

SMAI-ASSIGNMENT 2

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Abstract. The document is written to answer the following questions

- (1) Prove that the single-sample perceptron algorithm will always converge to a solution if one exists.
- (2) Implement Perceptron-based Linear Discriminant Functions, that is, Single-sample perceptron, Single-sample perceptron with margin, Relaxation algorithm with margin, Widrow-Hoff or Least Mean Squared (LMS) Rule
- (3) Implement a simple supervised, feed-forward, back-propagation network for the problem of optical character recognition for digits 0 and 7.

1. PROOF OF CONVERGENCE OF PERCEPTRON

Let the data be Linearly separable and let \hat{a} be solution vector perpendicular to the required plain. So, we know that $\hat{a}^t y_i > 0 \forall i$.

The Gradient Descent Equation is:

$$a(k+1) = a(k) + y^k$$

We subtracting $\alpha \hat{a}$ from both sides,

$$a(k+1) - \alpha \hat{a} = a(k) - \alpha \hat{a} + y^k$$

Squaring both sides

$$\|a(k+1) - \alpha \hat{a}\|^2 = \|a(k) - \alpha \hat{a} + y^k\|^2$$

$$\|a(k+1) - \alpha \hat{a}\|^2 = \|a(k) - \alpha \hat{a}\|^2 + \|y^k\|^2 + 2(a(k) - \alpha \hat{a})^t y^k$$

y^k was a misclassified point. Hence $\hat{a}^t y^k \leq 0$, so:

$$\|a(k+1) - \alpha \hat{a}\|^2 \leq \|a(k) - \alpha \hat{a}\|^2 + \|y^k\|^2 - 2\alpha \hat{a}^t y^k$$

Now, $\alpha^t y^k > 0$. Also,

We define β as the maximum length of the pattern vector. ie. $\beta^2 = \max_i \|y_i\|^2$. As this is a positive value, this is the maximum we can achieve from the second term.

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We define γ as the smaller inner product of the \hat{a} with any pattern vector. ie. $\gamma = \min_i [\hat{a}^i y_i]$. This is the minimum we can achieve from the third term which can be subtracted.

Hence our equation becomes,

$$\|a(k+1) - \alpha \hat{a}\|^2 \leq \|a(k) - \alpha \hat{a}\|^2 - 2\alpha\gamma + \beta^2$$

α was a random constant so let

$$\alpha = \beta^2 / \gamma.$$

Our equation reduces to,

$$\|a(k+1) - \alpha \hat{a}\|^2 \leq \|a(k) - \alpha \hat{a}\|^2 - \beta^2$$

Say that after some k_0 steps, the equation converges,

ie. $\|a(k_0+1) - \alpha \hat{a}\|^2 = 0$, hence,

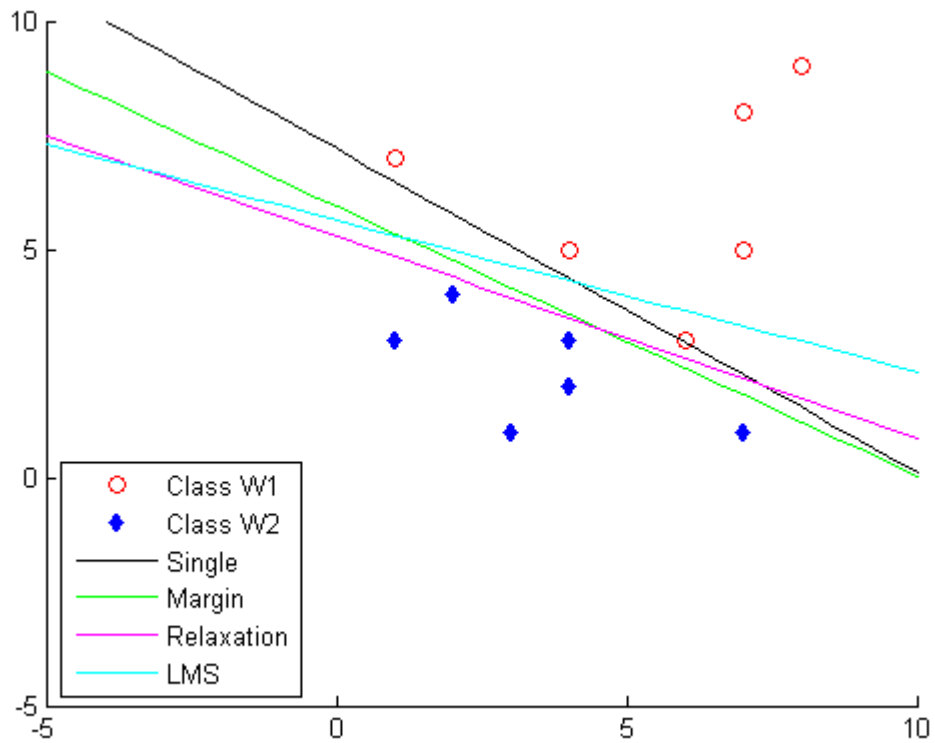
$$0 \leq \|a(k) - \alpha \hat{a}\|^2 - k_0 \beta^2$$

