

Department of Electrical and Computer Engineering


Course Number	EE8603
Course Title	Neural Networks and Deep Learning
Semester/Year	Summer/2018

Instructor	Dr. Kandasamy Illanko
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Assignment No.	2
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Assignment Title	Analytical and MATLAB Perceptron Training
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Submission Date	29 th May 2018
Due Date	29 th May 2018

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1 EXERSICE 3.5.2

1.1 MATLAB CODE

```
1  %%% PART A %%%
2  CP = [5 10; 10 5]; % Cluster center points
3  N = 25; % Number of points per cluster
4  C = size(CP,1); % Number of clusers
5  a = 2; % Standard deviation of the clusters
6  mean = 0; % Mean of the clusters
7  m = mean*ones(C*N, 1); % Array of means (zeros)
8  s = a*ones(C*N, 1); % Array of standard deviation (all a)
9  r = normrnd(m, s); % Random number array (Normally distributed)
10 theta = pi*rand(C*N, 1); % Random angles
11 x = zeros(C*N,2); % 2D array containing x_1 and x_2
12 col = ['b' 'r' 'g' 'y' 'm' 'c' 'k'];
13 figure
14 for c_i = 1:C
15     %disp(['c_i = ', num2str(c_i)]);
16     for i = 1+(c_i-1)*N:N+(c_i-1)*N
17         %disp(['i = ', num2str(i)]);
18         x(i,1) = CP(c_i, 1) + r(i)*sin(theta(i)); % x_1
19         x(i,2) = CP(c_i, 2) + r(i)*cos(theta(i)); % x_2
20     end
21     scatter(x(1+(c_i-1)*N:N+(c_i-1)*N,1), x(1+(c_i-1)*N:N+(c_i-1)*N,2), col(c_i), 'filled'); hold
22         on
23 end
24 hold off
25 %%% PART B %%%
26 points = 1:1:50; % Array of 50 integers
27 n_choose = 20; % Choose amount
28 sample = randsample(points, n_choose); % Choose 20 integers at random from 50 integers
29 chosen_points = zeros(n_choose, 2); % 2D array containing x_1 and x_2 of chosen points
30 for i = 1:n_choose
31     % Copy over into chosen array
32     chosen_points(i,1) = x(sample(i),1); % x_1
33     chosen_points(i,2) = x(sample(i),2); % x_2
34 end
35 %%% PART C %%%
36 n_other = (C*N)-n_choose;
37 other_points = zeros(n_other, 2); % 2D array containing x_1 and x_2 of other points
38 cursor = 1;
39 for i = 1:C*N
40     % If not chosen, copy over into other array
41     if(~ismember(sample, i))
42         other_points(cursor,1) = x(i,1); % x_1
43         other_points(cursor,2) = x(i,2); % x_2
44         cursor = cursor+1;
45     end
46 end
47 figure
48 scatter(chosen_points(:,1), chosen_points(:,2), 'r', 'o'); hold on
49 scatter(other_points(:,1), other_points(:,2), 'b', 's'); hold off
50
51 %%% PART D %%%
52 d_p = sqrt((CP(1, 1)-chosen_points(:,1)).^2 + (CP(1, 1)-chosen_points(:,1)).^2); % Distance from
53 d_q = sqrt((CP(2, 1)-chosen_points(:,1)).^2 + (CP(2, 1)-chosen_points(:,1)).^2); % Distance from Q
54 count_closer_to_p = 0;
55 for i = 1:n_choose
56     if(d_p(i) < d_q(i))
57         count_closer_to_p = count_closer_to_p + 1;
58     end
59 end
```

```
61 disp(['Out of ', num2str(n_choose), ' chosen points ', num2str(count_closer_to_p), ' points are closer to P than Q.']);
```

