

# 矩阵基本知识

- 矩阵基本运算
- 线性相关性
- 矩阵初等变换
- 随机向量与随机矩阵
- 矩阵直和
- 矩阵Hardmard积
- 矩阵Kronecker积

# 矩阵基本运算

■ 矩阵共轭转置

■ 矩阵加法

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & \cdots & a_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} \end{pmatrix}^H = \begin{pmatrix} a_{11}^* & a_{21}^* & \cdots & a_{M1}^* \\ a_{12}^* & a_{22}^* & \cdots & a_{M2}^* \\ \cdots & \cdots & \cdots & \cdots \\ a_{1N}^* & a_{2N}^* & \cdots & a_{MN}^* \end{pmatrix}$$

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & \cdots & a_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} \end{pmatrix} + \begin{pmatrix} b_{11} & b_{12} & \cdots & b_{1N} \\ b_{21} & b_{22} & \cdots & b_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ b_{M1} & b_{M2} & \cdots & b_{MN} \end{pmatrix}$$

$$= \begin{pmatrix} a_{11} + b_{11} & a_{12} + b_{12} & \cdots & a_{1N} + b_{1N} \\ a_{21} + b_{21} & a_{22} + b_{22} & \cdots & a_{2N} + b_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ a_{M1} + b_{M1} & a_{M2} + b_{M2} & \cdots & a_{MN} + b_{MN} \end{pmatrix}$$

# 矩阵基本运算

## ■ 矩阵标量乘法

## ■ 矩阵乘法

$$\alpha \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & \cdots & a_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} \end{pmatrix} = \begin{pmatrix} \alpha a_{11} & \alpha a_{12} & \cdots & \alpha a_{1N} \\ \alpha a_{21} & \alpha a_{22} & \cdots & \alpha a_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ \alpha a_{M1} & \alpha a_{M2} & \cdots & \alpha a_{MN} \end{pmatrix}$$

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & \cdots & a_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} \end{pmatrix} \begin{pmatrix} b_{11} & b_{12} & \cdots & b_{1N} \\ b_{21} & b_{22} & \cdots & b_{2K} \\ \cdots & \cdots & \cdots & \cdots \\ b_{N1} & b_{N2} & \cdots & b_{NK} \end{pmatrix} = \begin{pmatrix} \sum_{k=1}^N a_{1k} b_{k1} & \sum_{k=1}^N a_{1k} b_{k2} & \cdots & \sum_{k=1}^N a_{1k} b_{kK} \\ \sum_{k=1}^N a_{2k} b_{k1} & \sum_{k=1}^N a_{2k} b_{k2} & \cdots & \sum_{k=1}^N a_{2k} b_{kK} \\ \cdots & \cdots & \cdots & \cdots \\ \sum_{k=1}^N a_{Mk} b_{k1} & \sum_{k=1}^N a_{Mk} b_{k2} & \cdots & \sum_{k=1}^N a_{Mk} b_{kK} \end{pmatrix}$$

# 矩阵基本运算

- **加法交换律：**  $A + B = B + A$
- **乘法结合律：**  $(AB)C = A(BC)$
- **乘法分配律：**  $(A + B)C = AC + BC$   
 $A(B + C) = AB + AC$
- **逆矩阵：**  $AB = BA = I$ ;
- **幂等矩阵：**  $AA = A$ ；幂等矩阵例子？
- **幂单矩阵：**  $AA = I$ ；幂单矩阵例子？
- **幂零矩阵：**  $AA = 0$ ；

# 矩阵基本运算

- **幂等矩阵**:  $AA = A$ ; 幂等矩阵例子?
  - **A 可逆么? 如果可逆,  $A = ?$**
  - **如果A是Hermitian矩阵, 那么 $A = ?$**
- **幂单矩阵**:  $AA = I$ ; 幂单矩阵例子?
  - **A 可逆么?**
  - **如果A是Hermitian矩阵, 那么 $A = ?$**
- **幂零矩阵**:  $AA = 0$ ;
  - **非零 的幂零矩阵?**

# 矩阵函数

- 考虑矩阵的Taylor展开

$$f(x) = \sum_{n=0}^{+\infty} \frac{f^{(n)}(0)}{n!} x^n$$

- 有如下的矩阵函数定义

$$f(A) = \sum_{n=0}^{+\infty} \frac{f^{(n)}(0)}{n!} A^n$$

- 举例:

$$e^A = \sum_{n=0}^{+\infty} \frac{A^n}{n!} \quad e^{At} = \sum_{n=0}^{+\infty} \frac{A^n}{n!} t^n \quad \ln(I + A) = \sum_{n=0}^{+\infty} \frac{(-1)^{n-1}}{n} A^n$$

# 矩阵线性相关性

## ■ 线性无关，线性相关

$$\sum_{n=1}^N x_n \begin{pmatrix} a_{1n} \\ a_{2n} \\ \dots \\ a_{Mn} \end{pmatrix} = 0 \Leftrightarrow x_n = 0, \forall 1 \leq n \leq N.$$

## ■ 线性无关，线性方程

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \dots & \dots & \dots & \dots \\ a_{M1} & a_{M2} & \dots & a_{MN} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_N \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \dots \\ 0 \end{pmatrix} \Leftrightarrow \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_N \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \dots \\ 0 \end{pmatrix}$$