or we can write it as

$$k_0 = \|a(1) - \alpha \hat{a}\|^2 / \beta^2$$

2. LDF - GRADIENT DESCENT ALGORITHM

2.1. GRAPH



```

a1 : For single sample
a2 : For single sample with margin
a3 : For Relaxed
a4 : For LMS

```

```

Round 1
a = randi(5,[1,size(dataFinal,1)])
a1 =

```

```

-2.9500    0.2900    0.4100

```

```

Elapsed time is 0.005273 seconds.

```

4

SMAI

```
a = randi(5,[1,size(dataFinal,1)])  
a2 =
```

```
-5.9300    0.5900    1.0000
```

Elapsed time is 0.013373 seconds.

```
a = [-1,1,1];  
a3 =
```

```
-1.2917    0.1083    0.2451
```

Elapsed time is 0.915025 seconds.

```
a = [-1,1,1];  
a4 =
```

```
-1.5246    0.0898    0.2704
```

Elapsed time is 0.433580 seconds.

Round 2

```
a = randi(5,[1,size(dataFinal,1)])  
a1 =
```

```
-2.9500    0.2800    0.4600
```

Elapsed time is 0.006685 seconds.

```
a = randi(5,[1,size(dataFinal,1)])  
a2 =
```

```
-5.6500    0.5700    0.9400
```

Elapsed time is 0.011457 seconds.

```
a = [-5,1,1]  
a3 =
```

```
-5.0753    0.5898    0.8480
```

Elapsed time is 0.854140 seconds.

```
a = [-5,1,1]
```

```
a4 =
```

```
-5.0480    0.6693    0.6749
```

```
Elapsed time is 0.003884 seconds.
```

```
Round 3
```

```
a = randi(5,[1,size(dataFinal,1)])
```

```
a1 =
```

```
-3.2000    0.2200    0.6700
```

```
Elapsed time is 0.007913 seconds.
```

```
a = randi(5,[1,size(dataFinal,1)])
```

```
a2 =
```

```
-6.2000    0.5700    1.1200
```

```
Elapsed time is 0.011992 seconds.
```

```
a = [1,1,1]
```

```
a3 =
```

```
-5.0753    0.5898    0.8480
```

```
Elapsed time is 0.829555 seconds.
```

```
a = [1,1,1]
```

```
a4 =
```

```
0.6437   -0.1371    0.0696
```

```
Elapsed time is 0.019652 seconds.
```

```
But the answer obtained is not correct.
```

As one can see the time is almost same for any initial value of A for a1 and a2.

The time taken by a3 has a little dependence on A.

The time taken by a4 varies greatly with the initial value of A. If A is closer to a minimum then it Even though this may lead to wrong output.

ADDING DIFFERENT MARGIN

Round 1

b = 0.5

a2 =

-5.7100	0.5500	1.0000
---------	--------	--------

Elapsed time is 0.013883 seconds.

b = 0.1

a3 =

-1.2917	0.1083	0.2451
---------	--------	--------

Elapsed time is 0.932693 seconds.

Round 2

b = 1.5

a2 =

-14.9100	1.5500	2.3800
----------	--------	--------

Elapsed time is 0.033113 seconds.

b = 1.1

a3 =

-5.8612	0.6099	0.9103
---------	--------	--------

Elapsed time is 1.243246 seconds.