1.2 FIGURES

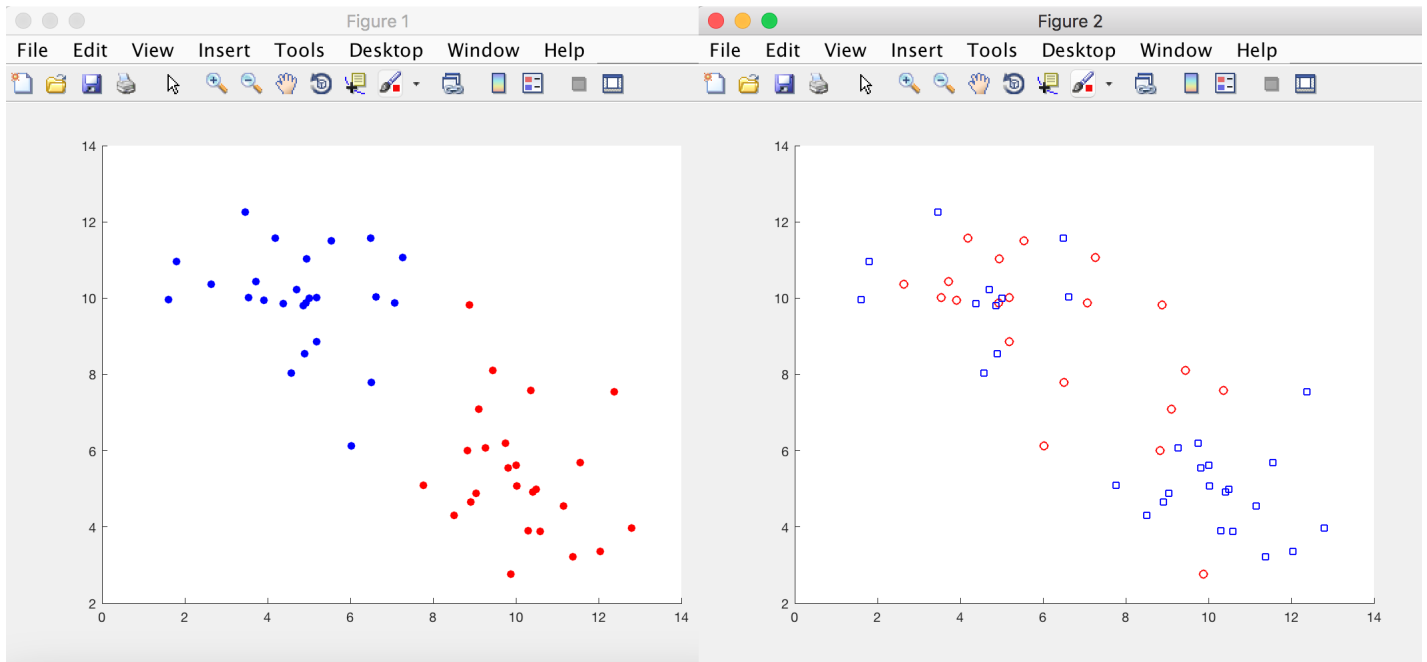


Figure 1.1: **Left:** cluster points marked with different colors.

Right: Chosen points are marked with red circles and other points are marked with blue squares.

1.3 OUTPUT

```
>> exercise_3_5_2
Out of 20 chosen points 11 points are closer to P than Q.
```

1.4 NUMERICAL ANSWER

Distance between two points in a R^2 space is given by:-

$$d = \sqrt{(y_2 - y_1)^2 + (x_2 - x_1)^2}$$

For points nearer to P than Q:

$$\text{nearPthanQ} = \begin{cases} 0 & \sqrt{(10 - y)^2 + (5 - x)^2} - \sqrt{(5 - y)^2 + (10 - x)^2} < 0 \\ 1 & \sqrt{(10 - y)^2 + (5 - x)^2} - \sqrt{(5 - y)^2 + (10 - x)^2} > 0 \end{cases}$$

Where y and x are coordinates of a point to test

2 EXERSICE 4.1.2

2.1 MATLAB CODE

```
1 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
2 % Data Creation
3 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
4 CB = [14 20; 0 6; 0 6; 0 6; 14 20; 14 20]; % Cluster boundaries
5 D = 3; % Space dimensions
6 N = 2000; % Number of points per cluster
7 C = size(CB,1)/D; % Number of clusers
8 t = zeros(C*N,1); % Preallocate for the input and the target vector
9 x = zeros(C*N,D); % Dimension array containing x_1, x_2, ... x_D
10 col = ['b' 'r' 'g' 'y' 'm' 'c' 'k'];
11 figure
12 for c_i = 1:C
13     %disp(['c_i = ', num2str(c_i)]);
14     for p_i = 1+(c_i-1)*N:N+(c_i-1)*N
15         %disp(['i = ', num2str(i)]);
16         for axis = 1:D
17             pos = ((c_i-1)*D)+axis;
18             range = CB(pos,2) - CB(pos,1);
19             x(p_i, axis) = range*rand + CB(pos, 1);
20         end
21         t(p_i) = c_i - 1;
22     end
23     start = 1+(c_i-1)*N;
24     finish = N+(c_i-1)*N;
25     scatter3(x(start:finish,1), x(start:finish,2), x(start:finish,3), 3, col(c_i), 'filled');
26     hold on
27 end
28 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
29 % Training and Testing
30 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
31 % Parameter intialization
32 w1 = 0.5;
33 w2 = 0.5;
34 w3 = 0.5;
35 b = 0.5;
36 error_rate = 1;
37 M = 0;
38 while error_rate > 0.1
39     M = M + 1; % Gradually increasing number of points used for training
40     % Indices for random selection of data for training and testing.
41     sp = randsample(C*N,M+1000);
42     for i = 1:M
43         y = b + w1*x(sp(i),1) + w2*x(sp(i),2) + w3*x(sp(i),3);
44         if(y < 0)
45             y = 0;
46         else
47             y = 1;
48         end
49         e = t(sp(i)) - y;
50         w1 = w1+e*x(sp(i),1);
51         w2 = w2+e*x(sp(i),2);
52         w3 = w3+e*x(sp(i),3);
53         b = b + e;
54     end
55     er = 0;
56     for i=M+1:M+1000
57         y = b + w1*x(sp(i),1) + w2*x(sp(i),2) + w3*x(sp(i),3);
58         if (y < 0)
59             y = 0;
60         else
61             y = 1;
```

```

    end
    e = abs(t(sp(i))-y);
    er = er + e;
65 end
    error_rate = er/1000
67 end
M
69 % Drawing a seperating plane
%Decide on a suitable showing range
71 x1Lim = [0 20];
x2Lim = [0 20];
73 [X1,X2] = meshgrid(x1Lim,x2Lim);
X3 = (w1*X1 + w2*X2 + b)/ (-w3);
75 reOrder = [1 2 4 3];
patch(X1(reOrder),X2(reOrder),X3(reOrder),'g');
77 grid on;
alpha(0.3);

