

CV Assignment 1 - Math Part(1-3)

1.

(1) Closure

$$\forall M_i, M_j \in \{M_i\},$$

$$\text{let } M = M_i \times M_j = \begin{bmatrix} R_i & \mathbf{t}_i \\ \mathbf{0}^T & 1 \end{bmatrix} \times \begin{bmatrix} R_j & \mathbf{t}_j \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} R_i R_j & R_i \mathbf{t}_j + \mathbf{t}_i \\ \mathbf{0}^T & 1 \end{bmatrix}.$$

To prove closure is to prove that $M \in \{M_i\}$.

$$\text{Obviously, } M \in \mathbb{R}^{4 \times 4}, R_i \in \mathbb{R}^{3 \times 3}, R_j \in \mathbb{R}^{3 \times 3}, R_i R_j \in \mathbb{R}^{3 \times 3}, R_i \mathbf{t}_j + \mathbf{t}_i \in \mathbb{R}^{3 \times 1}.$$

$$\det(R_i R_j) = \det(R_i) \det(R_j) = 1 \times 1 = 1.$$

$$(R_i R_j)(R_i R_j)^T = (R_i R_j)(R_j^T R_i^T) = R_i(R_j R_j^T)R_i^T = R_i I R_i^T = R_i R_i^T = I, \text{ where } I \in \mathbb{R}^{3 \times 3} \text{ indicates identity matrix.}$$

That means matrix $R_i R_j$ is an orthonormal matrix.

In summary, matrix M satisfies that $M \in \mathbb{R}^{4 \times 4}, R_i R_j \in \mathbb{R}^{3 \times 3}, R_i \mathbf{t}_j + \mathbf{t}_i \in \mathbb{R}^{3 \times 1}, R_i R_j$ is orthonormal, and $\det(R_i R_j) = 1$. So matrix $M \in \{M_i\}$.

Thus $\{M_i\}$ satisfies the property of closure.

(2) Associativity

$$\forall M_i, M_j, M_k \in \{M_i\},$$

$$(M_i M_j) M_k = \left(\begin{bmatrix} R_i & \mathbf{t}_i \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} R_j & \mathbf{t}_j \\ \mathbf{0}^T & 1 \end{bmatrix} \right) \begin{bmatrix} R_k & \mathbf{t}_k \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} R_i R_j & R_i \mathbf{t}_j + \mathbf{t}_i \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} R_k & \mathbf{t}_k \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} R_i R_j R_k & R_i R_j \mathbf{t}_k + R_i \mathbf{t}_j + \mathbf{t}_i \\ \mathbf{0}^T & 1 \end{bmatrix}$$

$$M_i (M_j M_k) = \begin{bmatrix} R_i & \mathbf{t}_i \\ \mathbf{0}^T & 1 \end{bmatrix} \left(\begin{bmatrix} R_j & \mathbf{t}_j \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} R_k & \mathbf{t}_k \\ \mathbf{0}^T & 1 \end{bmatrix} \right) = \begin{bmatrix} R_i & \mathbf{t}_i \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} R_j R_k & R_j \mathbf{t}_k + \mathbf{t}_j \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} R_i R_j R_k & R_i R_j \mathbf{t}_k + R_i \mathbf{t}_j + \mathbf{t}_i \\ \mathbf{0}^T & 1 \end{bmatrix}$$

$$\text{So } (M_i M_j) M_k = M_i (M_j M_k).$$

Thus $\{M_i\}$ satisfies the property of associativity.

(3) Existence of an identity element

$$\text{Let identity matrix } E = I_{4 \times 4} = \begin{bmatrix} I_{3 \times 3} & \mathbf{0} \\ \mathbf{0}^T & 1 \end{bmatrix}, \text{ where } \mathbf{0} \in \mathbb{R}^{3 \times 1}.$$

Firstly, we prove that $E \in \{M_i\}$.

$$\text{Obviously, } I_{3 \times 3} I_{3 \times 3}^T = I_{3 \times 3}^T I_{3 \times 3} = I_{3 \times 3}, \det(I_{3 \times 3}) = 1.$$

So E is an element of set $\{M_i\}$.

