

Matrices and Vectors

$$\underset{m \times n}{A} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & & \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}$$

m rows
n columns

Special Cases

$$\text{column vector } \underset{m \times 1}{\underline{a}} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_m \end{bmatrix}$$

$$\text{row vector } \underset{1 \times n}{\underline{b}} = [b_1, b_2, \dots, b_n]$$

Product of Two Matrices

$$C = A B$$

$$\begin{bmatrix} C_{11} & C_{12} & \dots & C_{1e} \\ C_{21} & C_{22} & \dots & C_{2e} \\ \vdots & & & \\ C_{n1} & C_{n2} & \dots & C_{ne} \end{bmatrix} = \underbrace{\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & & & \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{bmatrix}}_{A_{n \times m}} \underbrace{\begin{bmatrix} b_{11} & b_{12} & \dots & b_{1e} \\ b_{21} & b_{22} & \dots & b_{2e} \\ \vdots & & & \\ b_{m1} & b_{m2} & \dots & b_{me} \end{bmatrix}}_{B_{m \times e}}$$

$$C_{ij} = \sum_{k=1}^m a_{ik} b_{kj}$$

Example

$$\begin{bmatrix} 2 & -1 & 6 \\ 3 & 4 & 2 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 3 & 4 \\ 2 & -2 \end{bmatrix} = \begin{bmatrix} 11 & -18 \\ 19 & 9 \end{bmatrix}$$

Transpose of Matrix

$$A_{m \times n} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & & & \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}$$

$$A_{n \times m}^T = \begin{bmatrix} a_{11} & a_{21} & \cdots & a_{m1} \\ a_{12} & a_{22} & \cdots & a_{m2} \\ \vdots & \vdots & & \vdots \\ a_{1n} & a_{2n} & \cdots & a_{mn} \end{bmatrix}$$

$$\underline{a} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} \Rightarrow \underline{a}^T = [a_1 \ a_2 \ \cdots \ a_n]$$

Dot Product of Two Vectors (a.k.a. inner product)

$$\underline{a} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}, \quad \underline{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

$$\begin{aligned} \underline{a} \cdot \underline{b} &= \underline{a}^T \underline{b} = [a_1 \ a_2 \ \cdots \ a_n] \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix} \\ &= \sum_{k=1}^n a_k b_k \end{aligned}$$

Determinant

- only defined for square matrix, e.g. $A_{n \times n}$
- Notation: $\det(A)$ or $|A|$

Case $n = 2$

$$A = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \quad \det(A) = |A| = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11} a_{22} - a_{12} a_{21}$$

Case n=3

$$A = \begin{bmatrix} a_{11}^+ & a_{12}^- & a_{13}^+ \\ a_{21}^- & a_{22}^+ & a_{23}^- \\ a_{31}^+ & a_{32}^- & a_{33}^+ \end{bmatrix}$$

$$\det(A) = a_{11} \underbrace{\begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix}}_{\text{Co-factor of } a_{11}} + a_{12} \underbrace{\left(- \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} \right)}_{\text{Co-factor of } a_{12}} + a_{13} \underbrace{\begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}}_{\text{Co-factor of } a_{13}}$$

Example

$$A = \begin{bmatrix} 7 & 2 & 1 \\ 0 & 3 & -1 \\ -3 & 4 & -2 \end{bmatrix}, \quad |A| = 7 \cdot (-2) + 2 \cdot (3) + 1 \cdot (9) = 1$$

Matrix of Co-Factors

$$A = \begin{bmatrix} 7 & 2 & 1 \\ 0 & 3 & -1 \\ -3 & 4 & -2 \end{bmatrix}, \quad C = \begin{bmatrix} -2 & 3 & 9 \\ 8 & -11 & -34 \\ -5 & 7 & 21 \end{bmatrix}$$

↑
Co-factor matrix

Adjoint of Matrix

$$\text{adj}(A) = C^T = \begin{bmatrix} -2 & 8 & -5 \\ 3 & -11 & 7 \\ 9 & -34 & 21 \end{bmatrix}$$

Inverse of Matrix

- well-defined only for square matrix A with $\det(A) \neq 0$.