```

2.2 FIGURES

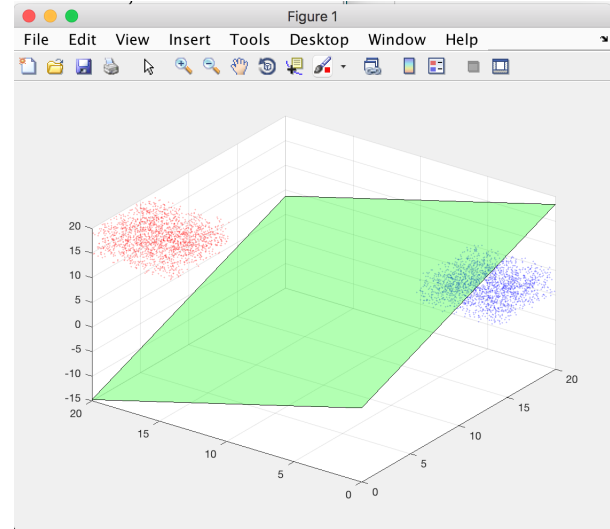
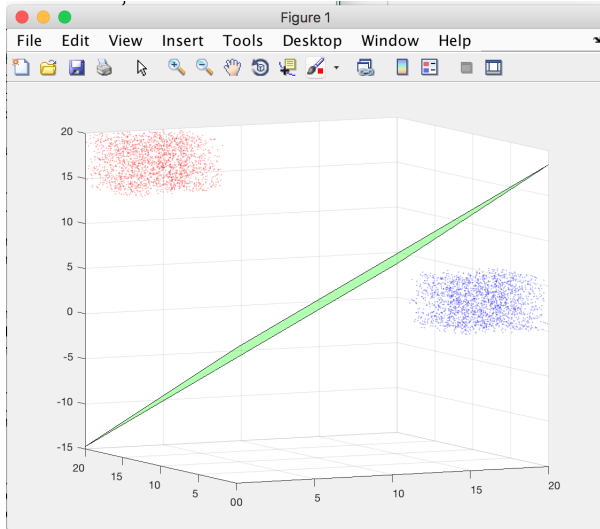


Figure 2.1: Separating plane obtained by training the basic perceptron. Separating plane is shown in green while two clusters are shown in red and blue.

```

>> exersice_4_1_2

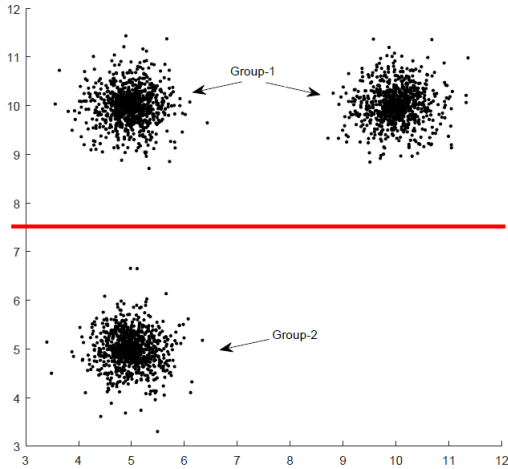
error_rate =          error_rate =
      0.5030              0

error_rate =          M =
      0.5070              3

```

Figure 2.2: The output of the program is shown. The error rate drops to 0 within 3 iterations.

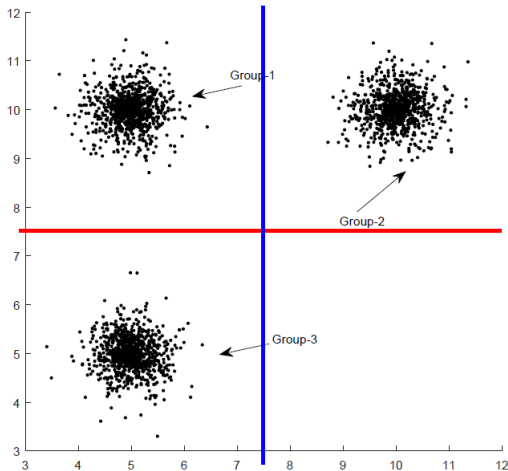
3 EXERSICE 4.1.3



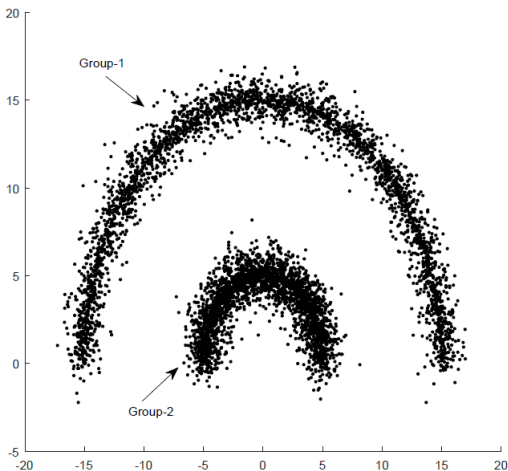
FOR PLOT 1: Since there are just 2 groups and both are linearly separable, a basic perceptron can separate the two groups. Line $y = 7.5$ CAN perfectly separate the two cluster groups.

$$y = \begin{cases} 1 & x_2 - 7.5 < 0 \\ 0 & x_2 - 7.5 > 0 \end{cases}$$

Where $b = -7.5$, $w_1 = 0$, and $w_2 = 1$.



FOR PLOT 2: Since there are 3 groups, a basic perceptron (a straight line separator) CANNOT separate these groups. They are however, separable using more than 1 perceptron.



FOR PLOT 3: There cannot be a straight line which can divide or separate the two groups. The groups are not linearly separable. Therefore, a basic perceptron CANNOT be used to separate the groups.

