

**4741****Home work 1 (Question 3 and 4)****Submitted by*****Aman jain (aj644), Vineel Yellapantula (vy45)***

```
In [145]: using Random # load the Random package to generate random numbers  
using Plots # load the Plots plotting library
```

Note: you can add text written in Markdown to this notebook to explain your work and your conclusions by adding a new cell (using the "+" icon on the toolbar), and setting it to be a markdown cell (by toggling the "Code" dropdown on the toolbar to say "Markdown" instead).

For examples of how to use Markdown, look at the examples in this notebook, or [this cheatsheet \(https://github.com/adam-p/markdown-here/wiki/Markdown-Cheatsheet\)](https://github.com/adam-p/markdown-here/wiki/Markdown-Cheatsheet).

## Generate data

Generate 50 random datapoints in 2 dimensions. Note that the last column is fixed at 1 to capture the offset

```

In [146]: ### DO NOT CHANGE THIS CODE BLOCK ###
# this code sets the random number generator,
# so everyone in the class will have consistent results
Random.seed!(3)

# set dimensions
n = 50
d = 2
# generate input: the rows of X are the data points
# notice the last column encodes the offset
X = [randn(n, d) ones(n)]

# Generate the target function w
w = randn(d+1)

# Store the correct decision for each datapoint in y
# sign of x, returning -1, 0, or +1
y = sign.(X*w);

```

```

In [147]: w

```

```

Out[147]: 3-element Array{Float64,1}:
 -1.1190703941943563
  1.2975091637129141
  0.4736813190084895

```

```

In [148]: # plot data points X and color according to y
# '!' instructs to plot on the same graph
scatter(X[y.== -1,1], X[y.== -1,2], color=:red, marker=:circle, label="y=-1")
scatter!(X[y.== 1,1], X[y.== 1,2], color=:blue, marker=:diamond, label="y=1")

# xlims and ylims define the graph size on X and Y axes

xlims!((-2,2))
ylims!((-2,2))

```

```

Out[148]:

```

**I have slightly modified below function to plot 2 Ws - one true value of  $w$ ,  $w_{\text{h}}$ , and another value of  $w$  as suggested by perceptron algorithm**

Although, Function will work even if a single value of  $w$  is provided.

```
In [151]: """
This function takes data points X, plots them, and colors them correctly according to y.
This function also plots the line  $w^T x = 0$ .
We are plotting just the first two dimensions of x;
the last (offset) coordinate of w sets the offset of the line from the origin.
"""

function plot_perceptron(X,y,wᵃ,w)
    # plot data points X and color according to y
    p = scatter(X[y.==1,1], X[y.==1,2], color=:blue, marker=:diamond, label="y=1")
    scatter!(p, X[y.==-1,1], X[y.==-1,2], color=:red, marker=:circle, label="y=-1")

    # plot vector w
    xlsamples = [minimum(X[:,1]), maximum(X[:,1])]
    if w[2]!=0
        plot!(p, xlsamples, [-(w[1]*x1 + w[3])/w[2] for x1 in xlsamples], color=:black, label="classification line")
    end

    # plot vector wᵃ
    if wᵃ[2]!=0
        plot!(p, xlsamples, [-(wᵃ[1]*x1 + wᵃ[3])/wᵃ[2] for x1 in xlsamples], color=:blue, label="True decision boundary")
    end

    # set figure limits
    xlims!(p, (-2,2))
    ylims!(p, (-2,2))

    return p
end
```

Out[151]: plot\_perceptron

```
In [152]: # Double check our function by plotting the points with the truth vector
plot_perceptron(X,y,wᵃ)
```

Out[152]:

## a) Code the missing parts of the Perceptron algorithm

Inputs to Perceptron are:

- required arguments:
  - $X$  = data points
  - $Y$  = true values
- optional arguments:
  - maxsteps = maximum number of updates to  $w$  (default: 100)
  - $w$  = initial guess of  $w$  (default: random)

Perceptron should start with the input guess of  $w$ , make at most maxsteps updates, and return the final  $w$ .

