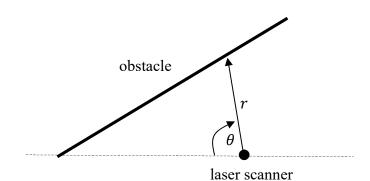
Time: 9:00-11:00 h. Marks: 100

#### **Open Book**

#### Attempt all questions.

Q.1 A laser scanner is used to create map of an unknown environment. The outputs from the laser scanner are range in cm, r, and direction in degree,  $\theta$ , of the obstacle respect to the laser scanner.  $\theta = 90^{\circ}$  is the direction of right in front of the laser scanner and clockwise direction is positive direction.



Assume a linear wall locates in the unknown environment and data from the laser scanner is collected and shown in the below table.

θ	0	30	60	90
r	50	56.12	92.03	1000

ADALINE network is used to determine the cartesian parameters, (m, c), of the linear wall according to

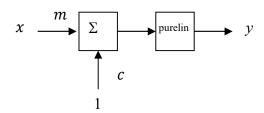
y = mx + c. Cartesian coordinate of the laser scanner location is set at (x, y) = (0, 0).

- (a) Convert the data from  $(r, \theta)$  to Cartesian coordinate (x, y). (10)
- (b) Draw the ADALINE network that can be used to determine the parameters. (5)
- (c) Determine all the parameters of the linear wall by LMS algorithm. (10)

#### <u>Solution</u>

(a)

x	-50	-48.60	-46.02	0
у	0	28.06	79.70	1000



(b)

(b)

$$F(x) = E[t^{2}] - 2x^{T}E[tz] + x^{T}E[zz^{T}]x = c - 2x^{T}h + x^{T}Rx$$
(1)

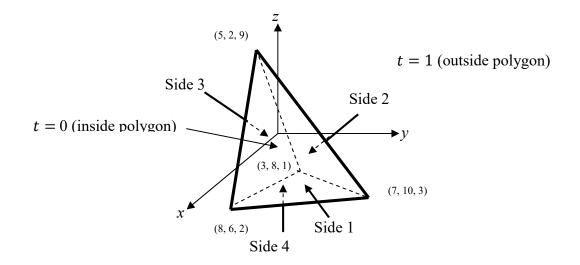
$$c = \frac{1}{4}(0^2 + 28.06^2 + 79.70^2 + 1000^2) = 251784.9$$
(2)

$$h = \frac{1}{4} \left( 0 \begin{bmatrix} -50\\1 \end{bmatrix} + 28.06 \begin{bmatrix} -48.60\\1 \end{bmatrix} + 79.70 \begin{bmatrix} -46.02\\1 \end{bmatrix} + 1000 \begin{bmatrix} 0\\1 \end{bmatrix} \right) = \begin{bmatrix} -1257.9\\276.9 \end{bmatrix}$$
(3)

$$R = \frac{1}{4} \left( \begin{bmatrix} -50\\1 \end{bmatrix} \begin{bmatrix} -50\\1 \end{bmatrix}^T + \begin{bmatrix} -48.60\\1 \end{bmatrix} \begin{bmatrix} -48.60\\1 \end{bmatrix}^T + \begin{bmatrix} -46.02\\1 \end{bmatrix} \begin{bmatrix} -46.02\\1 \end{bmatrix}^T + \begin{bmatrix} 0\\1 \end{bmatrix} \begin{bmatrix} 0\\1 \end{bmatrix}^T \right) = \begin{bmatrix} 1745.0 & -36.2\\-36.2 & 1 \end{bmatrix}$$
(4)

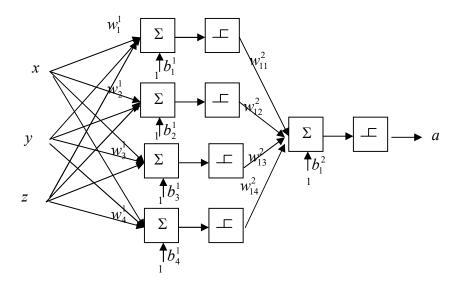
The minimum point is the stationary point of the quadratic function.