FOR PLOT 4: It can be seen that a straight line passing through points (17.5, 0) and (0, 17.5) **CAN** separate the two groups. Find the gradient (m) of separating line:

$$m = \frac{y_2 - y_1}{x_2 - x_1} = \frac{17.5 - 0}{0 - 17.5} = -1$$

Find the equation of separating line:

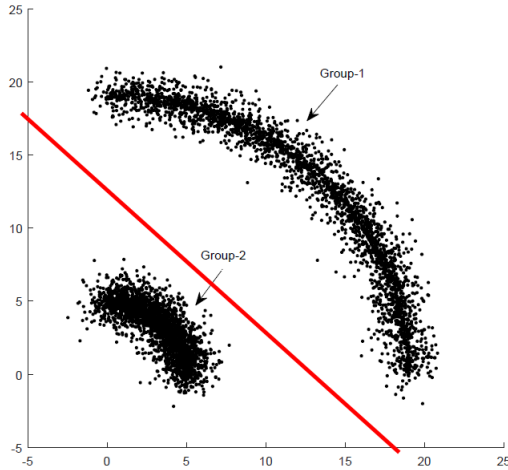
$$\begin{aligned} y - y_1 &= m(x - x_1) \\ y - 0 &= -1(x + 17.5) \\ y &= -x - 17.5 \\ y + x + 17.5 &= 0 \end{aligned}$$

Replacing the variables:

$$x_1 + x_2 + 17.5 = 0$$

$$y = \begin{cases} 1 & x_1 + x_2 + 17.5 > 0 \\ 0 & x_1 + x_2 + 17.5 < 0 \end{cases}$$

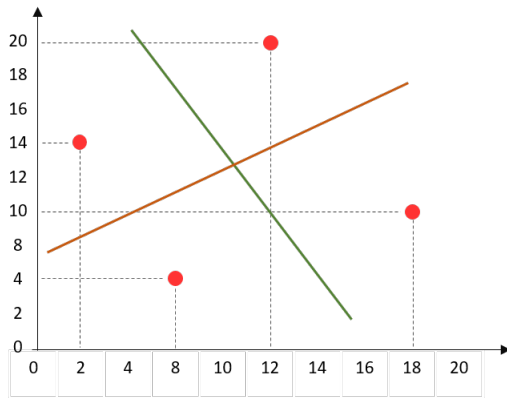
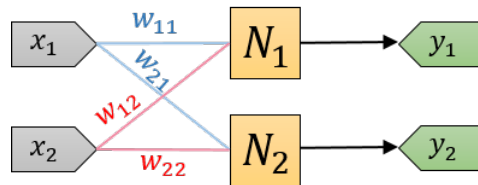
Where $w_1 = 1$, $w_2 = 1$, and $b = 17.5$.



4 EXERSICE 4.3.1

4.1 SOLUTION PART A

Since the search space is R^2 , the number of inputs to the neural network will be 2. Two lines can separate 4 linearly separable clusters. Therefore 2 perceptrons are required. Midpoints for x_1 separator:



$$MP_1 = \left(\frac{8 + 18}{2}, \frac{4 + 10}{2} \right) = (13, 7)$$

$$MP_2 = \left(\frac{18 + 12}{2}, \frac{10 + 20}{2} \right) = (15, 15)$$

$$MP_3 = \left(\frac{12 + 2}{2}, \frac{20 + 14}{2} \right) = (7, 17)$$

$$MP_4 = \left(\frac{2 + 8}{2}, \frac{14 + 4}{2} \right) = (5, 9)$$

Find the gradients (m_1 and m_2) of separating lines:

$$m_1 = \frac{y_2 - y_1}{x_2 - x_1} = \frac{15 - 9}{15 - 5} = \frac{3}{5} = 0.6$$

$$m_2 = \frac{y_2 - y_1}{x_2 - x_1} = \frac{7 - 17}{13 - 7} = -\frac{5}{3}$$

Equation of line 1 (Perceptron N_1):

$$y - y_1 = m(x - x_1)$$

$$y - 15 = \frac{3}{5}(x - 15)$$

$$y - 15 = \frac{3}{5}x - 9$$

$$y - 6 - \frac{3}{5}x = 0$$

Renaming variables:

$$-\frac{3}{5}x_1 + x_2 - 6 = 0$$

Therefore:

$$w_{11} = -\frac{3}{5}, w_{12} = 1, \text{ and } b_1 = -6$$

Equation of line 2 (Perceptron N_2):

$$y - y_1 = m(x - x_1)$$

$$y - 17 = -\frac{5}{3}(x - 7)$$

$$y - 17 = -\frac{5}{3}x + \frac{35}{3}$$

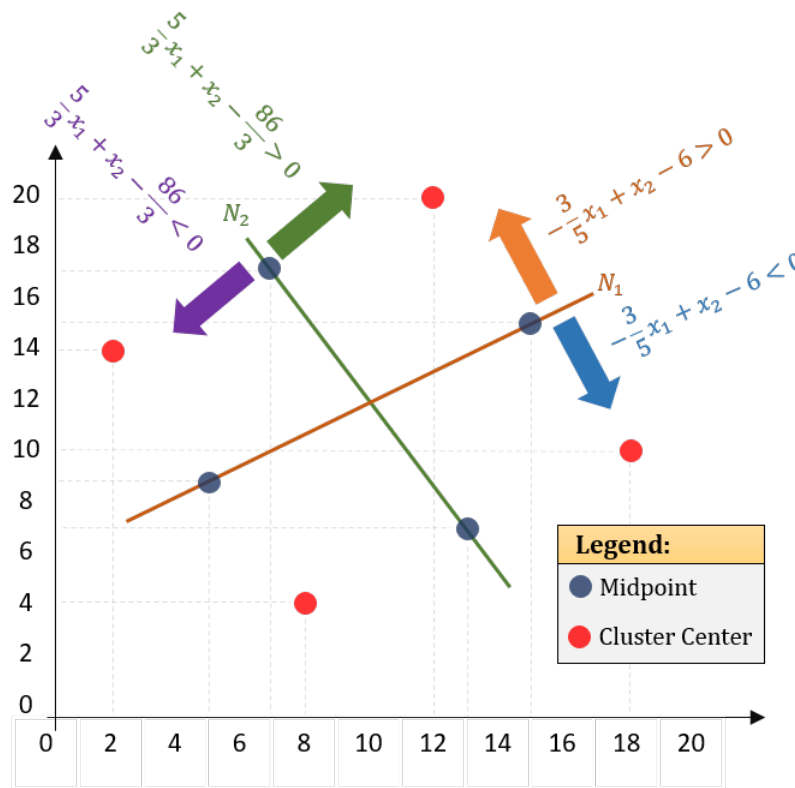
$$y - \frac{86}{3} + \frac{5}{3}x = 0$$

Renaming variables:

