

# Line Assignment

Gautam Singh

**Abstract**—This document contains a general solution to Question 16 of Exercise 2 in Chapter 11 of the class 12 NCERT textbook.

- 1) Find the shortest distance between the lines whose vector equations are

$$L_1 : \mathbf{x} = \mathbf{x}_1 + \lambda_1 \mathbf{m}_1 \quad (1)$$

$$L_2 : \mathbf{x} = \mathbf{x}_2 + \lambda_2 \mathbf{m}_2 \quad (2)$$

**Solution:** Let  $\mathbf{A}$  and  $\mathbf{B}$  be points on lines  $L_1$  and  $L_2$  respectively such that  $AB$  is normal to both lines. Define

$$\mathbf{M} \triangleq (\mathbf{m}_1 \quad \mathbf{m}_2) \quad (3)$$

$$\lambda \triangleq \begin{pmatrix} \lambda_1 \\ -\lambda_2 \end{pmatrix} \quad (4)$$

$$\mathbf{x} \triangleq \mathbf{x}_2 - \mathbf{x}_1 \quad (5)$$

Then, we have the following equations:

$$\mathbf{A} = \mathbf{x}_1 + \lambda_1 \mathbf{m}_1 \quad (6)$$

$$\mathbf{B} = \mathbf{x}_2 + \lambda_2 \mathbf{m}_2 \quad (7)$$

From (6) and (7), define the real-valued function  $f$  as

$$f(\lambda) \triangleq \|\mathbf{A} - \mathbf{B}\| \quad (8)$$

$$= \|\mathbf{M}\lambda - \mathbf{x}\| \quad (9)$$

$$= \sqrt{(\mathbf{M}\lambda - \mathbf{x})^\top (\mathbf{M}\lambda - \mathbf{x})} \quad (10)$$

Note that the norm function obeys the triangle inequality, which will be used later. To prove this, note that for vectors  $\mathbf{a}$  and  $\mathbf{b}$ ,

$$\left\| \mathbf{a} - \frac{\mathbf{a}^\top \mathbf{b}}{\|\mathbf{b}\|^2} \mathbf{b} \right\|^2 \geq 0 \quad (11)$$

$$\Rightarrow \|\mathbf{a}\|^2 - 2 \frac{(\mathbf{a}^\top \mathbf{b})^2}{\|\mathbf{b}\|^2} + \frac{(\mathbf{a}^\top \mathbf{b})^2}{\|\mathbf{b}\|^2} \geq 0 \quad (12)$$

$$\Rightarrow \|\mathbf{a}\|^2 - \frac{(\mathbf{a}^\top \mathbf{b})^2}{\|\mathbf{b}\|^2} \geq 0 \quad (13)$$

$$\Rightarrow \|\mathbf{a}\|^2 \|\mathbf{b}\|^2 \geq (\mathbf{a}^\top \mathbf{b})^2 \quad (14)$$

$$\Rightarrow \|\mathbf{a}\| \|\mathbf{b}\| \geq \mathbf{a}^\top \mathbf{b} \quad (15)$$

Using (15) as follows

$$\mathbf{a}^\top \mathbf{b} \leq \|\mathbf{a}\| \|\mathbf{b}\| \quad (16)$$

$$\|\mathbf{a}\|^2 + 2\mathbf{a}^\top \mathbf{b} + \|\mathbf{b}\|^2 \leq \|\mathbf{a}\|^2 + 2\|\mathbf{a}\| \|\mathbf{b}\| + \|\mathbf{b}\|^2 \quad (17)$$

$$\|\mathbf{a} + \mathbf{b}\|^2 \leq (\|\mathbf{a}\| + \|\mathbf{b}\|)^2 \quad (18)$$

$$\|\mathbf{a} + \mathbf{b}\| \leq \|\mathbf{a}\| + \|\mathbf{b}\| \quad (19)$$

This proves the triangle inequality.

We now show that  $f$  is convex. Indeed, consider  $\lambda_1$  and  $\lambda_2$  and let  $0 \leq \mu \leq 1$ . Then,

$$f(\mu\lambda_1 + (1-\mu)\lambda_2) \quad (20)$$

$$= \|\mathbf{M}(\mu\lambda_1 + (1-\mu)\lambda_2) - \mathbf{x}\| \quad (21)$$

$$= \|\mu(\mathbf{M}\lambda_1 - \mathbf{x}) + (1-\mu)(\mathbf{M}\lambda_2 - \mathbf{x})\| \quad (22)$$

$$\leq \mu \|\mathbf{M}\lambda_1 - \mathbf{x}\| + (1-\mu) \|\mathbf{M}\lambda_2 - \mathbf{x}\| \quad (23)$$

Where (23) follows from (19).