Round 3

b = 0.05

a2 =

-3.1200	0.2100	0.6500
---------	--------	--------

Elapsed time is 0.010261 seconds.

b = 0.01

a3 =

-1.2819 0.0810 0.2800

Elapsed time is 0.809574 seconds.

As one can see the time is greatly dependent on the value of the margin. Higher value of margin usually means more iterations. In almost all the cases a2 is much faster than a3.

2.2. CODE

```

1 clear;
2 clc;
3 data1 = [1,1,7;1,6,3;1,7,8;1,8,9;1,4,5;1,7,5]';
4 data2 = -[1,3,1;1,4,3;1,2,4;1,7,1;1,1,3;1,4,2]';
5 dataFinal = [data1,data2];
6
7 scatter(data1(2,:), data1(3,:), 'r'), axis([-5,10,-5,10]), hold on;
8 scatter(-data2(2,:), -data2(3,:), 'b', 'd', 'filled'), axis([-5,10,-5,10]);
9
10 x = -10:0.5:10;
11 tic, a1 = Single(dataFinal), toc;
12 tic, a2 = Margin(dataFinal), toc;
13 tic, a3 = Relax(dataFinal), toc;
14 tic, a4 = LMS(dataFinal), toc;
15
16 y1 = (a1(1)+a1(2)*x)/(-a1(3));
17 plot(x,y1, 'k'), axis([-5,10,-5,10]);
18 y2 = (a2(1)+a2(2)*x)/(-a2(3));
19 plot(x,y2, 'g'), axis([-5,10,-5,10]);
20 y3 = (a3(1)+a3(2)*x)/(-a3(3));
21 plot(x,y3, 'm'), axis([-5,10,-5,10]);
22 y4 = (a4(1)+a4(2)*x)/(-a4(3));
23 plot(x,y4, 'c'), axis([-5,10,-5,10]);
24 legend('Class W1', 'Class W2', 'Single', 'Margin', 'Relaxation',
        'LMS', 'Location', 'southwest'), hold off;
25
26
27 % In case of linearly classifiable data we assume a model  $G(x) = A*Y$ .
28 % So for properly classifying the points we need to determine the
    correct
29 % value of A. For this we first define an error function J and then
    try and
30 % minimize it. Now using the technique of Gradient Decent we update
    the value of A.
31 %
32 %  $A(k+1) = A(k) + (ETA)*J'$ ;

```

```

33% J' is the derivate of the error function we are trying to minimize
    and ETA is the learning rate.
34% Depending on our choice of J our algorithm changes.

35function [ a ] = Relax( dataFinal )
36%UNTITLED9 Summary of this function goes here
37% Detailed explanation goes here
38    a = [-1,1,1];
39    n = 0.1;           %learning rate
40    b = 0.1;           %margin
41    thres = 100000;
42    check = zeros(length(dataFinal),1);
43    for i = 0:thres
44        j = 1+mod(i,length(dataFinal));
45        q = a*dataFinal(:,j);
46        if q < b
47            a = a +
48                n*(((b-q)^2)/(norm(dataFinal(:,j)')^2))*(dataFinal(:,j)');
49        else
50            check(j)=1;
51        end
52        if sum(check) == length(check)
53            break;
54        end
55    end
56end
57
58% Till the previous algorithm more weightage was give to outliers.
    So to
59% correct this issue it is normalized with norm of the distance.
    Also the
60% concept of relaxation is introduced. If there is under relaxation
    then
61% the learning is slow but if there is over relaxation then it may
62% overshoot.
63% J' is  $((AY-B)^2)/(|Y|^2)*Y_k$ ;
64
65% This process was comparitively computationally expensive. But this
66% algorithm also gave correct results and this algorithm has the
    capacity
67% to hendle far off cases.

68function [ a ] = LMS( dataFinal )
69%UNTITLED8 Summary of this function goes here
70% Detailed explanation goes here
71    a = [-1,1,1];      % selected after trying for MANY possible
    a's.
72    % a = randi(5,[1,size(data,1)]);