$$\frac{5}{3}x_1 + x_2 - \frac{86}{3} = 0$$

Therefore:

$$w_{21} = \frac{5}{3}, w_{22} = 1, \text{ and } b_2 = -\frac{86}{3}$$



4.2 SOLUTION PART B

DATA CREATION

```

1 p = [2 14 8 4 12 20 18 10];
2 a = 2;
3 N = 1000;
4 M = 1000;
5 m = zeros(4*N, 1);
6 s = a*ones(4*N, 1);
7 r = normrnd(m, s);

```

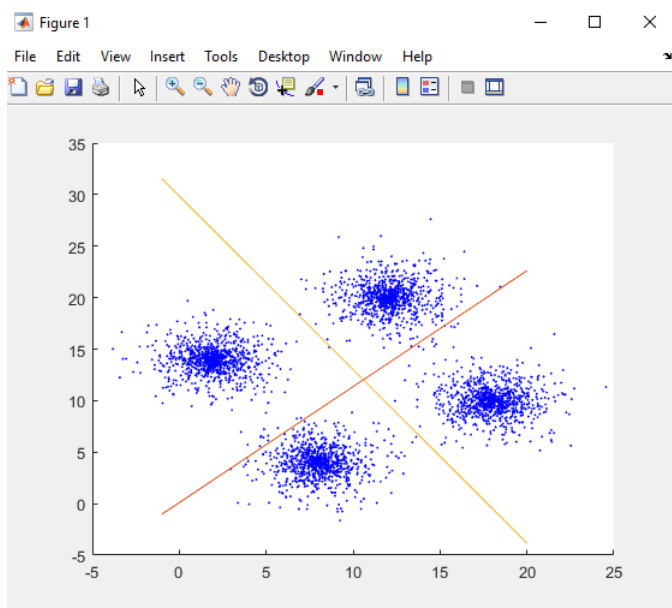


```

76 er=0;
77 for i = M+1:M+1000
78     y1 = b1 + w11*x(sp(i),1) + w12*x(sp(i),2);
79     if(y1<0)
80         y1=0;
81     else
82         y1=1;
83     end
84     e1 = abs(t(sp(i),1) - y1);
85     y2 = b2 + w21*x(sp(i),1) + w22*x(sp(i),2);
86     if(y2<0)
87         y2 = 0;
88     else
89         y2 = 1;
90     end
91     e2 = abs(t(sp(i),2) - y2);
92     if(e1 == 1 || e2 == 1)
93         er = er + 1;
94     end
95 end
96 er = er/1000;
97 fprintf('Iterations: %i, Error Rate: %.2f\n', M, er);
98 fprintf('Perceptron 1: %.2f x_1 + x_2 + %.2f = 0\n', w11/w12, b1);
99 fprintf('Perceptron 2: %.2f x_1 + x_2 + %.2f = 0\n', w21/w22, b2);
100 if(er < 0.1)
101     figure
102     scatter(x(:,1),x(:,2),3,'b','filled');
103     hold on
104     i = -1:20;
105     plot(i,(-b1-w11*i)/w12);
106     hold on
107     plot(i,(-b2-w21*i)/w22);
108     hold off
109 end

```

4.3 FIGURES



```

>> exercice_4_3_1_another_sol
Iterations: 1000, Error Rate: 0.01
Perceptron 1: -1.13 x_1 + x_2 + 3.84 = 0
Perceptron 2: 1.69 x_1 + x_2 + 103.53 = 0

```

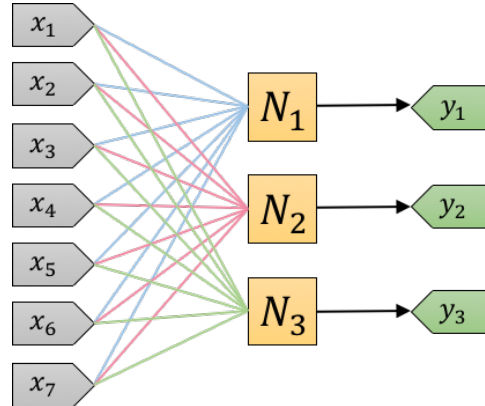
Figure 4.1: **Left:** Separating lines obtained by training.
Right: Output of program showing weights and biases.

5 EXERSICE 4.3.3

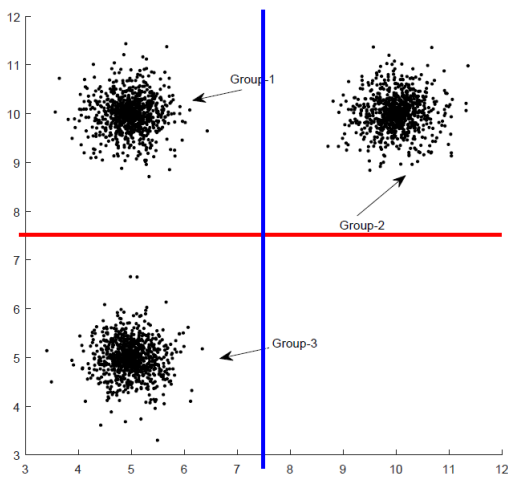
Since the search space is R^7 , the number of inputs to the neural network will be 7. The number of outputs and number of neurons depend on the number of clusters to classify. In classification neuron the output is usually kept binary. Therefore, number of minimal neurons is given by:

$$N_{neurons} = \lceil \log_2(N_{clusters}) \rceil = \lceil \log_2(5) \rceil = \lceil 2.32 \rceil = 3$$

