrix} (AXB)_1 \\ (AXB)_2 \\ \vdots \\ (AXB)_q \end{pmatrix} = \begin{pmatrix} (B)'_1 \otimes A \\ (B)'_2 \otimes A \\ \vdots \\ (B)'_q \otimes A \end{pmatrix} vec(X)$$

$$= (B' \otimes A)vec(X)$$

若A和B均为方阵,则

$$tr(A \otimes B) = tr(A) \cdot tr(B)$$

若x和y为列向量,则

$$xy' = x \otimes y' = y' \otimes x$$

 $若A为n阶方阵,其特征值为{\lambda_i,i=1,2,\dots,n},$ 相应的特征向量为 $\{x_i, i=1,2,\cdots,n\}$, B为m阶方 阵, 其特征值为 $\{\mu_i, i = 1, 2, \dots, m\}$, 相应的特征 向量{ y_i , $i = 1,2,\dots,m$ },则 $A \otimes B$ 的特征值为{ $\lambda_i \mu_i$, $i = 1, 2, \dots, n, i = 1, 2, \dots m$ },相应的特征向量为 $\{x_i \otimes y_i, i = 1, 2, \dots, n, j = 1, 2, \dots, m\}.$

证明:利用Jordan标准形,存在可逆矩阵P和Q,

使得
$$P^{-1}AP = egin{pmatrix} D_1 & 0 & \cdots 0 \\ 0 & D_2 & \cdots 0 \\ \cdots & \cdots & \cdots \\ 0 & 0 & \cdots D_k \end{pmatrix}$$

使得
$$P^{-1}AP = \begin{pmatrix} D_1 & 0 & \cdots & 0 \\ 0 & D_2 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & D_k \end{pmatrix}$$
其中 $D_i = \begin{pmatrix} \lambda_i & 1 & 0 & \cdots & 0 \\ 0 & \lambda_i & 1 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & 0 & \cdots & \lambda_i \end{pmatrix}$

$$Q^{-1}BQ = \begin{pmatrix} C_1 & 0 & \cdots & 0 \\ 0 & C_2 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & C_k \end{pmatrix}$$
其中 $C_i = \begin{pmatrix} \mu_i & 1 & 0 & \cdots & 0 \\ 0 & \mu_i & 1 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & 0 & \cdots & \mu_i \end{pmatrix}$

$$(P^{-1} \otimes Q^{-1})(A \otimes B)(P \otimes Q)$$

$$= \begin{pmatrix} D_1 & 0 & \cdots & 0 \\ 0 & D_2 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & D_k \end{pmatrix} \otimes \begin{pmatrix} C_1 & 0 & \cdots & 0 \\ 0 & C_2 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & C_k \end{pmatrix}$$

仍为上三角形矩阵,对 角线元为 $\lambda_i \mu_j$, $i=1,2,\cdots,n$, $j=1,2,\cdots m$, 所以 $A\otimes B$ 的特征值为 $\{\lambda_i \mu_i, i=1,2,\cdots,n, j=1,2,\cdots m\}$.

$$若Ax = \lambda x$$
, $By = \mu y$,

$$\Rightarrow (A \otimes B)(x \otimes y) = (Ax) \otimes (By)$$

$$= (\lambda x) \otimes (\mu y) = \lambda \mu(x \otimes y),$$

所以

$$\{x_i \otimes y_j, i = 1, 2, \dots, n, j = 1, 2, \dots, m\}$$

为相应的特征向量.

条件同性质11,则

$$\begin{vmatrix} \mathbf{A} \otimes \mathbf{B} \\ \mathbf{n} \times \mathbf{n} \end{vmatrix} = \begin{vmatrix} \mathbf{A} \\ \mathbf{n} \times \mathbf{n} \end{vmatrix}^{m} \begin{vmatrix} \mathbf{B} \\ \mathbf{m} \times \mathbf{m} \end{vmatrix}^{n}$$

§ 1.4 矩阵的微商和变换的雅可比

矩阵微商是通常微商的推广, 是求极大似然估计和最小二乘估计 的工具,利用它也可以方便地求多 变量积分变换的雅可比行列式。

- 一、矩阵对标量的微商
- 二、矩阵的标量函数对矩阵的微商
- 三、向量对向量的微商
- 四、矩阵对矩阵的微商

一、矩阵对标量的微商

定义1.4.1 设 $Y = (y_{ij}(t))$ 是 $p \times q$ 矩阵,它的元素是t的函数,

$$\frac{\partial \{Y\}}{\partial t} = \left(\frac{\partial y_{ij}(t)}{\partial t}\right)$$