```



```

73     n = 0.0005;           %learning rate
74     % b = randi(5,[1,size(data,2)]);
75     b = ones(1,size(dataFinal,2));
76     thres = 0.000025;
77     check = zeros(length(dataFinal),1);
78     j =1;
79     i =0;
80     while norm((n)*dataFinal(:,j)'+(a*dataFinal(:,j) - b(j))) > thres
81         j = 1+mod(i,length(dataFinal));
82         a = a - (n)*dataFinal(:,j)'+(a*dataFinal(:,j) - b(j));
83         i = i+1;
84     end
85 end
86
87 % Untill now all the algorithms that we saw used misclassified
88 % sample for
89 % updation purpose. This had a major issue as due to this outliers
90 % etc
91 % had large influence. So to correct this issue this method was
92 % introduced.
93 % In this method all the points are considered. It is assumed that
94 % each
95 % point has a distance bi from the line.
96 % But as correct distance cant be estimated error is introduced. So,
97 %  $e = AY - B$ ;
98 % Using this we try and update A to get a favorable vector.
99
100 % This algorithm give very different results for different
101 % initallised
102 % values. Even then it was not able to correctly identify all the
103 % points.
104
105 function [ a ] = Single( dataFinal )
106 %UNTITLED5 Summary of this function goes here
107 % Detailed explanation goes here
108 a = randi(5,[1,size(dataFinal,1)]);
109 n = 0.09;           %learning rate
110 b = 0;              %margin
111 thres = 50000;
112 check = zeros(length(dataFinal),1);
113 for i = 0:thres
114     j = 1+mod(i,length(dataFinal));
115     q = a*dataFinal(:,j);
116     if q <= b
117         a = a + n*dataFinal(:,j)';
118         check(j)=0;
119     else
120         check(j)=1;

```

```

115         end
116         if sum(check) == length(check)
117             break;
118         end
119     end
120 end

121 function [ a ] = Margin( dataFinal )
122 %UNTITLED5 Summary of this function goes here
123 % Detailed explanation goes here
124 a = randi(5,[1,size(dataFinal,1)]);
125 n = 0.09; %learning rate
126 b = 0.5; %margin
127 thres = 50000;
128 check = zeros(length(dataFinal),1);
129 for i = 0:thres
130     j = 1+mod(i,length(dataFinal));
131     q = a*dataFinal(:,j);
132     if q <= b
133         a = a + n*dataFinal(:,j)';
134         check(j)=0;
135     else
136         check(j)=1;
137     end
138     if sum(check) == length(check)
139         break;
140     end
141 end
142 end

143
144 % This is a little improvement of the previous technique. In this
145 % algorithm
146 % the concept margin is introduced so that the partitioning line is
147 % not too
148 % close to the points. This helps in obtaining a more general line .
149 % So our J is  $AY - b$ .
150
151 % This methos also gave very efficient results and was able to
152 % classify the
153 % results correctly in a short time. It is also computationally not
154 % that
155 % expensive. But if the data set was not as good as the one provided
156 % then
157 % this algorithm might not have such results.
158 % As a margin was kept it gave pretty good seperation line.

```

3. NEURAL NETWORK

```

154 clear;
155 clc;
156 a = fopen('optdigits-orig.cv');
157 i = 1;
158 while ~feof(a)
159     if mod(i,33)==0
160         sbl_v(i/33,:)=fgetl(a);
161     else
162         data_v(i,:)=fgetl(a);
163     end
164 i=i+1;
165 end
166 a = fopen('optdigits-orig.tra');
167 i = 1;
168 while ~feof(a)
169     if mod(i,33)==0
170         sbl_t(i/33,:)=fgetl(a);
171     else
172         data_t(i,:)=fgetl(a);
173     end
174 i=i+1;
175 end
176
177 sbl_t = sbl_t(:,2);
178 sbl_v = sbl_v(:,2);
179
180 a = data_t == '1';
181 data_t(a) = 255;
182 a = data_t == '0';
183 data_t(a) = 0;
184 a = data_v == '1';
185 data_v(a) = 255;
186 a = data_v == '0';
187 data_v(a) = 0;
188 data_t = double(data_t);
189 data_v = double(data_v);
190
191 a = find(sbl_t == '7' | sbl_t == '0');
192 train = zeros(8,9,length(a));
193 for i = 1:length(a)
194     if sbl_t(a(i)) == '7'
195         b = 1;
196     else
197         b = 0;
198     end

```

```

199     train(:,1:8,i) =
200         imresize(data_t((a(i)-1)*33+1:a(i)*33-1,:),0.25);
201     train(:,9,i) = b;
202 end
203 a = find(sbl_v == '7' | sbl_v == '0');
204 validate = zeros(8,9,length(a));
205 for i = 1:length(a)
206     if sbl_v(a(i)) == '7'
207         b = 1;
208     else
209         b = 0;
210     end
211     validate(:,1:8,i) =
212         imresize(data_v((a(i)-1)*33+1:a(i)*33-1,:),0.25);
213     validate(:,9,i) = b;
214 end
215 clear data_t;
216 clear data_v;
217 clear sbl_t;
218 clear sbl_v;
219 clear a;
220 clear b;
221 clear i;

222 clear;
223 clc;
224 load ./Q3_data.mat;
225 t = train;
226 v = validate;
227 d = size(t,1)*(size(t,2)-1)+1; % because 8X8 + x0;
228 nH = 39;
229 c = 2;
230 n = 0.1;
231 iter = 100;
232 thres = 10;
233 Wij = unifrnd(-1/sqrt(d),1/sqrt(d),d,nH);
234 Wjk = unifrnd(-1/sqrt(nH),1/sqrt(nH),nH,c);
235 netj = zeros(nH,1);
236 netk = zeros(c,1);
237 error = zeros(1,size(v,3));
238 e = zeros(1,size(v,3));
239
240 for j = 1:iter
241     for i = 1:size(t,3)
242         X = 1;
243         X(2:d) = t(:,1:8,i); % 1Xd
244         netj = (X*Wij); % 1Xd.dXnH => 1XnH

```

```

245
246     Y = (1+exp(-netj)).^(-1);           % 1XnH
247     netk = (Y*Wjk);                     % 1XnH.nHXc => 1Xc
248
249     Z = (1+exp(-netk)).^(-1);
250     if t(1,end,i) == 1
251         tc = [1,1];
252     else
253         tc = [0,0];
254     end
255     if (norm(Z)> 0.4 & t(1,end,i) == 1) | (norm(Z)<= 0.4 &
        t(1,end,i) == 0)
256         e(i) = 0;
257     else
258         e(i) = 1;
259     end
260     Dk = ((tc-Z).*(Z.*(1-Z)));
261     Wjk = Wjk + n*Y'*Dk;
262     Dj = ((Dk*Wjk').*(Y.*(1-Y)));
263     Wij = Wij + n*X'*Dj;
264 end
265 if sum(e) < thres
266     break;
267 end
268 end
269
270 for i = 1:size(v,3)
271     X = 1;
272     X(2:d) = v(:,1:8,i);                 % 1Xd
273     netj = (X*Wij);                       % 1Xd.dXnH => 1XnH
274
275     Y = (1+exp(-netj)).^(-1);             % 1XnH
276     netk = (Y*Wjk);                       % 1XnH.nHXc => 1Xc
277
278     Z = (1+exp(-netk)).^(-1);
279     if (norm(Z)> 0.5 & v(1,end,i) == 1) | (norm(Z)<= 0.5 &
        v(1,end,i) == 0)
280         error(i) = 0;
281     else
282         error(i) = 1;
283     end
284 end
285 q = find(error == 1);
286 fprintf('Number of misclassified "7": %d\n',sum(v(1,end,q)));
287 fprintf('Total misclassified samples: %d\n',length(q));
288 fprintf(' : %d\n',length(q));

```

For the purpose of recognizing letters '0' and '7'

Since the number of training samples was 390 I decided to use 39 hidden states.

Wij is 65X39:

0.00852377110196400 0.0168167108194295 0.0906906995135764 -0.0341521508520852
0.00915446012818496 -0.0768488757602270 0.0502128894586879 0.0279794871327043
0.0621662893305959 0.0759595787983735 0.105897485065351 0.0225581981968135
0.117390193773649 0.0214895782727926 -0.102615639285905 -0.0620640004927365
-0.0754294689332468 -0.0178795726500848 -0.0110213735800909 0.0727804090855225
-0.0439326189495888 0.00715505421676795 -0.0875192492853352 -0.0310464378044200
0.109104246623387 0.0316997788661457 -0.0117638930064226 0.0371779579678653
-0.0980467288909962 0.0637766875999936 -0.00738271533583162 0.102774214462131
0.0867514018326965 -0.0982989098190917 0.0850625005994987 0.0588708982926175
0.0776162569032200 0.123112547436863 -0.0543850089758245
0.0595261495178468 0.00714056952943048 0.0811628260928037 -0.0543772358575069
0.0604521855394331 -0.00463698755753382 0.0149190270129171 -0.00962944054533800
0.0348426394699797 0.105751071978376 -0.00130163968318400 -0.0932268375130779
0.0605169515811133 0.113086409361776 -0.105243020771068 -0.0918293699951344
0.0633270255349999 0.0276696150246281 0.0551462743085977 0.0609341949144762
0.0309797976289314 0.122647670708964 -0.0962274418719907 -0.0644266090832099
-0.0869059280930350 -0.0830993920358655 0.0315682349228637 -0.0687050112872726
-0.0788778677112658 -0.0894439736089960 -0.0555879414724181 0.111940823629141
-0.102190073481889 0.115545801728561 0.0729077843414233 0.0529256065176578
-0.0688455307960449 0.101037817289172 -0.0965540374183246
-0.109830646143318 0.0643703207794488 -0.0510154309450226 -0.111173648015054
-0.104774750720731 -0.0512938753621463 -0.0830650986088040 -0.120837949145864
-0.0633115424174788 0.0630578449510213 0.0577851075628820 0.0501532617037082
-0.0201904223622786 -0.0870993633519403 0.103613592078124 -0.0134111986724326
-0.123415622604114 -0.0284690187788827 0.111677824945701 -0.109004834976096
-0.0303006290010616 -0.0721013123039012 0.0441962771640571 -0.00625855068888152
-0.0296402637734693 0.00543342804006497 0.0366871065203631 -0.0827904583583095
0.0916374172128980 -0.00170096551716990 0.0908825317463318 -0.0264418823150248
0.0484401807211761 0.0687638588174358 -0.0529297910571880 -0.0207345205815580
0.0774297023751123 -0.0674558460383876 0.0288951106614635
-0.0775996271336807 -0.0354858986491108 0.0629885797464172 0.00329630169952517
0.00366246233055242 0.0927470099417445 0.0246810386538579 0.0397235116995779
-0.108078142634465 -0.121952161062692 -0.0100547123556675 0.111037991501745
0.0464084145353480 0.0967093501366771 -0.00670515450603525 0.101880141465277
0.00630307902247911 -0.0832242425839677 0.0336513946901884 -0.0340988001202030
0.0246847402122981 0.0270451310606321 -0.106392477300683 0.0737746673311120
0.0444435754981999 0.0470601757378302 -0.127280448215099 0.108799655935066
0.0189361108794941 0.0979002756323457 0.0566501830871377 0.123363766078253
-0.0211705114952527 -0.0158793591601205 0.0195125379740267 0.0807670769482862
0.124674283331374 0.0745102561387080 0.0601811156435985
0.000771526493604177 0.0630723614547452 0.0674020062402522 -0.0444395094839052
-0.107461400179258 -0.120991075915206 -0.0903457458305285 -0.0292480171837627
-0.0328254138786684 -0.0679071115991325 -0.0554740791000339 -0.120848877603020

-0.00398370138512121 0.125916167422714 0.0511030084448825 -0.120662612974428
0.119034558869075 -0.0640480667925480 0.108543865352031 -0.0121778929587795
-0.0550411125496900 -0.00121992069175365 0.0805420814576100 0.0865317399464556
-0.0300564171047141 -0.0371411208113495 0.0773612724774837 -0.0146339639866768
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Wjk is 39X2

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```

With the help of these weights I was able to get ZERO error in most of the cases(Weights are initial.

Procedure followed:

- First the data was extracted from the files taken from the online resouce.
- Then it was normalized and converted into favourable format.
- After some testing and using $m/10$ ($m = \#$ test cases) 39 was chosen as nH .
- Some random initial weights were picjed and then used for forward mapping.
- Using the training data and back propagation technique some weights were learned.
- They helped in classifying the incoming test cases.

3.1. DIAGRAM

