

Homework 1

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1. Prove the following properties of matrix rank:

- (a) $\text{rank}(A_{p \times q}) + \text{rank}(B_{q \times r}) - q \leq \text{rank}(AB) \leq \min\{\text{rank}(A), \text{rank}(B)\}$, i.e. $\text{rank}(AB) = \text{rank}(A)$, if $\text{rank}(B) = q$

Solution:

$$\begin{aligned}
 (1) \quad \text{rank}(A) + \text{rank}(B) &= \text{rank} \begin{pmatrix} A & O \\ O & B \end{pmatrix} \leq \text{rank} \begin{pmatrix} A & O \\ I_q & B \end{pmatrix} \\
 &= \text{rank} \begin{pmatrix} A & O \\ I_q & B \end{pmatrix} \begin{pmatrix} I_q & -B \\ O & I_r \end{pmatrix} \\
 &= \text{rank} \begin{pmatrix} I_p & -A \\ O & I_q \end{pmatrix} \begin{pmatrix} A & -AB \\ I_q & O \end{pmatrix} \\
 &= \text{rank} \begin{pmatrix} O & -AB \\ I_q & O \end{pmatrix} = \text{rank}(AB) + q
 \end{aligned}$$

$$(2) \quad L_{\text{col}}(AB) \subset L_{\text{col}}(A), \quad L_{\text{row}}(AB) \subset L_{\text{row}}(B)$$

- (b) $\text{rank}(AB) + \text{rank}(BC) \leq \text{rank}(B) + \text{rank}(ABC)$

Solution:

$$\begin{aligned}
 \text{rank}(AB) + \text{rank}(BC) &= \text{rank} \begin{pmatrix} AB & O \\ O & BC \end{pmatrix} \leq \text{rank} \begin{pmatrix} AB & O \\ B & BC \end{pmatrix} \\
 &= \text{rank} \begin{pmatrix} AB & O \\ B & BC \end{pmatrix} \begin{pmatrix} I & -C \\ O & I \end{pmatrix}
 \end{aligned}$$

$$\begin{aligned}
&= \text{rank} \begin{pmatrix} I & -A \\ O & I \end{pmatrix} \begin{pmatrix} AB & -ABC \\ B & O \end{pmatrix} \\
&= \text{rank} \begin{pmatrix} O & B \\ ABC & O \end{pmatrix} \\
&= \text{rank}(B) + \text{rank}(ABC)
\end{aligned}$$

(c) $\text{rank}(A + B) \leq \text{rank}(A) + \text{rank}(B)$

Solution:

$$L_{\text{col}}(A + B) \subset \text{Span}\{L_{\text{col}}(A) \cup L_{\text{col}}(B)\}$$

2. Prove the following properties of matrix trace:

(a) $\text{tr}(AB) = \text{tr}(BA)$

Solution:

$$\text{tr}(AB) = \sum_i (AB)_{ii} = \sum_i \sum_j a_{ij} a_{ji} = \sum_j \sum_i b_{ji} a_{ij} = \sum_j (BA)_{jj} = \text{tr}(BA).$$

(b) $\text{tr}(A) = \sum \lambda_i$, where λ_i are eigenvalues of A , if A is real symmetric.

Solution:

Since A is symmetric, it can be decomposed as $A = P\Lambda P^T$, where P is an orthogonal matrix satisfying $PP^T = P^T P = I$ and $\Lambda = \text{diag}(\lambda_i)$. Then we have $\text{tr}(A) = \text{tr}(P\Lambda P^T) = \text{tr}(P^T P \Lambda) = \text{tr}(\Lambda) = \sum \lambda_i$.