称为Y对t的微商.

$$\frac{\partial \{Y\}}{\partial t} = \begin{pmatrix}
\frac{\partial y_{11}(t)}{\partial t} & \frac{\partial y_{12}(t)}{\partial t} & \cdots & \frac{\partial y_{1q}(t)}{\partial t} \\
\frac{\partial y_{21}(t)}{\partial t} & \frac{\partial y_{22}(t)}{\partial t} & \cdots & \frac{\partial y_{2q}(t)}{\partial t} \\
\cdots & \cdots & \cdots \\
\frac{\partial y_{p1}(t)}{\partial t} & \frac{\partial y_{p2}(t)}{\partial t} & \cdots & \frac{\partial y_{pq}(t)}{\partial t}
\end{pmatrix}$$

$$\frac{\partial \{X+Y\}}{\partial t} = \frac{\partial \{X\}}{\partial t} + \frac{\partial \{Y\}}{\partial t}$$

$$\frac{\partial \{XY\}}{\partial t} = \frac{\partial \{X\}}{\partial t} Y + X \frac{\partial \{Y\}}{\partial t}$$

$$\frac{\partial \{X \otimes Y\}}{\partial t} = \frac{\partial \{X\}}{\partial t} \otimes Y + X \otimes \frac{\partial \{Y\}}{\partial t}$$

$$\left(\frac{\partial \{X\}}{\partial t}\right)' = \frac{\partial \{X'\}}{\partial t}$$

$$\frac{\partial \{X\}}{\partial x_{ij}} = E_{ij}$$

$$\frac{\partial \{AXB\}}{\partial x_{ij}} = AE_{ij}B$$

$$\frac{\partial \{X^{-1}\}}{\partial t} = -X^{-1} \frac{\partial \{X\}}{\partial t} X^{-1}$$

(作业)

提示: 利用 $XX^{-1} = I$ 和性质2.

二、矩阵的标量函数对矩阵的微商

定义1.4.2 设y = f(X)是 $m \times n$ 阶矩阵X的 函数,y对X的微商定义为

$$\frac{\partial y}{\partial \{X\}} = \left(\frac{\partial y}{\partial x_{ij}}\right)$$

$$\frac{\partial y}{\partial \{X\}} = \begin{pmatrix}
\frac{\partial y}{\partial x_{11}} & \frac{\partial y}{\partial x_{12}} & \cdots & \frac{\partial y}{\partial x_{1n}} \\
\frac{\partial y}{\partial \{X\}} & \frac{\partial y}{\partial x_{21}} & \frac{\partial y}{\partial x_{22}} & \cdots & \frac{\partial y}{\partial x_{2n}} \\
\cdots & \cdots & \cdots & \cdots \\
\frac{\partial y}{\partial x_{m1}} & \frac{\partial y}{\partial x_{m2}} & \cdots & \frac{\partial y}{\partial x_{mn}}
\end{pmatrix}$$

$$\left(\frac{\partial f(X)}{\partial \{X\}}\right)' = \frac{\partial f(X)}{\partial \{X'\}}$$

若X为n阶方阵,则

$$\frac{\partial tr(X)}{\partial \{X\}} = I_n$$

$$\frac{\partial tr(AXB)}{\partial \{X\}} = A'B',$$

特别
$$\frac{\partial tr(AX)}{\partial \{X\}} = A'$$

证明
$$tr(AXB) = \sum_{i} e'_{i}AXBe_{i}$$

$$\frac{\partial tr(AXB)}{\partial x_{kl}} = \sum_{i} e'_{i} A E_{kl} B e_{i} = \sum_{i} e'_{i} A e_{k} e'_{l} B e_{i}$$

$$= \sum_{i} a_{ik} b_{li} = \sum_{i} b_{li} a_{ik} = (BA)_{lk} = (A'B')_{kl}$$

所以
$$\frac{\partial tr(AXB)}{\partial \{X\}} = A'B'$$

$$\frac{\partial tr(AX)}{\partial \{X\}}$$

$$= \begin{cases} A' \\ A + A' - diag(a_{11}, a_{22}, \dots, a_{nn}) & 若X = X' \end{cases}$$

证明 仅对X为对称矩阵时给出证明即可.