**Q.2** Design a two-layer perceptron network with appropriate parameters, which generates output of 0 when the input vector is inside the polygon and generates output of 1 when the input vector is outside the polygon. Use integer for all coordinates of weight vectors. (25)



### **Solution**

Two-Layer Perceptron is selected. The first layer is used to create decision boundaries 1, 2, 3, and 4 when the weight vectors point outward from the polygon. The second layer is used to OR decision boundaries 1, 2, 3, and 4.



### For layer 1

**w** must point outward from the polygon and it must be perpendicular to decision boundary and edges.

**b** is determined from equating  $\mathbf{n} = 0$ ; and solve for **b** 

For the first decision boundary,

$$\begin{bmatrix} w_{11}^{1} & w_{12}^{1} & w_{13}^{1} \end{bmatrix} \begin{bmatrix} -1 \\ 4 \\ 1 \end{bmatrix} = -w_{11}^{1} + 4w_{12}^{1} + w_{13}^{1} = 0$$
(1)

$$\begin{bmatrix} w_{11}^{l} & w_{12}^{l} & w_{13}^{l} \end{bmatrix} \begin{bmatrix} -2 \\ -8 \\ 6 \end{bmatrix} = -2w_{11}^{l} - 8w_{12}^{l} + 6w_{13}^{l} = 0$$
(2)

$$\mathbf{w}_1^1 = \begin{bmatrix} 8\\1\\4 \end{bmatrix} \tag{3}$$

$$n_1^1 = w_{11}^1 x + w_{12}^1 y + w_{13}^1 z + b_1^1 = 0$$
(4)

at x = 8, y = 6, and z = 2;

$$8(8) + 1(6) + 4(2) + b_1^1 = 0; b_1^1 = -78$$
(5)

For the second decision boundary,

$$\begin{bmatrix} w_{21}^{1} & w_{22}^{1} & w_{23}^{1} \end{bmatrix} \begin{bmatrix} -2 \\ -8 \\ 6 \end{bmatrix} = -2w_{21}^{1} - 8w_{22}^{1} + 6w_{23}^{1} = 0$$
(6)

$$\begin{bmatrix} w_{21}^{1} & w_{22}^{1} & w_{23}^{1} \end{bmatrix} \begin{bmatrix} 2 \\ -6 \\ 8 \end{bmatrix} = 2w_{21}^{1} - 6w_{22}^{1} + 8w_{23}^{1} = 0$$
(7)

$$\mathbf{w}_2^1 = \begin{bmatrix} -1\\1\\1 \end{bmatrix}$$
 (8)

$$n_2^1 = w_{21}^1 x + w_{22}^1 y + w_{23}^1 z + b_2^1 = 0$$
(9)

at x = 7, y = 10, and z = 3

$$7(-1) + 10(1) + 3(1) + b_2^1 = 0; b_2^1 = -6$$
(10)

For the third decision boundary,

$$\begin{bmatrix} w_{31}^{1} & w_{32}^{1} & w_{33}^{1} \end{bmatrix} \begin{bmatrix} 2 \\ -6 \\ 8 \end{bmatrix} = 2w_{31}^{1} - 6w_{32}^{1} + 8w_{33}^{1} = 0$$
(11)

$$\begin{bmatrix} w_{31}^{1} & w_{32}^{1} & w_{33}^{1} \end{bmatrix} \begin{bmatrix} 5 \\ -2 \\ 1 \end{bmatrix} = 5w_{31}^{1} - 2w_{32}^{1} + 1w_{33}^{1} = 0$$
(12)

$$\mathbf{w}_{3}^{1} = \begin{bmatrix} -5\\ -19\\ -13 \end{bmatrix} \tag{13}$$

at x = 3, y = 8, and z = 1

$$3(-5) + 8(-19) + 1(-13) + b_3^1 = 0; b_3^1 = 180$$
(14)