# 初等行变换

## ■ 初等行变换

- 交换矩阵任意两行

- 一行乘以非零常数

- 某行乘以非零常数，加到另外一行

## ■ 行等价矩阵：矩阵A通过行初等变换→矩阵B

## ■ 初等行变换的矩阵表示

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & a & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & a & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$



# 随机向量

- 向量每个元素为随机变量

$$\bar{x} = (x_1 \quad x_2 \quad \dots \quad x_N)^T$$

- 随机向量期望

$$E\bar{x} = (Ex_1 \quad Ex_2 \quad \dots \quad Ex_N)^T$$

- 随机向量自相关与协方差

$$R_x = E[\bar{x}\bar{x}^H] \quad \text{cov}(\bar{x}, \bar{x}) = E[(\bar{x} - E\bar{x})(\bar{x} - E\bar{x})^H]$$

$$R_{xy} = E[\bar{x}\bar{y}^H] \quad \text{cov}(\bar{x}, \bar{y}) = E[(\bar{x} - E\bar{x})(\bar{y} - E\bar{y})^H]$$

# 高斯随机向量

- 高斯随机变量

$$\bar{x} = (x_1 \quad x_2 \quad \dots \quad x_N)^T$$

- 随机向量期望

$$f(\bar{x}) = \frac{1}{(2\pi)^{N/2} |\Gamma|^{1/2}} e^{-\frac{(\bar{x}-\bar{\mu})^T \Gamma^{-1} (\bar{x}-\bar{\mu})}{2}}$$

- 高斯随机的线性变换，仍然为高斯随机向量

$$\bar{y} = A\bar{x} = A(x_1 \quad x_2 \quad \dots \quad x_N)^T$$

# 随机矩阵

- 矩阵每个元素为随机变量

- 随机矩阵期望

  - 每个元素的期望

- 随机矩阵举例

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \dots & \dots & \dots & \dots \\ a_{M1} & a_{M2} & \dots & a_{MN} \end{pmatrix}$$

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \dots & \dots & \dots & \dots \\ a_{M1} & a_{M2} & \dots & a_{MN} \end{pmatrix} \in \{0,1\}^{M \times N} \quad A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1N} \\ a_{21} & a_{22} & \dots & a_{2N} \\ \dots & \dots & \dots & \dots \\ a_{M1} & a_{M2} & \dots & a_{MN} \end{pmatrix} \in [0, 1]^{M \times N}$$

# 随机矩阵

## ■ 随机矩阵的秩

### ■ 随机矩阵满秩的概率？

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & \cdots & a_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} \end{pmatrix} \in GF(q)^{M \times N}$$

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & \cdots & a_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} \end{pmatrix} \in \{0,1\}^{M \times N}$$

$$A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1N} \\ a_{21} & a_{22} & \cdots & a_{2N} \\ \cdots & \cdots & \cdots & \cdots \\ a_{M1} & a_{M2} & \cdots & a_{MN} \end{pmatrix} \in [0,1]^{M \times N}$$

# 矩阵直和

## ■ 矩阵直和定义

$$[A]_{m \times m} \oplus [B]_{n \times n} = \begin{bmatrix} A & 0 \\ 0 & B \end{bmatrix}$$

## ■ 矩阵直和性质

$$(A \pm B) \oplus (C \pm D) = A \oplus C \pm B \oplus D$$

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

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

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

# 矩阵Hardmard积

- 矩阵Hardmard积定义  $[A * B]_{ij} = a_{ij}b_{ij}$

$$[A * B] = \begin{pmatrix} a_{11}b_{11} & a_{12}b_{12} & \dots & a_{1N}b_{1N} \\ a_{21}b_{21} & a_{22}b_{22} & \dots & a_{2N}b_{2N} \\ \dots & \dots & \dots & \dots \\ a_{M1}b_{M1} & a_{M2}b_{M2} & \dots & a_{MN}b_{MN} \end{pmatrix}$$

- 矩阵Hardmard积性质

$$A * B = B * A$$

$$(A * B) * C = A * (B * C)$$

$$A * (B \pm C) = A * B \pm A * C$$

# 矩阵Kronecker积

■ 矩阵Kronecker积定义

$$A \otimes B = \begin{pmatrix} a_{11}B & a_{12}B & \dots & a_{1N}B \\ a_{21}B & a_{22}B & \dots & a_{2N}B \\ \dots & \dots & \dots & \dots \\ a_{M1}B & a_{M2}B & \dots & a_{MN}B \end{pmatrix}$$

■ 矩阵Kronecker积性质

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

$$AB \otimes CD = (A \otimes C)(B \otimes D) \quad A^{-1} \otimes B^{-1} = (A \otimes B)^{-1}$$

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

$$\text{rank}(A \otimes B) = \text{rank}(A)\text{rank}(B)$$

# 矩阵Kronecker积

## ■ 矩阵Kronecker积性质证明

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

$$= \begin{pmatrix} a_{11}C & a_{12}C & \dots & a_{1N}C \\ a_{21}C & a_{22}C & \dots & a_{2N}C \\ \dots & \dots & \dots & \dots \\ a_{M1}C & a_{M2}C & \dots & a_{MN}C \end{pmatrix} \begin{pmatrix} b_{11}D & b_{12}D & \dots & b_{1N}D \\ b_{21}D & b_{22}D & \dots & b_{2N}D \\ \dots & \dots & \dots & \dots \\ b_{N1}D & b_{N2}D & \dots & b_{NK}D \end{pmatrix}$$
$$= (AB) \otimes (CD)$$

## ■ 矩阵分析的语言