若
$$X = X'$$
,则

$$(1)\frac{\partial tr(AX)}{\partial x_{kk}} = \frac{\partial \sum_{i} e'_{i} AX e_{i}}{\partial x_{kk}} = \sum_{i} e'_{i} AE_{kk} e_{i}$$

$$=\sum_{i}e_{i}'Ae_{k}\underline{e_{k}'e_{i}}=a_{kk}$$

(2)若 $k \neq l$,则

$$\frac{\partial tr(AX)}{\partial x_{kl}} = \frac{\partial \sum_{i} e'_{i} AX e_{i}}{\partial x_{kl}} = \sum_{i} e'_{i} AE_{kl} e_{i} + \sum_{i} e'_{i} AE_{lk} e_{i}$$

$$=\sum_{i}e'_{i}Ae_{k}e'_{\underline{l}}e_{i}+\sum_{i}e'_{i}Ae_{l}e'_{\underline{k}}e_{i}=a_{lk}+a_{kl}$$

所以
$$\frac{\partial tr(AX)}{\partial \{X\}} = A + A' - diag(a_{11}, a_{22}, \dots, a_{nn})$$

$$\frac{\partial x'Ax}{\partial \{x\}} = (A + A')x$$

$$\frac{\partial tr(X'AX)}{\partial \{X\}} = (A + A')X$$

作业

证明
$$x'Ax = tr(x'Ax) = tr(Axx') = \sum_{i} e'_{i}Axx'e_{i}$$

$$\frac{\partial x' A x}{\partial x_l} = \sum_{i} \left(\frac{\partial e_i' A x}{\partial x_l} \right) x' e_i + \sum_{i} e_i' A x \left(\frac{\partial x' e_i}{\partial x_l} \right)$$

$$= \sum_{i} e'_{i} A e_{l} x' e_{i} + \sum_{i} e'_{i} A x e'_{l} e_{i} = \sum_{i} a_{il} x_{i} + e'_{l} A x$$

$$\frac{\partial \{XY\}}{\partial t} = \frac{\partial \{X\}}{\partial t} Y + X \frac{\partial \{Y\}}{\partial t}$$

$$\sum_{i} a_{il} x_{i} + e'_{l} A x$$

$$= e'_{l} A' x + e'_{l} A x$$

$$= e'_{l} (A' + A) x = ((A' + A) x)_{l}$$

所以
$$\frac{\partial x'Ax}{\partial \{x\}} = (A+A')x$$

$$\frac{\partial a'x}{\partial \{x\}} = a$$

$$\frac{\partial \det(X)}{\partial \{X\}} = (X')^{-1} \det(X)$$

证明
$$\det(X) = \sum_{i} x_{ij} X_{ij}$$
, 其中 X_{ij} 是 x_{ij} 的

代数余子式. 易见
$$\frac{\partial \det(X)}{\partial x_{ij}} = X_{ij}$$

$$X \quad X^{-1} = \frac{1}{\det(X)} \begin{pmatrix} X_{11} & X_{21} & \cdots & X_{n1} \\ X_{12} & X_{22} & \cdots & X_{n2} \\ & & & & \\ X_{1n} & X_{2n} & \cdots & X_{nn} \end{pmatrix}$$

所以
$$\frac{\partial \det(X)}{\partial \{X\}} = (X')^{-1} \det(X)$$

三、向量对向量的微商

定义1.4.3 设x为n维向量, $y = (y_1, y_2, \dots, y_m)'$

=
$$(f_1(x), f_2(x), \dots, f_m(x))'$$

则y对x的微商定义为

$$\frac{\partial y'}{\partial x} = \left(\frac{\partial y_j}{\partial x_i}\right)$$

$$\frac{\partial y_1}{\partial x_1} \qquad \frac{\partial y_2}{\partial x_1} \qquad \cdots \qquad \frac{\partial y_m}{\partial x_1}$$

$$\frac{\partial y'}{\partial x} = \begin{bmatrix} \frac{\partial y_1}{\partial x_2} & \frac{\partial y_2}{\partial x_2} & \cdots & \frac{\partial y_m}{\partial x_2} \\ \cdots & \cdots & \cdots \\ \frac{\partial y_1}{\partial x_n} & \frac{\partial y_2}{\partial x_n} & \cdots & \frac{\partial y_m}{\partial x_n} \end{bmatrix}$$