Because E is an identity matrix, we know that

$$\forall M \in \{M_i\}, EM = ME = M$$

Thus $\{M_i\}$ satisfies the property of existence of an identity element.

(4) Existence of an inverse element for each group element

$$\text{For } \forall M \in \{M_i\}, M = \begin{bmatrix} R & \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix}, R \text{ is orthonormal, so } RR^T = I, \text{ where } I \text{ indicates identity matrix.}$$

Let $M^{-1} = \begin{bmatrix} R^T & -R^T \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix}$.

Firstly, we prove that $M^{-1} \in \{M_i\}$.

Obviously, $R^T \in \mathbb{R}^{3 \times 3}$, $-R^T \mathbf{t} \in \mathbb{R}^{3 \times 1}$,

$R^T(R^T)^T = R^T R = I$, where I indicates identity matrix,

and $\det(R^T) = \det(R) = 1$.

So M^{-1} is an element of set $\{M_i\}$.

To prove that M^{-1} is an inverse element for each group element, we compute:

$$MM^{-1} = \begin{bmatrix} R & \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} R^T & -R^T \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} RR^T & -RR^T \mathbf{t} + \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} I_{3 \times 3} & \mathbf{0} \\ \mathbf{0}^T & 1 \end{bmatrix} = I_{4 \times 4}$$

$$M^{-1}M = \begin{bmatrix} R^T & -R^T \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} \begin{bmatrix} R & \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} R^T R & R^T \mathbf{t} - R^T \mathbf{t} \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} I_{3 \times 3} & \mathbf{0} \\ \mathbf{0}^T & 1 \end{bmatrix} = I_{4 \times 4}$$

So $M^{-1} \in \{M_i\}$ is an inverse element for each group element.

2.

a)

To prove that the $n \times n$ real symmetric matrix M is semi-positive definite is to prove that its eigenvalues satisfy that $\lambda_1 \geq 0, \lambda_2 \geq 0$.

We know that:

$$\lambda_1 \lambda_2 = \det(M) = \sum_{(x_i, y_i) \in w} (I_x)^2 \sum_{(x_i, y_i) \in w} (I_y)^2 - \left(\sum_{(x_i, y_i) \in w} (I_x I_y) \right)^2$$

$$\lambda_1 + \lambda_2 = \text{trace}(M) = \sum_{(x_i, y_i) \in w} (I_x)^2 \sum_{(x_i, y_i) \in w} (I_y)^2.$$

Obviously, $\sum_{(x_i, y_i) \in w} (I_x)^2 \geq 0, \sum_{(x_i, y_i) \in w} (I_y)^2 \geq 0$, so $\lambda_1 + \lambda_2 \geq 0$.

Let's apply Cauchy-Schwarz Inequality:

$$\sum_{(x_i, y_i) \in w} (I_x)^2 \sum_{(x_i, y_i) \in w} (I_y)^2 \geq \left(\sum_{(x_i, y_i) \in w} (I_x I_y) \right)^2,$$

so $\lambda_1 \lambda_2 \geq 0$.

$$\lambda_1 \lambda_2 \geq 0 \wedge \lambda_1 + \lambda_2 \geq 0 \iff \lambda_1 \geq 0 \wedge \lambda_2 \geq 0.$$

Thus M is positive semi-definite.

b)

For simplicity, we note $\sum_{(x_i, y_i) \in w} (I_x)^2 = A, \sum_{(x_i, y_i) \in w} (I_x I_y) = B, \sum_{(x_i, y_i) \in w} (I_y)^2 = C$.

$$\text{So } M = \begin{bmatrix} A & B \\ B & C \end{bmatrix}.$$

$$\begin{bmatrix} x & y \end{bmatrix} M \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x & y \end{bmatrix} \begin{bmatrix} A & B \\ B & C \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = Ax^2 + 2Bxy + Cy^2 = 1.$$

According to the theory of conic sections, to prove that $Ax^2 + 2Bxy + Cy^2 = 1$ represents an ellipse is to prove that $A > 0, C > 0$ and the discriminant $4AC - (2B)^2 > 0$.