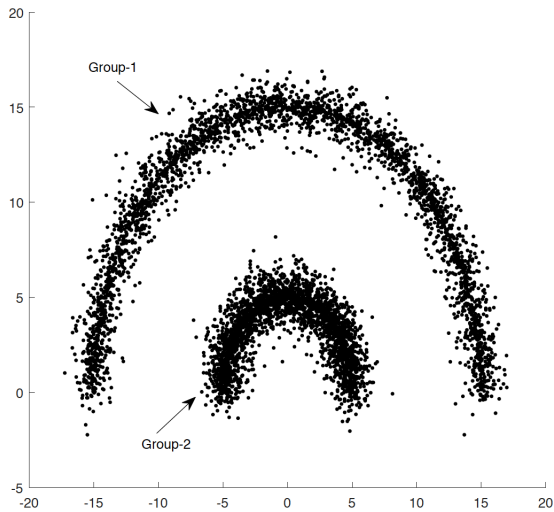
Another analogy to think of while determining number of minimal neurons required is stack and cut cake. If a person takes a whole cake and cuts it once, the cake is divided into two slices. Then if the cake's two peices are stacked on top of each other then cut again, we end up with four slices. If this is repeated, we end up with eight slices. So on an so forth. In this analogy the knife is a separating hyper-plane and the slices are clusters.



6 EXERSICE 4.3.4



FOR PLOT 1: Since the 3 groups are linearly sperable from each other, a single layer multi-perceptron **CAN** seperate these groups. A single layer with 2 neurons (each neuron having two inputs) can seperate these groups.



FOR PLOT 2: Since the groups are not linearly separable, a single layer multi-perceptron **CANNOT** separate them.

7 EXERCISE 5.1.1

$$z = f(x, y) = \frac{1}{2}x^2 + y^2$$

7.1 SOLUTION PART A

$$\frac{\partial z}{\partial x} = x$$

$$\frac{\partial z}{\partial y} = 2y$$

$$\therefore \nabla f(x, y) = (x, 2y)$$

$$\nabla f(1, 1) = (1, 2)$$

Therefore, required direction is $(1, 2)$ and maximum rate of change in this direction or greatest rate of change is $\|\nabla f\| = \sqrt{1^2 + 2^2} = \sqrt{5}$

7.2 SOLUTION PART B

Given vector:

$$u = (3, 4)$$

$$\|u\| = \sqrt{3^2 + 4^2} = \sqrt{9 + 16} = 5$$

Unit vector of u:

$$\hat{u} = \frac{u}{\|u\|} = \frac{(3, 4)}{5} = \left(\frac{3}{5}, \frac{4}{5}\right)$$

The directional derivative of f in the direction given by the vector u is given by:

$$D_u f = \nabla f \cdot \frac{u}{\|u\|}$$

$$= (x, 2y) \cdot \left(\frac{3}{5}, \frac{4}{5}\right)$$

$$= \frac{3}{5}x + \frac{8}{5}y$$

$$D_u f(1, 1) = \frac{3}{5}x + \frac{8}{5}y = \frac{3}{5}(1) + \frac{8}{5}(1) = \frac{11}{5}$$

7.3 SOLUTION PART C

Steepest Decent at point (1,1):

$$\nabla f(x, y) = (x, 2y) = (1, 2)$$

Direction unit vector of most rapid increase/decrease:

$$\hat{d} = \frac{\nabla f(x, y)}{\|\nabla f(x, y)\|} = \frac{\nabla f(1, 1)}{\|\nabla f(1, 1)\|} = \frac{(1, 2)}{\sqrt{1^2 + 2^2}} = \frac{(1, 2)}{\sqrt{5}}$$
$$\therefore \hat{d}_{increase} = \left(\frac{1}{\sqrt{5}}, \frac{2}{\sqrt{5}} \right)$$

To get direction of most rapid decrease at point (1,1), we multiply by negative:

$$\therefore \hat{d}_{decrease} = -\left(\frac{1}{\sqrt{5}}, \frac{2}{\sqrt{5}} \right) = \left(-\frac{1}{\sqrt{5}}, -\frac{2}{\sqrt{5}} \right)$$

8 EXERSICE 5.1.2

$$z = f(x, y) = xy^2$$

$$x = r \cos(\theta)$$

$$y = r \sin(\theta)$$

To find $\frac{\partial z}{\partial r}$, we have to find dependencies. One dependency of r comes through x and another through y:

$$\frac{\partial z}{\partial r} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial r} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial r}$$
$$= y^2 \cos(\theta) + 2xy \sin(\theta)$$

Expanding x and y:-

$$= (r \sin(\theta))^2 \cos(\theta) + 2(r \cos(\theta))(r \sin(\theta)) \sin(\theta)$$
$$= r^2 \sin^2(\theta) \cos(\theta) + 2r^2 \sin^2(\theta) \cos(\theta) = 3r^2 \sin^2(\theta) \cos(\theta)$$
$$\therefore \frac{\partial z}{\partial r} = 3r^2 \sin^2(\theta) \cos(\theta)$$

To find $\frac{\partial z}{\partial \theta}$, we have to find dependencies. One dependency of r comes through x and another through y:

$$\frac{\partial z}{\partial \theta} = \frac{\partial z}{\partial x} \frac{\partial x}{\partial \theta} + \frac{\partial z}{\partial y} \frac{\partial y}{\partial \theta}$$
$$= 2rxy \cos(\theta) - ry^2 \sin(\theta)$$