- A^{-1} satisfies : $A^{-1}A = AA^{-1} = I$

$$I_{n \times n} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ 0 & 0 & \cdots & 1 \end{bmatrix}_{n \times n}$$

Identity
matrix

Result

$$A^{-1} = \frac{1}{|A|} \text{adj}(A)$$

Example

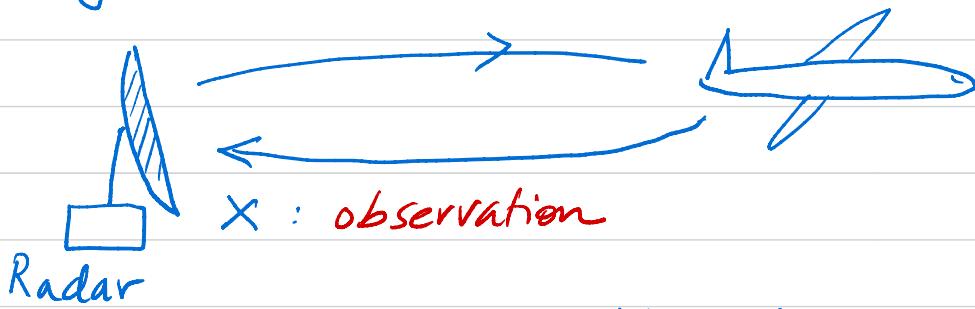
$$A = \begin{bmatrix} 7 & 2 & 1 \\ 0 & 3 & -1 \\ -3 & 4 & -2 \end{bmatrix}, \quad C = \begin{bmatrix} -2 & 3 & 9 \\ 8 & -11 & -34 \\ -5 & 7 & 21 \end{bmatrix}$$

$$\det(A) = |A| = 1$$

$$\text{adj}(A) = \begin{bmatrix} -2 & 8 & -5 \\ 3 & -11 & 7 \\ 9 & -34 & 21 \end{bmatrix} = A^{-1} \quad (\text{since } |A|=1)$$

$$\text{check: } \begin{bmatrix} 7 & 2 & 1 \\ 0 & 3 & -1 \\ -3 & 4 & -2 \end{bmatrix} \begin{bmatrix} -2 & 8 & -5 \\ 3 & -11 & 7 \\ 9 & -34 & 21 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Binary Hypothesis Testing (ECE 313)



H_0 : plane is absent

H_1 : plane is present

Goal : Decide which hypothesis is true based on X

Prior information $P(H_0) = \pi_0$, $P(H_1) = \pi_1 = 1 - \pi_0$

Observation pdf / pmf $p(x | H_0)$, $p(x | H_1)$

Likelihood Ratio $\Lambda(x) = \frac{p(x | H_1)}{p(x | H_0)}$

Performance

Probability of False Alarm $P_{FA} = P(\text{decide } H_1 | H_0)$

Probability of Miss $P_M = P(\text{decide } H_0 | H_1)$

Overall probability of error $P_e = \pi_0 P_{FA} + \pi_1 P_M$

MAP Rule decide 1 if $\Lambda(x) \geq \frac{\pi_0}{\pi_1}$
decide 0 if $\Lambda(x) < \frac{\pi_0}{\pi_1}$

$\Lambda(x) \geq \frac{\pi_0}{\pi_1}$ MAP rule minimizes P_e .

Example $\pi_0 = 0.5, \pi_1 = 0.5$.

$$p(x|H_0) = \begin{cases} 0.1 & x=0 \\ 0.2 & x=1 \\ 0.3 & x=2 \\ 0.4 & x=3 \\ 0 & \text{otherwise} \end{cases} \quad p(x|H_1) = \begin{cases} 0.25 & x=0 \\ 0.25 & x=1 \\ 0.2 & x=2 \\ 0.3 & x=3 \\ 0 & \text{otherwise} \end{cases}$$

(a) MAP Rule :

$$\Lambda(x) = \frac{p(x|H_1)}{p(x|H_0)} \stackrel{>}{\underset{0}{\gtrless}} \frac{\pi_0}{\pi_1} = 1$$

$\Lambda(x) > 1$ for $x=0, \text{ and } x=1$

$\Lambda(x) < 1$ for $x=2, \text{ and } x=3$

MAP rule decides $\begin{cases} 1 & \text{if } x=0 \text{ or } x=1 \\ 0 & \text{if } x=2 \text{ or } x=3 \end{cases}$

(b) Performance :

$$P_{FA} = P(\text{decide } H_1 | H_0) = p(0|H_0) + p(1|H_0) \\ = 0.1 + 0.2 = 0.3$$

$$P_M = P(\text{decide } H_0 | H_1) = p(2|H_1) + p(3|H_1) \\ = 0.2 + 0.3 = 0.5$$

$$P_e = \pi_0 P_{FA} + \pi_1 P_M \\ = 0.5 (0.3 + 0.5) = 0.4$$