M is positive definite, so its eigenvalues satisfy $\lambda_1 > 0, \lambda_2 > 0$.

M satisfies that $\det(M) = AC - B^2 = \lambda_1 \lambda_2 > 0$, so the discriminant $4AC - (2B)^2 = 4(AC - B^2) > 0$.

$\text{trace}(M) = AC = \lambda_1 + \lambda_2 > 0$, and obviously $A = \sum_{(x_i, y_i) \in w} (I_x)^2 \geq 0, C = \sum_{(x_i, y_i) \in w} (I_y)^2 \geq 0$, so $A > 0, C > 0$.

In summary, $Ax^2 + 2Bxy + Cy^2 = 1$ satisfies that $A > 0, C > 0$ and the discriminant $4AC - (2B)^2 > 0$.

Thus $\begin{bmatrix} x & y \end{bmatrix} M \begin{bmatrix} x \\ y \end{bmatrix}$ represents an ellipse.

c)

We note α_1, α_2 ($\|\alpha_1\|_2^2 = 1, \|\alpha_2\|_2^2 = 1$) the eigenvectors of M , corresponding to λ_1, λ_2 respectively.

According to the condition that eigenvalues of the $n \times n$ real symmetric matrix M satisfy $\lambda_1 > \lambda_2 > 0$, its eigenvectors are linearly independent, so M can be diagonalized.

Diagonalize M :

$$M = P^T \Lambda P, P = [\alpha_1, \alpha_2], \Lambda = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

Rewrite $\begin{bmatrix} x & y \end{bmatrix} M \begin{bmatrix} x \\ y \end{bmatrix} = 1$:

$$\begin{bmatrix} x & y \end{bmatrix} M \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x & y \end{bmatrix} P^T \Lambda P \begin{bmatrix} x \\ y \end{bmatrix} = (P \begin{bmatrix} x \\ y \end{bmatrix})^T \Lambda (P \begin{bmatrix} x \\ y \end{bmatrix}) = 1$$

Note $X = \begin{bmatrix} u \\ v \end{bmatrix} = P \begin{bmatrix} x \\ y \end{bmatrix}$, we have

$$X^T \Lambda X = X^T \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} X = \lambda_1 u^2 + \lambda_2 v^2 = \frac{u^2}{\frac{1}{\lambda_1}} + \frac{v^2}{\frac{1}{\lambda_2}} = 1.$$

We know $\lambda_1 > \lambda_2 > 0$, so $\frac{1}{\lambda_2} > \frac{1}{\lambda_1} > 0$

Thus for the ellipse, the length of its semi-major axis is $\frac{1}{\sqrt{\lambda_2}}$, the length of its semi-minor axis is $\frac{1}{\sqrt{\lambda_1}}$.

3.

Firstly, let's prove that $\text{rank}(A^T A) = \text{rank}(A) = n$, for which we prove that the homogeneous linear equation system $A\mathbf{x} = 0$ and $A^T A\mathbf{x} = 0$ have the same solution

Obviously, the solutions of $A\mathbf{x} = 0$ are also solutions of $A^T A\mathbf{x} = 0$

Let vector α be the solution of $A^T A\mathbf{x} = 0$, then $\alpha^T A^T A\alpha = (A\alpha)^T (A\alpha) = 0$. The inner product of vector $A\alpha$ is 0, then $A\alpha = \mathbf{0}$. So the solutions of $A^T A\mathbf{x} = 0$ are also solutions of $A\mathbf{x} = 0$.

Now we know that $A\mathbf{x} = 0$ and $A^T A\mathbf{x} = 0$ have the same solution, so $\text{rank}(A^T A) = \text{rank}(A) = n$.

Obviously, $A^T A \in \mathbb{R}^{n \times n}$. Since $\text{rank}(A^T A) = n$, $A^T A$ is invertible.