***I have filled the below code for perceptron.***

***2 print commands tells us if perceptron is convergence for a given  $X$  and  $Y$  or not - if convergent how many iterations it took.***

***Below function is linear in time as well.***

```

In [153]: function perceptron(X,y;
            maxsteps=100, #
            w = rand(size(X,2))) # initial guess for w

    for i = 1:maxsteps
        y_pred = sign.(X*w)
        y_missclassified = ones(size(X,1)) - (y_pred.==y)*1
        if sum(y_missclassified) == 0
            print("algorithm converged ", i-1, "iterations only")
            break
        end
        y_multiply = y.*y_missclassified
        w_incremental = X'*y_multiply
        w = w+w_incremental
        if i >= maxsteps
            print("algorithm did not converged even after", maxsteps, "iterations")
        end
    end

    ## YOUR CODE HERE p

    return w
end

```

Out[153]: perceptron (generic function with 1 method)

```

In [154]: ## Rough work
           #y = [1 2 3]'
           #y_pred = [1 2 3]'
           #y_missclassified = ones(3) - (y_pred.==y)*1
           #sum(y_missclassified)

           #y.*y_missclassified
           #i = 2
           #print("breaking off at ", i, "only")

```

## b) Run perceptron on linearly separable data set

**For this part, perceptron converges after 6 iterations only.**

**Plot containing both  $w_{\text{true}}$  and  $w_{\text{calculated}}$  has been provided. Please note that both of these are not equal as many classification boundaries are possible in given linearly separable data set.**

```
In [155]: w_final = perceptron(X,y)
```

```
algorithm converged 6 iterations only
```

```
Out[155]: 3-element Array{Float64,1}:  
-15.2565198534639  
16.427392611962926  
5.851999190276057
```

```
In [156]: plot_perceptron(X,y,wh,w_final)
```

```
Out[156]:
```

## c) Run perceptron on data sets of size $n = 20, 100$ , and $1000$

**Some key observations:**

- number of iterations for convergence increased with  $n$ .
- number of iterations for convergence in case b ( $n=50$ ), was found to be in between number of iterations for  $n=20$  and  $n=100$ .
- convergence was achieved in all three cases.  $n = 20, 100$  and  $1000$ .

**Perceptron run for  $n = 20$**

```
In [161]: # set dimensions
n_20 = 20
n_100 = 100
n_1000 = 1000
d = 2
# generate input: the rows of X are the data points
# notice the last column encodes the offset
X_20 = [randn(n_20, d) ones(n_20)]
X_100 = [randn(n_100, d) ones(n_100)]
X_1000 = [randn(n_1000, d) ones(n_1000)]

# Generate the target function w
w2 = randn(d+1)
```

```
Out[161]: 3-element Array{Float64,1}:
 0.6579088352742319
-1.23718415848714
-1.0562109659819299
```

```
In [162]: # Store the correct decision for each datapoint in y
# sign of x, returning -1, 0, or +1
y_20 = sign.(X_20*w2);
X_20;
w_20_final = perceptron(X_20,y_20)
plot_perceptron(X_20,y_20,w2,w_20_final)
```

algorithm converged 4 iterations only

```
Out[162]:
```

### **Perceptron run for n = 100**

```
In [163]: # Store the correct decision for each datapoint in y
# sign of x, returning -1, 0, or +1
y_100 = sign.(X_100*w2);
w_100_final = perceptron(X_100,y_100)
plot_perceptron(X_100,y_100,w2,w_100_final)
```

algorithm converged 6 iterations only

```
Out[163]:
```

**Perceptron run for  $n = 20$** 

```
In [164]: # Store the correct decision for each datapoint in y
# sign of x, returning -1, 0, or +1
y_1000 = sign.(X_1000*w2_0);
w_1000_final = perceptron(X_1000,y_1000)
plot_perceptron(X_1000,y_1000,w2_0,w_1000_final)
```

algorithm converged 17 iterations only

Out[164]:

## d) Run perceptron on data set of size $n = 1000$ with dimension $d = 10$

**Perceptron run for  $n = 1000$  in 10 dimensions.**

**As data is in more than two dimensions, plot (which is 2-D) is not very intuitive in terms of line of separation. Even though perceptron converges in 13 iterations.**



```
In [99]: # set dimensions
d_10 = 10
# generate input: the rows of X are the data points
# notice the last column encodes the offset
X_1000_10 = [randn(n_1000, d_10) ones(n_1000)]
# Generate the target function w
w10 = randn(d_10+1)
```

```
Out[99]: 11-element Array{Float64,1}:
 0.09295287529837899
 1.2405275121117183
-0.7572815213974944
 0.6898626257391324
-0.4241884982054228
 0.4203561481382908
-0.13689000099878507
-0.4384395621992105
 0.40744988014903866
-0.7395093467288412
-0.7082525169144553
```

```
In [101]: # Store the correct decision for each datapoint in y
# sign of x, returning -1, 0, or +1
y_1000_10 = sign.(X_1000_10*w10);
w_1000_10_final = perceptron(X_1000_10,y_1000_10)
plot_perceptron(X_1000_10,y_1000_10,w10,w_1000_10_final)
```

algorithm converged 13iterations only

```
Out[101]:
```

## e) Run perceptron on dataset from (d) 100 times.

Plot a histogram for the number of updates that the algorithm takes to converge.

For this portion, make sure your perceptron implementation picks the next consideration point  $x(t)$  randomly instead of deterministically.

