

# ESE 680 Autonomous Racing

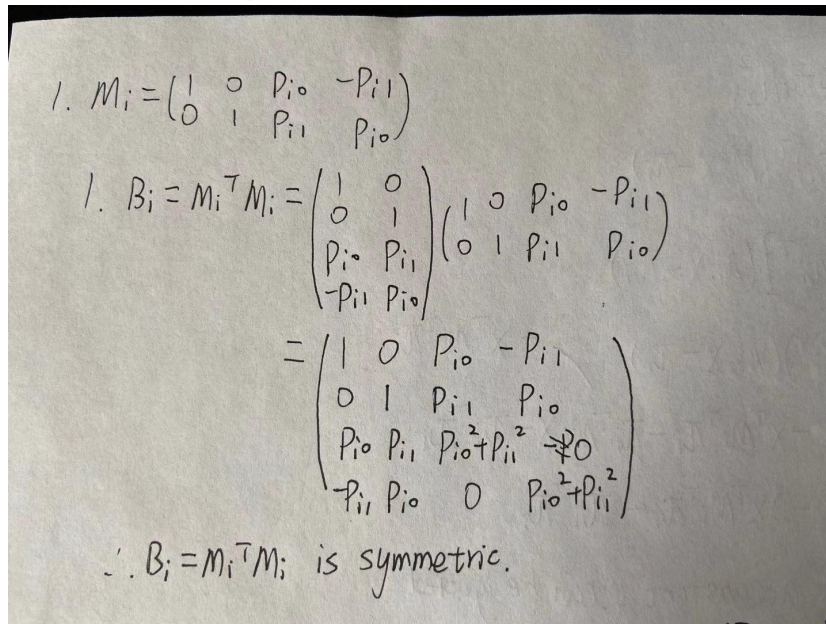
## Lab 5: Scan\_matching - PLICP method - Localization

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### 1 Theoretical Questions (20)

1.  $M_i = \begin{pmatrix} 1 & 0 & p_{i0} & -p_{i1} \\ 0 & 1 & p_{i1} & p_{i0} \end{pmatrix}$

1. Show that  $B_i := M_i^T M_i$  is symmetric.



Handwritten derivation:

$$1. M_i = \begin{pmatrix} 1 & 0 & p_{i0} & -p_{i1} \\ 0 & 1 & p_{i1} & p_{i0} \end{pmatrix}$$

$$1. B_i = M_i^T M_i = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ p_{i0} & p_{i1} \\ -p_{i1} & p_{i0} \end{pmatrix} \begin{pmatrix} 1 & 0 & p_{i0} & -p_{i1} \\ 0 & 1 & p_{i1} & p_{i0} \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 & p_{i0} & -p_{i1} \\ 0 & 1 & p_{i1} & p_{i0} \\ p_{i0} & p_{i1} & p_{i0}^2 + p_{i1}^2 & 0 \\ -p_{i1} & p_{i0} & 0 & p_{i0}^2 + p_{i1}^2 \end{pmatrix}$$

$\therefore B_i = M_i^T M_i$  is symmetric.

2. Demonstrate that  $B_i$  is positive semi-definite.

As from [http://theanalysisofdata.com/probability/C\\_4.html](http://theanalysisofdata.com/probability/C_4.html), Proposition C.4.1. We know: A symmetric matrix is psd (positive semi definite) if and only if all eigenvalues are non-negative. Eigenvalues for  $B_i$  are calculated using Matlab:

```
>> positive_semi_definite_matrix
e =
     0
     0
 pi0^2 + pi1^2 + 1
 pi0^2 + pi1^2 + 1
```

All eigenvalues are non-negative. Thus,  $B_i$  is a positive semi-definite matrix.

$$\begin{aligned} x^* = \operatorname{argmin}_{x \in \mathbb{R}^4} \quad & \sum_{i=1}^n \|M_i x - \pi_i\|_2^2 \\ \text{s.t.} \quad & x_3^2 + x_4^2 = 1 \end{aligned}$$
$$\begin{aligned} x^* = \operatorname{argmin}_{x \in \mathbb{R}^4} & x^T M x + g^T x \\ \text{s.t. } & x^T W x = 1 \end{aligned}$$

$$\begin{aligned}
 2. \sum_{i=1}^n \|M_i X - \pi_i\|_2^2 \\
 &= \sum_{i=1}^n (M_i X - \pi_i)^T (M_i X - \pi_i) \\
 &= \sum_{i=1}^n [(M_i X)^T - \pi_i^T] (M_i X - \pi_i) \\
 &= \sum_{i=1}^n (X^T M_i^T - \pi_i^T) (M_i X - \pi_i) = X^T M_i^T \pi_i \\
 &= \sum_{i=1}^n X^T M_i^T M_i X - X^T M_i^T \pi_i - \pi_i^T M_i X - \pi_i^T \pi_i \\
 &= \sum_{i=1}^n X^T \underbrace{M_i^T M_i}_M X - 2X^T M_i^T \pi_i - \pi_i^T \pi_i \\
 &\because -\pi_i^T \pi_i \text{ is a constant, } \therefore \text{can be ignored} \\
 &= \sum_{i=1}^n X^T M X - \underbrace{2X^T M_i^T \pi_i}_{g^T} \\
 &\therefore M = M_i^T M_i, \quad g^T = M_i^T \pi_i \quad M = \sum_i M_i^T M_i, \quad g = \sum_i -2\pi_i^T M_i \\
 &\text{when } W = \begin{bmatrix} O_{2 \times 2} & O_{2 \times 2} \\ O_{2 \times 2} & I_{2 \times 2} \end{bmatrix}, \quad X_3^2 + X_4^2 = 1 \\
 &\quad \quad \quad \downarrow \\
 &\quad \quad \quad X^T W X = 1
 \end{aligned}$$

Assuming a non-zero column vector  $z$

$$\mathbf{z}^T \mathbf{W} \mathbf{z} = \begin{bmatrix} z_1 & z_2 & z_3 & z_4 \end{bmatrix} \begin{bmatrix} 0_{2 \times 2} & 0_{2 \times 2} \\ 0_{2 \times 2} & \mathbf{I}_{2 \times 2} \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \\ z_4 \end{bmatrix}$$

$$= z_3^2 + z_4^2 \geq 0$$

$\therefore W$  is positive semi-definite matrix.

As from the calculation shown in the screenshot, we can tell that:

$$M = \sum_{i=1}^n M_i^T M_i, g = \sum_{i=1}^n -2\Pi^T M_i, W = \begin{bmatrix} 0_{2 \times 2} & 0_{2 \times 2} \\ 0_{2 \times 2} & I_{2 \times 2} \end{bmatrix}$$

Which are the answers for question a.

B. Show that M and W are positive semi definite matrix.

The prove of W is shown in the above screenshot using the psd definition.

The prove of M is shown in the below screenshot using the prove in question 1.

From Q1, we know  $M_i^T M_i$  is psd.

$$\therefore M = \sum_{i=1}^n M_i^T M_i, z^T M z = z^T \sum_{i=1}^n M_i^T M_i z$$

$$= z^T (M_1^T M_1 + M_2^T M_2 + \dots + M_n^T M_n) z$$

$$= z^T M_1^T M_1 z + z^T M_2^T M_2 z + \dots + z^T M_n^T M_n z$$

$\therefore$  every items on the right side are non-negative  
as from the results in Q1.

$\therefore$  the sum of them are non-negative.

$\therefore M \succeq 0$

$\therefore M$  is psd.