For the fourth decision boundary,

$$\begin{bmatrix} w_{41}^{1} & w_{42}^{1} & w_{43}^{1} \end{bmatrix} \begin{bmatrix} -1 \\ 4 \\ 1 \end{bmatrix} = -w_{41}^{1} + 4w_{42}^{1} + w_{43}^{1} = 0$$
(15)

$$\begin{bmatrix} w_{41}^{1} & w_{42}^{1} & w_{43}^{1} \end{bmatrix} \begin{bmatrix} 4\\2\\2 \end{bmatrix} = 4w_{41}^{1} + 2w_{42}^{1} + 2w_{43}^{1} = 0$$
(16)

$$\mathbf{w}_{4}^{1} = \begin{bmatrix} 1\\1\\-3 \end{bmatrix} \tag{17}$$

at x = 8, y = 6, and z = 2

$$1(8) + 1(6) - 3(2) + b_4^1 = 0; b_4^1 = -8$$
(18)

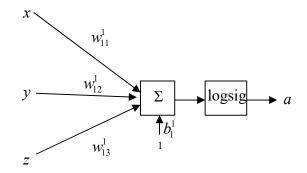
### For layer 2

Since this is OR layer, if we select

$$w_{11}^2 = w_{12}^2 = w_{13}^2 = w_{14}^2 = 1$$
(19)

$$b_1^2 = -0.5$$
 (20)

**Q.3** Consider a 3-1 neural network as shown below.



Train this network by SDBP using both training sets for one round with learning rate,  $\alpha = 0.5$ . The training sets are given as

$$\left\{ p_1 = \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix}, t_1 = [0] \right\}, \left\{ p_2 = \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}, t_1 = [1] \right\}$$

The initial weights and bias are as follows. $w_{11}^1 = 0.1, w_{12}^1 = -0.2, w_{13}^1 = 0.3, b_1^1 = -0.4.$  (25)

# <u>Solution</u>

Determine derivative of the transfer function,

$$f^{1} = (1 + e^{-n})^{-1}, \dot{f}^{1} = (1 - a^{1})(a^{1})$$
(1)

Present  $p_1$ ,

$$n_{1}^{1} = \begin{bmatrix} 0.1 & -0.2 & 0.3 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix} + \begin{bmatrix} -0.4 \end{bmatrix} = -0.4, a_{1}^{1} = (1 + e^{-n_{1}^{1}})^{-1} = (1 + e^{0.4})^{-1} = 0.4013$$
(2)

$$e = t - a = 0 - (0.4013) = -0.4013 \tag{3}$$

$$s^{1} = -2\dot{f}^{1}e = -2(1 - 0.4013)(0.4013)(-0.4013) = 0.1928$$
(4)

$$w^{1}(1) = w^{1}(0) - \alpha s^{1}(a^{0})^{T} = \begin{bmatrix} 0.1 & -0.2 & 0.3 \end{bmatrix} - 0.5 \begin{bmatrix} 0.1928 \end{bmatrix} \begin{bmatrix} 1 & -1 & -1 \end{bmatrix} = \begin{bmatrix} 0.0036 & -0.1036 & 0.3964 \end{bmatrix}$$
(5)

$$b^{1}(1) = b^{1}(0) - \alpha s^{1} = [-0.4] - 0.5[0.1928] = [-0.4964]$$
(6)

Present  $p_2$ ,

$$n_{1}^{1} = \begin{bmatrix} 0.0036 & -0.1036 & 0.3964 \end{bmatrix} \begin{bmatrix} 1\\ 1\\ -1 \end{bmatrix} + \begin{bmatrix} -0.4964 \end{bmatrix} = -0.9928, a_{1}^{1} = (1 + e^{-n_{1}^{1}})^{-1} = (1 + e^{0.9928})^{-1} = 0.2704$$
(7)

$$e = t - a = 1 - (0.2704) = 0.7296 \tag{8}$$

$$s^{1} = -2\dot{f}^{1}e = -2(1 - 0.2704)(0.2704)(0.7296) = -0.2879$$
(9)

$$w^{1}(2) = w^{1}(1) - \alpha s^{1}(a^{0})^{T} = [0.0036 - 0.1036 - 0.3964] - 0.5[-0.2879][1 - 1 - 1] = [0.1475 - 0.0403 - 0.2525](10)$$

$$b^{1}(2) = b^{1}(1) - \alpha s^{1} = [-0.4964] - 0.5[-0.2879] = [-0.3525]$$
(11)