3. Let A be a $n \times n$ such that $A^T A = A^2$. Show that A is symmetric.

Solution:

$$\text{tr}[(A - A^T)^T(A - A^T)] = \text{tr}(A^T A - AA - AA^T + AA^T) = 0.$$

4. Calculate the Vandermonde determinant.

$$V(x_1, \dots, x_n) = \begin{vmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-2} & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-2} & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-2} & x_n^{n-1} \end{vmatrix}$$

Solution:

Starting from the left column, we minus x_1 multiply the column left to it.

$$\begin{aligned} V(x_1, \dots, x_n) &= \begin{vmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{n-2} & x_1^{n-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{n-2} & x_2^{n-1} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & x_n & x_n^2 & \cdots & x_n^{n-2} & x_n^{n-1} \end{vmatrix} \\ &= \begin{vmatrix} 1 & 0 & 0 & \cdots & 0 & 0 \\ 1 & x_2 - x_1 & x_2^2 - x_1x_2 & \cdots & x_2^{n-2} - x_1x_2^{n-3} & x_2^{n-1} - x_1x_2^{n-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 1 & x_n - x_1 & x_n^2 - x_1x_n & \cdots & x_n^{n-2} - x_1x_n^{n-3} & x_n^{n-1} - x_1x_n^{n-2} \end{vmatrix} \\ &= \begin{vmatrix} x_2 - x_1 & x_2^2 - x_1x_2 & \cdots & x_2^{n-2} - x_1x_2^{n-3} & x_2^{n-1} - x_1x_2^{n-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ x_n - x_1 & x_n^2 - x_1x_n & \cdots & x_n^{n-2} - x_1x_n^{n-3} & x_n^{n-1} - x_1x_n^{n-2} \end{vmatrix} \\ &= (x_2 - x_1)(x_3 - x_1) \cdots (x_n - x_1) \begin{vmatrix} 1 & x_2 & \cdots & x_2^{n-3} & x_2^{n-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & \cdots & x_n^{n-3} & x_n^{n-2} \end{vmatrix} \\ &= \prod_{j>1} (x_j - x_1) V(x_2, \dots, x_n) \\ &= \prod_{1 \leq i < j \leq n} (x_i - x_j) \end{aligned}$$

5. (Sherman-Morrison formula) Suppose the inverse of $A = A_{n \times n}$ is known, now we perturb each element of the matrix by $A_{ij} \rightarrow A_{ij} + \Delta x_i \Delta y_j = B_{ij}$, denote the new matrix by $B = \{B_{ij}\}$. Calculate the inverse of matrix B .

Solution:

Write $B = A + \mathbf{u}\mathbf{v}^T$, where \mathbf{u}, \mathbf{v} are $n \times 1$ vectors where $u_i = \Delta x_i, v_i = \Delta y_i$. The Sherman-Morrison formula provides a quick way to compute the inverse of matrix B :

$$B^{-1} = (A + \mathbf{u}\mathbf{v}^T)^{-1} = A^{-1} - \frac{A^{-1}\mathbf{u}\mathbf{v}^T A^{-1}}{1 + \mathbf{v}^T A^{-1}\mathbf{u}}$$

6. Consider the equicorrelation matrix

$$A = \begin{pmatrix} 1 & \rho & \cdots & \rho \\ \rho & 1 & \cdots & \rho \\ \vdots & \vdots & \ddots & \vdots \\ \rho & \rho & \cdots & 1 \end{pmatrix}$$

Find the explicit expressions of $\det(A)$ and A^{-1} .

Solution:

We can write A as follows:

$$A = (1 - \rho)I_n + \rho \mathbf{1}_n \mathbf{1}'_n = (1 - \rho) \left(I_n + \frac{\rho}{1 - \rho} \mathbf{1}_n \mathbf{1}'_n \right).$$

Therefore, the determinant of A is

$$\det(A) = (1 - \rho)^n \det \left(I_n + \frac{\rho}{1 - \rho} \mathbf{1}_n \mathbf{1}'_n \right) = (1 - \rho)^n \det \left(1 + \frac{\rho}{1 - \rho} \mathbf{1}'_n \mathbf{1}_n \right) = (1 - \rho)^n \left(1 + \frac{n\rho}{1 - \rho} \right).$$

The inverse of A is:

$$\begin{aligned} A^{-1} &= (1 - \rho)^{-1} \left\{ I_n^{-1} - I_n^{-1} \mathbf{1}_n \mathbf{1}'_n I_n^{-1} / \left(\frac{1 - \rho}{\rho} + \mathbf{1}'_n I_n^{-1} \mathbf{1}_n \right) \right\} \\ &= \frac{1}{1 - \rho} I_n - \frac{\rho}{(1 - \rho)^2 + \rho(1 - \rho)n} \mathbf{1}_n \mathbf{1}'_n. \end{aligned}$$