## Assignment 1 Assignment 1

### 1

Using the code given in the appendix we created the scatterplot in figure 1.

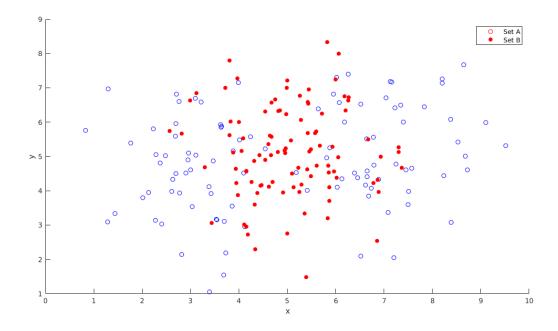


Figure 1: Scatterplot for the two classes.

The plot shows that there are at least three prototypes needed to approach a fairly well classification of these data. Two for set A, which should probably be located around (3, 4.5) and (7.5, 5.5), and one for set B somewhere around (5, 5).

### 2

The code in the appendix shows our implementation of the LVQ1 algorithm. We acquired the following results for the different settings.

### a 1 Prototype for class A and 1 prototype for class B

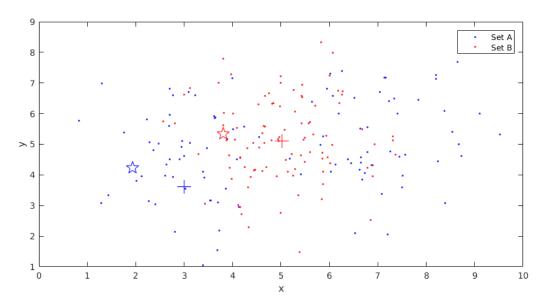


Figure 2: Scatterplot for the two classes.

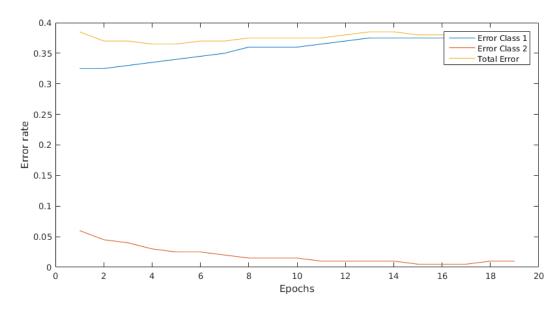


Figure 3: Training error rate to number of epochs.

As is expected with only one prototype per class, the prototype for class A is formed quite well, allowing it to correctly classify at least the data points that belong to class A. However, since class B is distributed in two groups, with A in between them, the prototype for class B is formed in the center of one of the two clusters, which means it can only correctly classify that cluster correctly. The other cluster will be incorrectly classified as class A.

### **Appendix**

### $../Code/Ass1_1.m$

```
hold off;
hold off;

kold on;
scatter(matA(:,1),matA(:,2), 'blue');
scatter(matB(:,1),matB(:,2), 'red', 'filled');
klabel('x'); ylabel('y'); legend('Set_A', 'Set_B');
```

### $../Code/Ass1_2.m$

```
load('data_lvq_A') % matA
1
2
   load ('data_lvq_B') % matB
3
4
   close all
5
   figure;
   plot (matA(:,1),matA(:,2), 'bp', 'markersize', 2);
7
   plot (matB(:,1), matB(:,2), 'rp', 'markersize', 2);
   xlabel('x'); ylabel('y');
   legend('Set_A', 'Set_B');
10
11
12
   data = [matA ; matB];
   data_labels = (floor((0:length(data)-1) * 2 / length(data))).;
13
   data = [data data_labels];
14
15
16
   % The prototypes
17
   w_A = 1;
18
   w_B = 1;
19
   w = zeros(w_A + w_B, ndims(data)+1);
21
   eta = 0.01;
22
   nrEpochs = 500;
23
24
   E_{-1} = zeros(1, nrEpochs);
25
   E_{-2} = \mathbf{zeros}(1, \text{nrEpochs});
26
27
   % Randomly initialize the prototypes between the minimum and maximum values
28
   % last value being their class
29
   \mathbf{for} \ i = 1 : \ \mathbf{size}(w, 1)
30
        if i \leftarrow w_A
31
            w(i,:) = [mean(matA) + rand()*2*std(matA)-std(matA) 0];
32
        else
33
            w(i,:) = [mean(matB) + rand()*2*std(matB)-std(matB) 1];
34
        end
35
   end
36
   plot(w(1:w_A,1), w(1:w_A,2), 'b+', 'markersize', 15);
37
   plot(w(w_A+1:size(w,1),1), w(w_A+1:size(w,1),2), 'r+', 'markersize', 15);
38
39
40
   for epoch = 1:nrEpochs
41
        % Training
42
        for point = 1 : size(data, 1)
```

```
43
                % Find the row with the nearest prototype
               rowMin = find(pdist2(data(point,1:2), w(:,1:2)) = min(pdist2(data(point,1:2)))
44
                    (1:2), w(:,1:2)), (1);
               % If the classes of the data point and the nearest prototype are the same
45
               if \ w(\operatorname{rowMin},end) == \operatorname{data}(\operatorname{point},\ end)
46
47
                    % Move the row closer to the data point
48
                    w(rowMin, 1:2) = w(rowMin, 1:2) + eta * (data(point, 1:2) - w(rowMin, 1:2));
49
                    w(rowMin, 1:2) = w(rowMin, 1:2) - eta * (data(point, 1:2) - w(rowMin, 1:2));
50
51
               end
52
         end
53
         \% Testing
54
55
         for point = 1 : size(data, 1)
               % Find the row with the nearest prototype
56
57
               rowMin = find(pdist2(data(point,1:2), w(:,1:2)) = min(pdist2(data(point
                    (1:2), w(:,1:2)), (1:2)
58
               if w(rowMin,end) ~= data(point, end)
59
                    if point <= size (matA, 1)
60
                         E_{-1}(epoch) = E_{-1}(epoch) + 1;
61
                         E_2(epoch) = E_2(epoch) + 1;
62
63
                    end
64
               \mathbf{end}
65
         end
66
         E = E_{-1} + E_{-2};
67
          if (epoch > 10 && var(E(:, epoch - 4: epoch)) < 0.05)
68
69
               epoch
70
               E_{-1}(:, epoch+1:end) = [];
               E_{-2}(:, epoch+1:end) = [];
71
72
              E(:, epoch+1:end) = [];
73
               break
74
         end
75
    end
76
    plot(w(1:w<sub>-</sub>A,1), w(1:w<sub>-</sub>A,2), 'bp', 'markersize', 15);
77
    \mathbf{plot}\left(\mathbf{w}(\mathbf{w}.\mathbf{A}+1:\mathbf{size}\left(\mathbf{w},1\right)\;,1\right)\;,\;\;\mathbf{w}(\mathbf{w}.\mathbf{A}+1:\mathbf{size}\left(\mathbf{w},1\right)\;,2\right)\;,\;\;\mathsf{'rp'}\;,\;\;\;\mathsf{'markersize'}\;,\;\;15)\;;
78
79
80
    figure;
    plot (E<sub>-</sub>1/200)
81
82
    \mathbf{hold} on;
83
    plot (E_2/200);
    plot (E/200);
    legend('Training_error_Class_1', 'Training_error_Class_2', 'Total_training_Error');
    xlabel ('Epochs')
    ylabel('Error_rate')
```

# Pattern Recognition Practical 4

Group 24: Maikel Withagen (s1867733) Steven Bosch (s1861948)

October 8, 2015