若
$$y = Ax$$
,则 $\frac{\partial y'}{\partial x} = A'$

证明:
$$y_j = \sum_k a_{jk} x_k$$
 $j = 1, 2, \dots, m$.

$$\frac{\partial y_j}{\partial x_i} = a_{ji} = (A')_{ij}$$

所以
$$\frac{\partial y'}{\partial x} = A'$$

若
$$Y = AXB$$
,则

$$\frac{\partial (vec(Y))'}{\partial vec(X)} = B \otimes A'$$

证明

$$vec(Y) = (B' \otimes A)vec(X)$$

由性质1立即得证.

四、矩阵对矩阵的微商

定义1.4.4 设X是 $m \times n$ 阶矩阵,F(X)是 $p \times q$

阶矩阵,记
$$F(X) = (f_{ij}(X)) = (f_{ij})$$
则

$$\frac{\partial F(X)}{\partial X} = \frac{\partial [vec(F(X))]'}{\partial vec(X)}$$

称为F(X)对X的微商。

(依行的先后次序)第一行的元素

$$\begin{pmatrix}
\frac{\partial f_{11}}{\partial x_{11}} & \frac{\partial f_{11}}{\partial x_{21}} & \dots & \frac{\partial f_{11}}{\partial x_{m1}} \\
\frac{\partial f_{11}}{\partial x_{12}} & \frac{\partial f_{11}}{\partial x_{22}} & \dots & \frac{\partial f_{11}}{\partial x_{m2}} \\
\dots & \dots & \dots \\
\frac{\partial f_{11}}{\partial x_{1n}} & \frac{\partial f_{11}}{\partial x_{2n}} & \dots & \frac{\partial f_{11}}{\partial x_{mn}}
\end{pmatrix}$$

(依行的先后次第一列的元素

$$\begin{pmatrix}
\frac{\partial f_{12}}{\partial x_{12}} & \frac{\partial f_{22}}{\partial x_{12}} & \cdots & \frac{\partial f_{p^2}}{\partial x_{12}} \\
\frac{\partial f_{12}}{\partial x_{22}} & \frac{\partial f_{22}}{\partial x_{22}} & \cdots & \frac{\partial f_{p^2}}{\partial x_{22}} \\
\cdots & \cdots & \cdots \\
\frac{\partial f_{12}}{\partial x_{m^2}} & \frac{\partial f_{22}}{\partial x_{m^2}} & \cdots & \frac{\partial f_{p^2}}{\partial x_{m^2}}
\end{pmatrix}$$

若将其分块,每块均为m行 每块均为m行 p列的矩阵, 则第(2,2)个分 块矩阵。

若X是 $m \times n$ 阶矩阵,则

$$\frac{\partial X}{\partial X} = \frac{\partial (vec(X))'}{\partial (vec(X))} = I_{mn}$$

若F(X)是 $p \times q$ 阶矩阵,G(X)是 $q \times r$ 阶矩阵,

X是 $m \times n$ 阶矩阵,则

$$\frac{\partial F(X)G(X)}{\partial X}$$

$$= \frac{\partial F(X)}{\partial X} (G(X) \otimes I_p) + \frac{\partial G(X)}{\partial X} (I_r \otimes F'(X))$$

$$\frac{\partial (AXB)}{\partial X} = B \otimes A'$$

可以先拉直,再求导.