We need to minimize  $f$  as a function of  $\lambda$ . Thus, differentiating (10) using the chain rule,

$$\frac{df(\lambda)}{d\lambda} = \frac{\mathbf{M}^\top (\mathbf{M}\lambda - \mathbf{x}) + \mathbf{M} (\mathbf{M}\lambda - \mathbf{x})^\top}{2 \|\mathbf{M}\lambda - \mathbf{x}\|} \quad (24)$$

$$= \frac{\mathbf{M}^\top (\mathbf{M}\lambda - \mathbf{x})}{\|\mathbf{M}\lambda - \mathbf{x}\|} \quad (25)$$

Setting (25) to zero gives

$$\mathbf{M}^\top \mathbf{M} \lambda = \mathbf{M}^\top \mathbf{x} \quad (26)$$

We have the following cases:

- a) There exists a  $\lambda$  satisfying

$$\mathbf{M}\lambda = \mathbf{x} \quad (27)$$

$$\Rightarrow \lambda_1 \mathbf{m}_1 - \lambda_2 \mathbf{m}_2 = \mathbf{x}_2 - \mathbf{x}_1 \quad (28)$$

$$\Rightarrow \mathbf{x}_1 + \lambda_1 \mathbf{m}_1 = \mathbf{x}_2 + \lambda_2 \mathbf{m}_2 \quad (29)$$

Thus, both lines intersect at a point and the shortest distance between them is 0. To check for the existence of such a  $\lambda$ , we can bring the augmented matrix  $(\mathbf{M} \quad \mathbf{x})$  into row-reduced echelon form and check whether there is a pivot in the last column.

- b)  $\mathbf{M}^\top \mathbf{M}$  is singular. Since  $\mathbf{M}^\top \mathbf{M}$  is a square

matrix of order 2, its rank must be 1. Further,

$$\det(\mathbf{M}^\top \mathbf{M}) = \begin{vmatrix} \mathbf{m}_1^\top \mathbf{m}_1 & \mathbf{m}_1^\top \mathbf{m}_2 \\ \mathbf{m}_1^\top \mathbf{m}_2 & \mathbf{m}_2^\top \mathbf{m}_2 \end{vmatrix} \quad (30)$$

$$= (\|\mathbf{m}_1\| \cdot \|\mathbf{m}_2\|)^2 - (\mathbf{m}_1^\top \mathbf{m}_2)^2 \quad (31)$$

Thus, equating the determinant to zero gives

$$\|\mathbf{m}_1\| \cdot \|\mathbf{m}_2\| = |\mathbf{m}_1^\top \mathbf{m}_2| \quad (32)$$

which implies that both lines are parallel to each other. Setting  $\mathbf{m}_2 = k\mathbf{m}_1, k \in \mathbb{R} \setminus \{0\}$ , we obtain one equation from (26).

$$\mathbf{m}_1^\top \mathbf{m}_1 (\lambda_1 - k\lambda_2) = \mathbf{m}_1^\top \mathbf{x} \quad (33)$$

$$\implies \lambda_1 - k\lambda_2 = \frac{\mathbf{m}_1^\top \mathbf{x}}{\|\mathbf{m}_1\|^2} \quad (34)$$

Therefore, the required shortest distance is

$$\|\mathbf{A} - \mathbf{B}\| = \left\| \frac{\mathbf{m}_1^\top \mathbf{x} \mathbf{m}_1}{\|\mathbf{m}_1\|^2} - \mathbf{x} \right\| \quad (35)$$

c)  $\mathbf{M}^\top \mathbf{M}$  is nonsingular. This implies that the lines are skew. From (26),

$$\lambda = (\mathbf{M}^\top \mathbf{M})^{-1} \mathbf{M}^\top \mathbf{x} \quad (36)$$

and therefore, the shortest distance is

$$\|\mathbf{A} - \mathbf{B}\| = \left\| \left( \mathbf{M} (\mathbf{M}^\top \mathbf{M})^{-1} \mathbf{M}^\top - \mathbf{I}_n \right) \mathbf{x} \right\| \quad (37)$$

where  $\mathbf{I}_n$  is the identity matrix of order  $n$ .