**Q.4** LVQ network is used to recognize 3 patterns according to the following training sets.

$$\left\{ p_1 = \begin{bmatrix} 10\\10\\10 \end{bmatrix}, t_1 = \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}, \left\{ p_2 = \begin{bmatrix} 10.5\\9.8\\9.6 \end{bmatrix}, t_2 = \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}, \left\{ p_3 = \begin{bmatrix} 9.7\\10.3\\10.2 \end{bmatrix}, t_3 = \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}, \left\{ p_4 = \begin{bmatrix} 5\\-15\\0 \end{bmatrix}, t_4 = \begin{bmatrix} 0\\1\\0 \end{bmatrix} \right\}, \left\{ p_5 = \begin{bmatrix} 4.7\\-14.8\\0.3 \end{bmatrix}, t_5 = \begin{bmatrix} 0\\1\\0 \end{bmatrix} \right\}, \left\{ p_6 = \begin{bmatrix} -20\\-20\\-20 \end{bmatrix}, t_6 = \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}, \left\{ p_{11} = \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}, \left\{ p_{12} = \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}, \left\{ p_{13} = \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}, \left$$

$$\left\{ p_7 = \begin{bmatrix} -19.6 \\ -20.2 \\ -20.3 \end{bmatrix}, t_7 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\}, \left\{ p_8 = \begin{bmatrix} -20.3 \\ -19.5 \\ -19.1 \end{bmatrix}, t_8 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \right\}, \left\{ p_9 = \begin{bmatrix} -5 \\ 20 \\ -15 \end{bmatrix}, t_9 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}, \left\{ p_{10} = \begin{bmatrix} -4.8 \\ 19.7 \\ -15.4 \end{bmatrix}, t_{10} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}, \left\{ p_{11} = \begin{bmatrix} 15 \\ 15 \\ 15 \\ 15 \end{bmatrix}, t_{11} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}, \left\{ p_{12} = \begin{bmatrix} 15.4 \\ 14.7 \\ 14.9 \end{bmatrix}, t_{12} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

If a very small learning rate,  $\alpha$ , is applied, determine the weights of the first and the second layer of the LVQ network. (25)

## **Solution**

From the training sets, there are 5 sub classes.

In the first layer of LVQ, the trained weights vector will be at the center of each sub class.

$$w_{1} = \frac{1}{3} \begin{bmatrix} 10+10.5+9.7\\10+9.8+10.3\\10+9.6+10.2 \end{bmatrix} = \begin{bmatrix} 10.07\\10.03\\9.93 \end{bmatrix}$$
(1)

$$w_{2} = \frac{1}{2} \begin{bmatrix} 5+4.7\\-15-14.8\\0+0.3 \end{bmatrix} = \begin{bmatrix} 4.85\\-14.90\\0.15 \end{bmatrix}$$
(2)

$$w_{3} = \frac{1}{3} \begin{bmatrix} -20 - 19.6 - 20.3 \\ -20 - 20.2 - 19.5 \\ -20 - 20.3 - 19.1 \end{bmatrix} = \begin{bmatrix} -19.97 \\ -19.90 \\ -19.80 \end{bmatrix}$$
(3)

$$w_{4} = \frac{1}{2} \begin{bmatrix} -5 - 4.8\\ 20 + 19.7\\ -15 - 15.4 \end{bmatrix} = \begin{bmatrix} -4.90\\ 19.85\\ -15.20 \end{bmatrix}$$
(4)

$$w_{5} = \frac{1}{2} \begin{bmatrix} 15+15.4\\15+14.7\\15+14.9 \end{bmatrix} = \begin{bmatrix} 15.20\\14.85\\14.95 \end{bmatrix}$$
(5)

The weight matrix of the first layer is thus

$$W^{1} = \begin{bmatrix} 10.07 & 4.85 & -19.97 & -4.90 & 15.20 \\ 10.03 & -14.90 & -19.90 & 19.85 & 14.85 \\ 9.93 & 0.15 & -19.80 & -15.20 & 14.95 \end{bmatrix}$$
(6)

The weight matrix of the second layer combines and recognizes subclasses 1 and 3 as class 1, subclass 2 as class 2 and sub classes 4 and 5 as class 3 respectively.

$$W^{2} = \begin{bmatrix} 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \end{bmatrix}$$
(7)