***For this problem, slightly modified version of perceptron function is used.***

***Rather than final  $W$ , this function return the number of iterations it took for convergence.***

```
In [125]: function perceptron_iterations(X,y;
           maxsteps=100, #
           w = rand(size(X,2))) # initial guess for w

    for i = 1:maxsteps
        y_pred = sign.(X*w)
        y_missclassified = ones(size(X,1)) - (y_pred.==y)*1
        if sum(y_missclassified) == 0
            return i
            break
        end
        y_multiply = y.*y_missclassified
        w_incremental = X'*y_multiply
        w = w+w_incremental
    end

    ## YOUR CODE HERE p

    return i
end
```

Out[125]: perceptron\_iterations (generic function with 1 method)

***no\_exp is variable governing number of iterations.***

***please note that  $h[j] = 100$  means that perceptron did not converge within 100 iterations for that experiment.***

```
In [137]: no_exp = 100
h = zeros(no_exp)
for j = 1:no_exp
    # set dimensions
    d_10 = 10
    # generate input: the rows of X are the data points
    # notice the last column encodes the offset
    X_1000_10 = [randn(n_1000, d_10) ones(n_1000)]
    # Generate the target function w
    w10 = randn(d_10+1)
    # Store the correct decision for each datapoint in y
    # sign of x, returning -1, 0, or +1
    y_1000_10 = sign(X_1000_10*w10);
    h[j] = perceptron_iterations(X_1000_10,y_1000_10)
end
```

```
In [138]: h
```

```
Out[138]: 100-element Array{Float64,1}:
```

```
11.0
10.0
13.0
16.0
15.0
12.0
11.0
14.0
12.0
15.0
20.0
2.0
7.0
⋮
23.0
13.0
14.0
10.0
20.0
8.0
12.0
14.0
11.0
11.0
15.0
93.0
```

```
In [144]: using Plots
```

```
p1 = histogram(h,xlabel="number of experiments", ylabel = "Number of iterations for convergence")
plot(p1)
```

```
Out[144]:
```

*As histogram above shows, most of the experiments converged within 30 iterations.*

## g) Add an outlier. What happens?

```
In [158]: ### DO NOT CHANGE THIS CODE BLOCK ###
# (re)generate data with same random seed, adding one outlier
Random.seed!(3)

# set dimensions
n = 50
d = 2
# generate input
X = [randn(n, d) ones(n)]
w = randn(d+1)
y = sign.(X*w);
y[1] *= -1 # flip classification of the first data point
```

Out[158]: 1.0

```
In [159]: # run the perceptron and see what happens
perceptron(X,y,maxsteps=50)
```

algorithm did not converged even after 50 iterations

```
Out[159]: 3-element Array{Float64,1}:
-8.91156595752908
 9.86614075726375
 6.851999190276057
```

***As we can see from the print message, perceptron did not converge after 50 iterations.***

## h) Try out one idea to fix perceptron

***One way to fix this problem is allow mis-classifications for some numbers of (x,y).***

***In following perceptron\_fix function, this problem is fixed by breaking out as soon as number of misclassifications goes below n\*tolerance.***

***Default tolerance in below function is 5%.***

```

In [122]: function perceptron_fix(X,y;
           maxsteps=100, #
           w = rand(size(X,2)), tolerance = 0.05) # initial guess for w

    for i = 1:maxsteps
        y_pred = sign.(X*w)
        y_missclassified = ones(size(X,1)) - (y_pred.==y)*1
        if sum(y_missclassified) < size(X,1)*tolerance
            print("algorithm converged ", i-1, "iterations only")
            break
        end
        y_multiply = y.*y_missclassified
        w_incremental = X'*y_multiply
        w = w+w_incremental
        if i >= maxsteps
            print("algorithm did not converged even after", maxsteps, "iterations")
        end
    end

    ## YOUR CODE HERE p

    return w
end

```

Out[122]: perceptron\_fix (generic function with 1 method)

```
In [123]: perceptron_fix(X,y,maxsteps=50)
```

algorithm converged 5iterations only

```

Out[123]: 3-element Array{Float64,1}:
 -11.871244775722971
  16.686249800219308
   5.915068922393004

```

**As can be seen above perceptron converges after 5 iterations.**

## Question 4

***Question 1 took about 3 hours. Even though i did this part on python - which i am familier with, It took time as i was exploring plotting lat longs - which was new for me.***

***Question 2 took about 1 hours.***

***Question 3 took about 2 hours. It was quite suprising for me. It was first time i was working on Julia but it was very quick. Starting code ofcourse helped a lot.***

***In total it took around 8 hours - including clean up, discussion & reviewing it with partner and making it ready for submission.***

In [ ]: