ECG 703- Spring 2021

## Homework Set 2

a) Create a set of 2D data ( $d=2$, features: $\boldsymbol{x}=\left(x_{1}, x_{2}\right)$ )
b) Randomly create a set of 20 data points $(N=20)$ such that for each point $\boldsymbol{x}=\left(x_{1}, x_{2}\right)$, the coordinates $x_{1}, x_{2}$ be integers. $x_{1}$, $x_{2}$ are to be limited to the $[-30,+30]$ range and uncorrelated.
c) Choose the line $x_{1}+2 x_{2}-1.1=0$ as your target function, where the points on one side of the line map to $y=+1$ ( $\left.f=x_{1}+2 x_{2}-1.1>0\right)$ and the other points map to $y=-1$ $\left(f=x_{1}+2 x_{2}-1.1<0\right)$. Now, you have a set of 20 data points $(\boldsymbol{x}, y)$ as your separable data points.
d) Plot the points on the 2D plane labeling them with " + " or "-" or Red and Blue.
e) Implement and run the simple perceptron algorithm. You must write the perceptron code from scratch as opposed to using the code in software packages. Make sure to include declarations next to code lines for ease of readability. How many iterations does it take to arrive at the solution boundary (estimated target function)? Plot 4 iterations including the final one showing the boundary line at each iteration. Draw function $f$ line on the last plot and explain why they are different.

## Assignment 2: Perceptron Student: Francisco Mata

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## Part a)

```
                INSTRUCTIONS
                Create a set of 2D data ( d=2, features: x=(x_1,x_2) )
                making some arbitrary vectors to use on the 2d
                    Implementation
                    Two row vectors were created for x1 and x2 with 20 columns each
format compact
%initilizing and allocating x1 and x2 vectors
disp('Initializing and allocating x1 and x2 vectors');
x1=zeros(1,20);
x2=zeros(1,20);
disp(x1);
disp(x2);
Initializing and allocating x1 and x2 vectors
    Columns 1 through 13
        0
    0
    Columns 14 through 20
            0
    Columns 1 through 13
            0
    0 0
        Columns 14 through 20
            0
```


## Part b)

## INSTRUCTIONS

Randomly create a set of 20 data points ( $\mathrm{N}=20$ ) such that for each point $x=\left(x \_1, x \_2\right)$, the coordinates $x \_1, x \_2$ be integers. $x \_1, x \_2$ are to be limited to the $[-30,+30]$ range and uncorrelated.

```
            Implementation
                A matrix of two rows and 20 columns with random variables was
                created. The first row vector was set to x1 and the second row
                to x2.
X = randi([-30,30], [2,20]); % array of 2x20 random values within - 30
    and 30 range
x1 = X(1,:); % making the first row of X equal to x1
x2 = X(2,:); % making the first row of X equal to x1
```


## Part c)

## INSTRUCTIONS

Choose the line $\mathrm{x} \_1+2 \mathrm{x} \_2-1.1=0$ as your target function, where the points on one side of the line map to $y=+1$
(f=x_1+2x_2-1.1>0) and the other points map to $y=-1$
(f=x_1+2x_2-1.1<0). Now, you have a set of 20 data points (x,y) as your separable data points.

## Implementation

A for loop grabs every random point's $x 1$ and $x 2$ from the random vector and puts those values in the target line function. If the output is posive x 1 and x 2 are store in a vector varialbe called "y_plus_x1" and "y_plus_x2". If the output is negative then $x 1$ and $x 2$ are store in y_minus_x1 and y_minus_x2. Figure 1 shows the first plot with the target line and the 20 random points.

```
% line below is to make the target line function the same range as
% x1 maximum and minimum values
x1_sort = sort(x1, 'ascend');
% The variables and vectors declared below were used for the first
    plot
% to identified the points above and below the target line
jp=1;
jm=1;
y_plus_x1=zeros(1,2);
y_plus_x2=zeros(1,2);
y_minus_x1=zeros(1,2);
y_minus_x2=zeros(1,2);
y = zeros(1,2);
% the for loop below checks for all random points (x1,x2) and
    allocates
% them on 2 different vectors; one for the points above the line and
% another one for the ones below the line
for i=1:20
    f = x1(i) + 2*x2(i) - 1.1; % target function
    if f > 0
        y_plus_x1(jp) = x1(i);
        y_plus_x2(jp) = x2(i);
```

```
            y(i) = 1;
            jp= jp+1;
        else
            y_minus_x1(jm)=x1(i);
            y_minus_x2(jm)=x2(i);
            y(i)=-1;
            jm = jm+1;
        end
end
%Vlength=length(x1)
% the lines below displays the y output, -1 or 1
disp('y output of +1(above the line) or -1(below the line');
disp(y);
% the lines below display thr random points above the line and below
disp('Final Vector Values From Target Function');
disp('Random Points Above Line');
disp(y_plus_x1);
disp(y_plus_x2);
disp('Random Point Below Line');
disp(y_minus_x1);
disp(y_minus_x2);
%disp(jp);
%disp(jm);
x2_lf = -(x1_sort./2) +0.55; % target line function x1 + 2x2 - 1.1 = 0
clf(figure(1))
figure(1)
plot(x1, x2, 'b*'); % plotting the random values of x1 and x2
title('Random Points and Target Line Function');
suptitle('Figure 1')
xlabel('x1');
ylabel('x2');
hold on
plot(x1_sort,x2_lf, 'm' ); %plotting the line function, the x-value is
    the
                            %range of the maximun and minimum of x1 row
legend('Random Points', 'Target Line Function')
grid on
hold off
y output of +1(above the line) or -1 (below the line
    Columns 1 through 13
            1 
    1 -1
    Columns 14 through 20
            1 1 1 1 1 1 1 1 -1 
Final Vector Values From Target Function
Random Points Above Line
    Columns 1 through 13
        21
    24 16
        Columns 14 through 15
```



Figure 1
Random Points and Target Line Function


## Part d)

```
            INSTRUCTIONS
            Plot the points on the 2D plane labeling them with "+" or "_"
                    or Red and Blue.
                    Implementation
                    The figure below plots the random points above the target line
                    with a "+" in blue for points above the target line and a "+"
                    in red for the points below the line
clf(figure(2))
figure(2)
plot(y_plus_x1, y_plus_x2, 'b+');
title('Labeling Random Points');
suptitle('Figure 2');
```

```
xlabel('x1');
ylabel('x2');
hold on
plot(y_minus_x1, y_minus_x2, 'r+');
grid on
plot(x1_sort, x2_lf, 'm')
legend('Points Above', 'Points Below', 'Target Function')
hold off
```

Figure 2
Labeling Random Points


## Part e)

INSTRUCNTIONS: Implement and run the simple perceptron algorithm. You must write the perceptron code from scratch as opposed to using the code in software packages. Make sure to include declarations next to code lines for ease of readability. How many iterations does it take to arrive at the solution boundary (estimated target function)? Plot 4 iterations including the final one showing the boundary line at each iteration. Draw function f line on the last plot and explain why they are different.

Implementation: A fixed starting line was chosen with given (w) values which are $w(x)$ _old. Then it goes into a while loop with if statements and a couple for lopps that take the first misclassified point, takes the coordinates and produces new weights (w). Once it implements the new set of weights a new function line is created, and then it goes through the loop again to recheck how many and which are the misclassified points. There is also a varialble counter that counts the iterations. Figure 3 shows a plot with the path and number of iterations the line function is taking. Figure 4 shows the first 3 iterations and the last itiration along with the target line in red and the random points above and below. The last itiration is in magenta color and it should be close to the target line function. The starting and first itiration is in blue, the second in green, third in cyan and the last one in magenta.

The total number of iterations is shown in figure 3 and at the bottom of the outputs section.
The last function line and the target line are not the same for several resons. First, there are infinite number of line functions that can separate the data linearly using 20 random points. Second, depending on how clotter the data is around the target function, the learning line will get closer towards the target line but not exact unless the number of data points keeps increasing.

```
% starting line function is 0.5x2 + x1 + 0 = 0, which for this
% case is fixed and is the first itiration since it checks for
% misclassified points.
% below are the starting weight (w) values the x range of the
% training line and the learning rate
n = 0.005; % learning rate
wO_old = 0; % starting weight 0
w1_old = 1; % starting weight 1
w2_old = 0.5; % starting weight 2
x1_new = -30:30; % x1 range of values used on plots
%starting line function 0.5x2 + x1 + 0 = 0
new_y_plus_xl=zeros(1,2); % stores xl points above the learinig line
new_y_plus_x2=zeros(1,2); % stores x2 points above the learinig line
new_y_minus_xl=zeros(1,2); % stores xl points below the learinig line
new_y_minus_x2=zeros(1,2); % stores x2 points below the learinig line
new_y = zeros(1,20); % stores new y output, -1 or 1
misclassified_x1 = zeros(1,2); % holds x1 misclassified values
misclassified_x2 = zeros(1,2); % holds x2 misclassified values
miss = 1; % counts misclassified point each iteration
iterations = 0; %initializing iteration counter
clf(figure(3)) % clears the figure 3 each time the code is run
clf(figure(4)) % clears the figure 4 each time the code is run
while miss ~= 0 % while the counter,"miss", that holds the number of
    % misclassified points is not equal to zero
% incrementing variables used with vectors to store x1 and x2 points
% above and below the learing line
new_jp=1;
new_jm=1;
    % statememt below decides if between the starting weight values
    % given above or the new weight values depending on the number
    % of iterations
    if iterations == 0
        w0 = w0_old;
        w1 = w1_old;
        w2 = w2_old;
    else
        w0 = w0_new;
        w1 = w1_new;
        w2 = w2_new;
```


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```
end
% the for loop below takes the current weight (w) values and puts
% them with the line function and rechecks for new misclassified
% values
for i=1:20
    f2 = w1*x1(i) + w2*x2(i) + w0;
    if ((f2 > 0) && (new_jp < jp))
            new_y_plus_x1(new_jp) = x1(i);
            new_y_plus_x2(new_jp) = x2(i);
            new_y(i) = 1;
            %disp(new_y);
            new_jp = new_jp + 1;
        elseif ((f2 <= 0)&&(new_jm < jm))
            new_y_minus_x1(new_jm)=x1(i);
            new_y_minus_x2(new_jm)=x2(i);
            new_y(i)=-1;
            %disp (new_y);
            new_jm = new_jm + 1;
        end
    end
    miss=0;
    Y_should_be = zeros(1:2); % initializing vector for the output
                                    % value the misclassified point
                        % should be
% the loop below checks for the target line y output of -1 and 1
% and compares it with each line iteration to see if there is any
% misclassified points. If there are any misclassified points
% then the variable "miss" will count up, however the "miss"
% number value should decrease per iteration
for i = 1:20
        if new_y(i) ~= y(i)
            miss = miss + 1;
            misclassified_x1(miss) = x1(i);
            misclassified_x2(miss) = x2(i);
            y_should_be(miss) = y(i);
                disp("misclassified points");
                    disp(misclassified_x1);
            disp(misclassified_x2);
            disp(miss);
    end
end
    iterations = iterations + 1; % iteration counter
w0_new = w0 + n*y_should_be(1)*1; % creates new weight, w0
w1_new = w1 + n*y_should_be(1)*misclassified_x1(1); % ", w1
w2_new = w2 + n*y_should_be(1)*misclassified_x2(1); % ", w2
```

```
    % f_line variable below receives a returning line from
    % funciton "fun", which takes weight values and x range
    f_line = fun(w0, w1, w2, x1_new);
    % figure 3 below plots all the iterations
    figure(3)
    plot(x1_new, f_line, 'b')
    title('Total Number of Iterations')
    suptitle('Figure 3');
% subtitle('3');
    text = ['Number of Iterations = ', num2str(iterations)];
    legend(text)
    hold on
    xlabel('x1');
    ylabel('x2');
    grid on
    % figure 4 below plots 4 iterations and the target line. Blue,
    % green, and cyan are the first 3 iterations and magenta is
    % the last iteration. The target line function is in red
    figure(4)
    plot(y_plus_x1, y_plus_x2, 'b+');
    title('4 Iterations with Target Function Looping in Real Time')
    suptitle('Figure 4');
    xlabel('x1');
    ylabel('x2');
    hold on
    plot(y_minus_x1, y_minus_x2, 'r+');
    grid on
    hold on
    plot(x1_sort,x2_lf, 'r') % plotting target line
    hold on
    if iterations == 1
            first_iteration = f_line;
            plot(x1_new, f_line, 'b')
            %first_iteration = f_line;
            hold on
elseif iterations == 2
            second_iteration = f_line;
            plot(x1_new, f_line, 'g')
            hold on
elseif iterations == 3
            third_iteration = f_line;
    plot(x1_new, f_line, 'c')
    hold on
end
```

```
plot(xl_new, f_line, 'm') % plots the last iteration line for
```

plot(xl_new, f_line, 'm') % plots the last iteration line for
hold off % figure 4

```
    hold off % figure 4
```

end

```
% the lines below was used for testing the misclassified outputs -1
    and 1
% disp("new_y output of +1(above the line) or -1(below the line)");
disp("New_y vector output to be verified with first set of y output");
disp(new_y);
disp("Final Vector Values from Perceptrom Algorithm");
disp("Random Points Above Last Itiration Line ");
disp(new_y_plus_x1);
disp(new_y_plus_x2);
disp("Random Point Below Last Itiration Line");
disp(new_y_minus_x1);
disp(new_y_minus_x2);
disp("misclassified points");
disp(misclassified_x1);
disp(misclassified_x2);
disp(miss);
disp("The value the misclassified should be");
disp(y_should_be);
disp("Number of iterations");
disp(iterations);
%disp(first_iteration);
clf(figure(5))
figure(5)
plot(x1_new, first_iteration, 'b');
title('4 Iterations with Target Function');
suptitle('Figure 5');
hold on
plot(x1_new, second_iteration, 'g');
plot(x1_new, third_iteration, 'c');
plot(x1_new, f_line, 'm')
plot(x1_sort,x2_lf, 'r')
plot(y_plus_x1, y_plus_x2, 'b+');
plot(y_minus_x1, y_minus_x2, 'r+');
legend('First Iteration', 'Second Iteration', 'Third Iteration', 'Last
    Iteration', 'Target Function','Points above', 'Points Below')
grid on
hold off
    % funtion below returns a line function from a set of weights given
    function f1 = fun(w0, w1, w2, x1_new)
        x2_new = ((-w1*x1_new) - w0*1)/w2 ;
        f1 = x2_new;
    end
```


## Output Section

```
New_y vector output to be verified with first set of y output
    Columns 1 through 13
    lllllllllllllll
```

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```
Columns 14 through 20
    1
Final Vector Values from Perceptrom Algorithm
Random Points Above Last Itiration Line
        Columns 1 through 13
            21
    24 16
        Columns 14 through 15
            26 22
        Columns 1 through 13
\begin{tabular}{llllllllllll} 
& 4 & 12 & 19 & 30 & 22 & 30 & 24 & 21 & 5 & 10 & 14
\end{tabular}
    Columns 14 through 15
        5 -1
Random Point Below Last Itiration Line
        2 18 -25 -29 21
        -1 -17 
misclassified points
        2 2 18 -15 21
        -1 lllll
        O
    The value the misclassified should be
        0 0
Number of iterations
        1 4
```

Figure 2
Labeling Random Points


Figure 3


Figure 4


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Figure 5


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