Expanding x and y:-

$$= 2r(r \cos(\theta))(r \sin(\theta)) \cos(\theta) - r(r \sin(\theta))^2 \sin(\theta)$$
$$= 2r^3 \cos^2(\theta) \sin(\theta) - r^3 \sin^3(\theta)$$
$$\therefore \frac{\partial z}{\partial \theta} = 2r^3 \cos^2(\theta) \sin(\theta) - r^3 \sin^3(\theta)$$

9 EXERCISE 5.1.3

$$y_1 = x_1 - 2x_2 + 3x_3$$

$$y_2 = 3x_1 + 4x_2 - 2x_3$$

$$yy = 7y_1 + 11y_2$$

To find $\frac{\partial yy}{\partial x_1}$, we have to find dependencies. One dependency of x_1 comes through y_1 and another through y_2 :

$$\begin{aligned} \frac{\partial yy}{\partial x_1} &= \frac{\partial yy}{\partial y_1} \frac{\partial y_1}{\partial x_1} + \frac{\partial yy}{\partial y_2} \frac{\partial y_2}{\partial x_1} \\ &= 7(1) + 11(3) = 40 \\ \therefore \frac{\partial yy}{\partial x_1} &= 40 \end{aligned}$$

To find $\frac{\partial yy}{\partial x_3}$, we have to find dependencies. One dependency of x_3 comes through y_1 and another through y_2 :

$$\begin{aligned} \frac{\partial yy}{\partial x_3} &= \frac{\partial yy}{\partial y_1} \frac{\partial y_1}{\partial x_3} + \frac{\partial yy}{\partial y_2} \frac{\partial y_2}{\partial x_3} \\ &= 7(3) + 11(-2) = -1 \\ \therefore \frac{\partial yy}{\partial x_3} &= -1 \end{aligned}$$

10 EXERCISE 5.1.4

$$y_1 = ax_1 + bx_2 + cx_3$$

$$y_2 = dx_1 + ex_2 + fx_3$$

$$yy_1 = py_1 + qy_2$$

$$yy_2 = ry_1 + sy_2$$

$$yyy = ky_1 + ly_2$$

To find $\frac{\partial yyy}{\partial y_1}$, we have to find dependencies. One dependency of y_1 comes through yy_1 and another through yy_2 :

$$\begin{aligned} \frac{\partial yyy}{\partial y_1} &= \frac{\partial yyy}{\partial yy_1} \frac{\partial yy_1}{\partial y_1} + \frac{\partial yyy}{\partial yy_2} \frac{\partial yy_2}{\partial y_1} \\ \therefore \frac{\partial yyy}{\partial y_1} &= kp + lr \end{aligned}$$

To find $\frac{\partial yyy}{\partial x_3}$, we have to find dependencies. One dependency of x_3 comes through yy_1 , yy_2 where both have a path to x_1 through y_1 and y_2 :

$$\begin{aligned} \frac{\partial yyy}{\partial x_3} &= \frac{\partial yyy}{\partial yy_1} \frac{\partial yy_1}{\partial x_3} + \frac{\partial yyy}{\partial yy_2} \frac{\partial yy_2}{\partial x_3} \\ \frac{\partial yy_1}{\partial x_3} &= \frac{\partial yy_1}{\partial y_1} \frac{\partial y_1}{\partial x_3} + \frac{\partial yy_1}{\partial y_2} \frac{\partial y_2}{\partial x_3} \\ \frac{\partial yy_2}{\partial x_3} &= \frac{\partial yy_2}{\partial y_1} \frac{\partial y_1}{\partial x_3} + \frac{\partial yy_2}{\partial y_2} \frac{\partial y_2}{\partial x_3} \\ \frac{\partial yyy}{\partial x_3} &= \frac{\partial yyy}{\partial yy_1} \left(\frac{\partial yy_1}{\partial y_1} \frac{\partial y_1}{\partial x_3} + \frac{\partial yy_1}{\partial y_2} \frac{\partial y_2}{\partial x_3} \right) + \frac{\partial yyy}{\partial yy_2} \left(\frac{\partial yy_2}{\partial y_1} \frac{\partial y_1}{\partial x_3} + \frac{\partial yy_2}{\partial y_2} \frac{\partial y_2}{\partial x_3} \right) \\ &= k(pc + qf) + l(rc + sf) \\ \therefore \frac{\partial yyy}{\partial x_3} &= kpc + kqf + lrc + lsf \end{aligned}$$

11 EXERCISE 5.1.5

$$z = T(y)$$

$$y = S(x)$$

$$T = \begin{bmatrix} ww_{11} & ww_{12} & ww_{13} \\ ww_{21} & ww_{22} & ww_{23} \\ ww_{31} & ww_{32} & ww_{33} \end{bmatrix}, \quad S = \begin{bmatrix} w_{11} & w_{12} & w_{13} \\ w_{21} & w_{22} & w_{23} \\ w_{31} & w_{32} & w_{33} \end{bmatrix}$$

$$\begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} ww_{11} & ww_{12} & ww_{13} \\ ww_{21} & ww_{22} & ww_{23} \\ ww_{31} & ww_{32} & ww_{33} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} w_{11} & w_{12} & w_{13} \\ w_{21} & w_{22} & w_{23} \\ w_{31} & w_{32} & w_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

Find $\frac{\partial z_2}{\partial x_3}$:

$$\frac{\partial z_2}{\partial x_3} = \frac{\partial z_2}{\partial y_1} \frac{\partial y_1}{\partial x_3} + \frac{\partial z_2}{\partial y_2} \frac{\partial y_2}{\partial x_3} + \frac{\partial z_2}{\partial y_3} \frac{\partial y_3}{\partial x_3} = ww_{21} \cdot w_{13} + ww_{22} \cdot w_{23} + ww_{23} \cdot w_{33}$$