证明:利用性质2

$$\frac{\partial (AXB)}{\partial X} = \frac{\partial (AX)}{\partial X} (B \otimes I) + \frac{\partial B}{\partial X} (I \otimes X'A')$$

$$= \left[\frac{\partial(A)}{\partial X}(X \otimes I) + \frac{\partial X}{\partial X}(I \otimes A')\right](B \otimes I)$$

$$= B \otimes A'$$

$$\frac{\partial X^{-1}}{\partial X} = -[X^{-1} \otimes (X^{-1})']$$

证明 由
$$XX^{-1} = I$$
,及性质2

$$\frac{\partial (XX^{-1})}{\partial X} = \frac{\partial X}{\partial X}(X^{-1} \otimes I) + \frac{\partial X^{-1}}{\partial X}(I \otimes X')$$

$$= (X^{-1} \otimes I) + \frac{\partial X^{-1}}{\partial X} (I \otimes X') = 0$$

所以
$$\frac{\partial X^{-1}}{\partial X} = -(X^{-1} \otimes I)(I \otimes (X^{-1})')$$

$$=-[X^{-1}\otimes (X^{-1})']$$

五、复合函数求导的公式

定理1.4.1(复合函数求导公式)设 $\Psi(X)$ 是矩阵变量X的数值函数,X的每个元素 x_{ij} 均为变量t的函数,则

$$\frac{\partial \psi(X)}{\partial t} = tr \left[\frac{\partial \{X\}}{\partial t} \left(\frac{\partial \psi}{\partial \{X\}} \right)' \right]$$

证明: 由复合函数求导公式得到

$$\frac{\partial \psi(X)}{\partial t} = \sum_{j=1}^{n} \sum_{i=1}^{m} \frac{\partial \psi(X)}{\partial x_{ij}} \frac{\partial x_{ij}}{\partial t}$$

$$= tr \left[\frac{\partial \{X\}}{\partial t} \left(\frac{\partial \psi}{\partial \{X\}} \right)' \right]$$

定理1.4.2(矩阵的复合函数求导公式)设F(G(X))是两个矩阵函数的复合函数,则

$$\frac{\partial F(G(X))}{\partial X} = \frac{\partial G(X)}{\partial X} \frac{\partial F(G)}{\partial G}$$

$$\frac{\partial \det(X(t))}{\partial t} = \det(X)tr\left(X^{-1}\frac{\partial\{X\}}{\partial t}\right)$$

证明:

$$\frac{\partial \det(X(t))}{\partial t} = tr \left| \frac{\partial \{X\}}{\partial t} \left(\frac{\partial \det(X)}{\partial \{X\}} \right)' \right|$$

$$= tr \left[\frac{\partial \{X\}}{\partial t} ((X')^{-1} \det(X))' \right]$$

$$= tr \left[\frac{\partial \{X\}}{\partial t} \left(X^{-1} \det(X) \right) \right] = \det(X) tr \left(X^{-1} \frac{\partial \{X\}}{\partial t} \right)$$

$$\frac{\partial \ln \det(X)}{\partial t} = tr \left(X^{-1} \frac{\partial \{X\}}{\partial t} \right)$$

证明:

$$\frac{\partial \ln \det(X)}{\partial t} = \frac{1}{\det(X)} \frac{\partial \det(X)}{\partial t}$$

$$= tr \left(X^{-1} \frac{\partial \{X\}}{\partial t} \right)$$

六、雅可比(Jacobi)行列式的计算

$$\int_{D} g(x_{1}, x_{2}, \dots, x_{n}) dx_{1} dx_{2} \cdots dx_{n} \qquad D \subset \mathbb{R}^{n}$$

$$= \int_T g(f^{-1}(y)) |J(x \to y)| dy$$

式中
$$T = \{y \mid y = f(x), x \in D\}, |J(x \to y)| = \left| \frac{\partial x'}{\partial y} \right|_{+}$$

f是一一变换,其中 $|A|_{+}$ 表示A的行列式的绝对值.

定义1.4.5 设 $X \in R^{m \times n}$ 是矩阵变量,

$$Y = F(X) \in \mathbb{R}^{m \times n}$$
是一一变换,

$$F(X)$$
可微,则变换 $Y = F(X)$ 的

雅可比行列式定义为

$$J(Y \to X) = J(Y : X) = \left| \frac{\partial Y}{\partial X} \right|_{+}$$

(一)、当矩阵变量为一般矩阵的情况

性质1

若Y = AXB, 且A为m阶非奇异方阵,

B为n阶非奇异方阵,则有

$$J(Y \to X) = \left| (\det(A))^n (\det(B))^m \right| = \left| A \right|_+^n \left| B \right|_+^m$$

特别
$$y = Ax, J(y \rightarrow x) = |A|_+$$

证明: 因为 $vec(Y) = (B' \otimes A)vec(X)$

所以

$$J(Y \to X) = \left| \frac{\partial Y}{\partial X} \right|_{+} = \left| \frac{\partial [vec(Y)]'}{\partial vec(X)} \right|_{+}$$

$$= |B \otimes A'|_{+} = |B|_{+}^{m} |A'|_{+}^{n} = |A|_{+}^{n} |B|_{+}^{m}$$

若X为n阶可逆方阵,作变换 $Y = X^{-1}$,

则有

$$J(Y \to X) = |X|^{-2n}$$

证明:由矩阵对矩阵的微商中的性质4知,

$$\frac{\partial X^{-1}}{\partial X} = -[X^{-1} \otimes (X^{-1})']$$

所以有

$$J(Y \to X) = \left| \frac{\partial Y}{\partial X} \right|_{+} = \left| -\left(X^{-1} \otimes (X^{-1})' \right) \right|_{+}$$

$$= |X^{-1}|_{+}^{n} |(X^{-1})'|_{+}^{n} = |X|^{-2n}$$

(二)、当矩阵变量为三角矩阵的情况

性质3

设G为给定的n阶非奇异下三角矩阵,

X为n阶下三角矩阵变量,令Y = GX,

则有
$$J(Y \to X) = \left| \prod_{i=1}^{n} g_{ii}^{i} \right|$$
, 其中 g_{11}, g_{22}, \dots ,

 g_{nn} 是G的主对角元.

$$Y = GX$$

$$J(Y \to X) = \left| \prod_{i=1}^{n} g_{ii}^{i} \right|$$

条件同性质3, 令变换Y = XG, 则有

$$\mathbf{J}(Y \to X) = \left| \prod_{i=1}^n \mathbf{g}_{ii}^{n-i+1} \right|$$

设X为下三角矩阵变量,令变换

$$Y = XX'$$
,则有

$$J(Y \to X) = 2^n \left| \prod_{i=1}^n x_{ii}^{n-i+1} \right|$$

注意:若要求X的对角线元素非负,则

$$Y = XX'$$
是一一变换.

类似对X为上三角矩阵时,G 为非奇异上三角矩阵,有类似性 质3、性质4和性质5的结论。

(三)、当矩阵变量为对称矩阵的情况

性质6

设
$$X' = X$$
, G 是非奇异上三角矩阵, $n \times n$

$$J(Y \to X) = \left| G \right|_{+}^{n+1}$$

设X' = X,G是非奇异下三角矩阵,

令
$$Y = G'XG$$
,则有

$$J(Y \to X) = \left| G \right|_{+}^{n+1}$$

设X' = X,P是非奇异矩阵,令

$$Y = P'XP$$
,则有

$$\mathbf{J}(Y \to X) = \left| P \right|_{+}^{n+1}$$

作业

- 1、证明AB和BA有相同的非零特征根。
- 2、设P为p阶非奇异矩阵,U为 $p \times q$ 矩阵,V为 $q \times p$ 矩阵,而p阶方阵Q=P+UV和q阶方阵 $I_q+VP^{-1}U$ 也非奇异,则 $Q^{-1}=(P+UV)^{-1}$

$$= P^{-1} - P^{-1}U(I_q + VP^{-1}U)^{-1}VP^{-1}$$

特别
$$(P+xy')^{-1}=P^{-1}-(1+y'P^{-1}x)^{-1}P^{-1}xy'P^{-1}$$

3、若A>0(或 ≥ 0),则存在B>0(或 ≥ 0), 使得 $A=B^2$,称B为A的平方根矩阵,记 为 $B=A^{\frac{1}{2}}$.

4、证明加号逆的性质5,即

$$(A^{T}A)^{+} = A^{+}(A^{+})^{T}$$

、若A>0,将A剖分为

$$A = \begin{pmatrix} A_{11} & A_{12} \\ & & \\ A_{21} & A_{22} \end{pmatrix}.$$

其中411为方阵,则

$$A_{11} > 0$$
, $A_{22} > 0$, $A_{11\cdot 2} > 0$, $A_{22\cdot 1} > 0$.

6、证明:

$$\frac{\partial \{X^{-1}\}}{\partial t} = -X^{-1} \frac{\partial \{X\}}{\partial t} X^{-1}$$