Find $\frac{\partial z_3}{\partial x_1}$:

$$\frac{\partial z_3}{\partial x_1} = \frac{\partial z_3}{\partial y_1} \frac{\partial y_1}{\partial x_1} + \frac{\partial z_3}{\partial y_2} \frac{\partial y_2}{\partial x_1} + \frac{\partial z_3}{\partial y_3} \frac{\partial y_3}{\partial x_1} = ww_{31} \cdot w_{11} + ww_{32} \cdot w_{21} + ww_{33} \cdot w_{31}$$

Find $\frac{\partial z_1}{\partial ww_{12}}$:

$$\frac{\partial z_1}{\partial ww_{12}} = y_2$$

Find $\frac{\partial z_1}{\partial ww_{22}}$:

$$\begin{aligned} \frac{\partial z_1}{\partial ww_{22}} &= \frac{\partial z_1}{\partial y_1} \frac{\partial y_1}{\partial ww_{22}} + \frac{\partial z_1}{\partial y_2} \frac{\partial y_2}{\partial ww_{22}} + \frac{\partial z_1}{\partial y_3} \frac{\partial y_3}{\partial ww_{22}} \\ &= (ww_{11})(0) + (ww_{12})(x_2) + (ww_{13})(0) \\ &\therefore \frac{\partial z_1}{\partial ww_{22}} = ww_{12}x_2 \end{aligned}$$

Find $\frac{\partial z_2}{\partial ww_{13}}$:

$$\begin{aligned} \frac{\partial z_2}{\partial ww_{13}} &= \frac{\partial z_2}{\partial y_1} \frac{\partial y_1}{\partial ww_{13}} + \frac{\partial z_2}{\partial y_2} \frac{\partial y_2}{\partial ww_{13}} + \frac{\partial z_2}{\partial y_3} \frac{\partial y_3}{\partial ww_{13}} \\ &= (ww_{21})(x_3) + (ww_{22})(0) + (ww_{23})(0) \\ &\therefore \frac{\partial z_2}{\partial ww_{13}} = ww_{21}x_3 \end{aligned}$$

12 EXERCISE 5.2.1

Determining if the following matrix is positive definite:

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 2 & 3 & 3 \\ 1 & -1 & 1 \end{bmatrix}$$

$$1 > 0, \det \begin{bmatrix} 1 & 1 \\ 2 & 3 \end{bmatrix} = 3 - 2 > 0, \text{ and } \det(A) = 1(3 + 3) - 1(2 - 3) + 0(-2 - 3) = 6 + 1 = 7 > 0,$$

Hence A is positive definite.

Determining if the following matrix is positive definite:

$$B = \begin{bmatrix} 1 & 1 & 1 \\ 2 & 3 & 3 \\ 1 & -1 & 1 \end{bmatrix}$$

$1 > 0$, $\det \begin{bmatrix} 1 & 1 \\ 2 & 3 \end{bmatrix} = 3 - 2 > 0$, and $\det(A) = 1(3 + 3) - 1(2 - 3) + 1(-2 - 3) = 6 + 1 - 5 = 2 > 0$,
Hence B is positive definite.

13 EXERSICE 5.2.2

13.1 SOLUTION FOR PART A

Find Hessian matrix for following function:

$$f = x^2y + y^3x$$

$$H = \begin{bmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{bmatrix}$$

$$f_x = \frac{\partial f}{\partial x} = 2xy + y^3$$

$$f_{xx} = \frac{\partial f_x}{\partial x} = 2y$$

$$f_{xy} = \frac{\partial f_x}{\partial y} = 2x + 3y^2$$

$$f_y = \frac{\partial f}{\partial y} = x^2 + 3xy^2$$

$$f_{yy} = \frac{\partial f_y}{\partial y} = 6xy$$

$$f_{yx} = \frac{\partial f_y}{\partial x} = 2x + 3y^2$$

$$\therefore H = \begin{bmatrix} 2y & 2x + 3y^2 \\ 2x + 3y^2 & 6xy \end{bmatrix}$$

13.2 SOLUTION FOR PART B

Find Hessian matrix for following function:

$$g = x^2 + 2y^2 + 3z^2 + xy + yz + zx + xyz$$

$$H = \begin{bmatrix} g_{xx} & g_{xy} & g_{xz} \\ g_{yx} & g_{yy} & g_{yz} \\ g_{zx} & g_{zy} & g_{zz} \end{bmatrix}$$

$$g_x = \frac{\partial g}{\partial x} = 2x + y + z + yz$$

$$g_y = \frac{\partial g}{\partial y} = 4y + x + z + xz$$

$$g_z = \frac{\partial g}{\partial z} = 6z + y + x + xy$$

$$\therefore H = \begin{bmatrix} 2 & 1 + z & 1 + y \\ 1 + z & 4 & 1 + x \\ 1 + y & 1 + x & 6 \end{bmatrix}$$

14 EXERCISE 5.2.3

Prove that the following function is convex in the first quadrant of its domain:

$$f = x^3 + 3y^2 + 6xy \quad (14.1)$$

$$H = \begin{bmatrix} f_{xx} & f_{xy} \\ f_{yx} & f_{yy} \end{bmatrix}$$

$$f_x = \frac{\partial f}{\partial x} = 3x^2 + 6y$$

$$f_y = \frac{\partial f}{\partial y} = 6y + 6x$$

$$H = \begin{bmatrix} 6x & 6 \\ 6 & 6 \end{bmatrix}$$

In first quadrant, $x > 0$ and $y > 0$ which makes the first value of matrix greater than 0 (ie. $6x > 0$) and $\det(A) = 36x - 6 > 0$ which makes the hessian matrix positive definite. When hessian matrix of a function is positive definite in domain D, the function is convex in domain D.