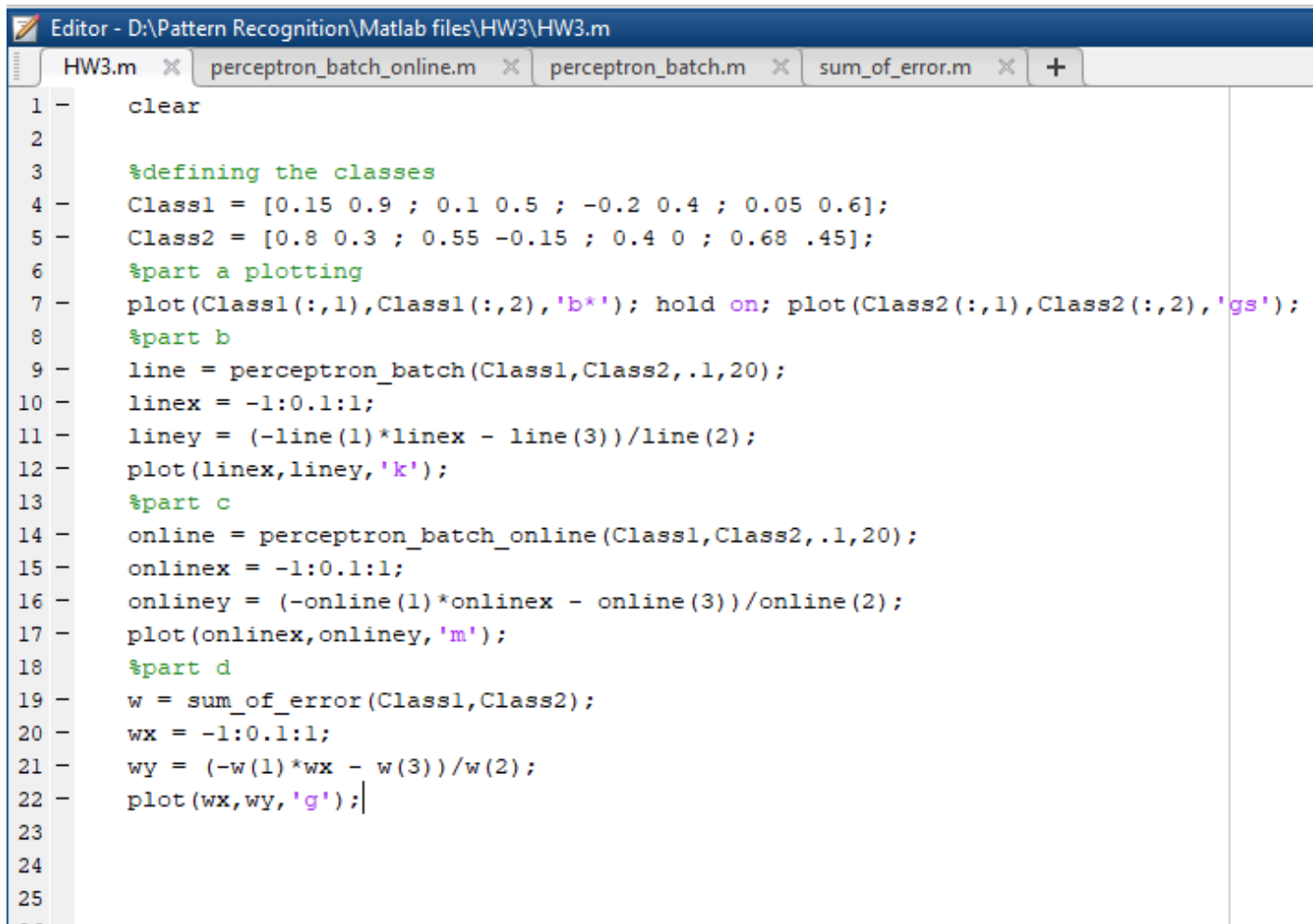


Homework 2

Main Code



```
Editor - D:\Pattern Recognition\Matlab files\HW3\HW3.m
HW3.m x perceptron_batch_online.m x perceptron_batch.m x sum_of_error.m x +
1 - clear
2
3 %defining the classes
4 - Class1 = [0.15 0.9 ; 0.1 0.5 ; -0.2 0.4 ; 0.05 0.6];
5 - Class2 = [0.8 0.3 ; 0.55 -0.15 ; 0.4 0 ; 0.68 .45];
6 %part a plotting
7 - plot(Class1(:,1),Class1(:,2),'b*'); hold on; plot(Class2(:,1),Class2(:,2),'gs');
8 %part b
9 - line = perceptron_batch(Class1,Class2,.1,20);
10 - linex = -1:0.1:1;
11 - liney = (-line(1)*linex - line(3))/line(2);
12 - plot(linex,liney,'k');
13 %part c
14 - online = perceptron_batch_online(Class1,Class2,.1,20);
15 - onlinex = -1:0.1:1;
16 - onliney = (-online(1)*onlinex - online(3))/online(2);
17 - plot(onlinex,onliney,'m');
18 %part d
19 - w = sum_of_error(Class1,Class2);
20 - wx = -1:0.1:1;
21 - wy = (-w(1)*wx - w(3))/w(2);
22 - plot(wx,wy,'g');
23
24
25
```

Figure 1 Main code

Part a) plotting Class1 and Class2

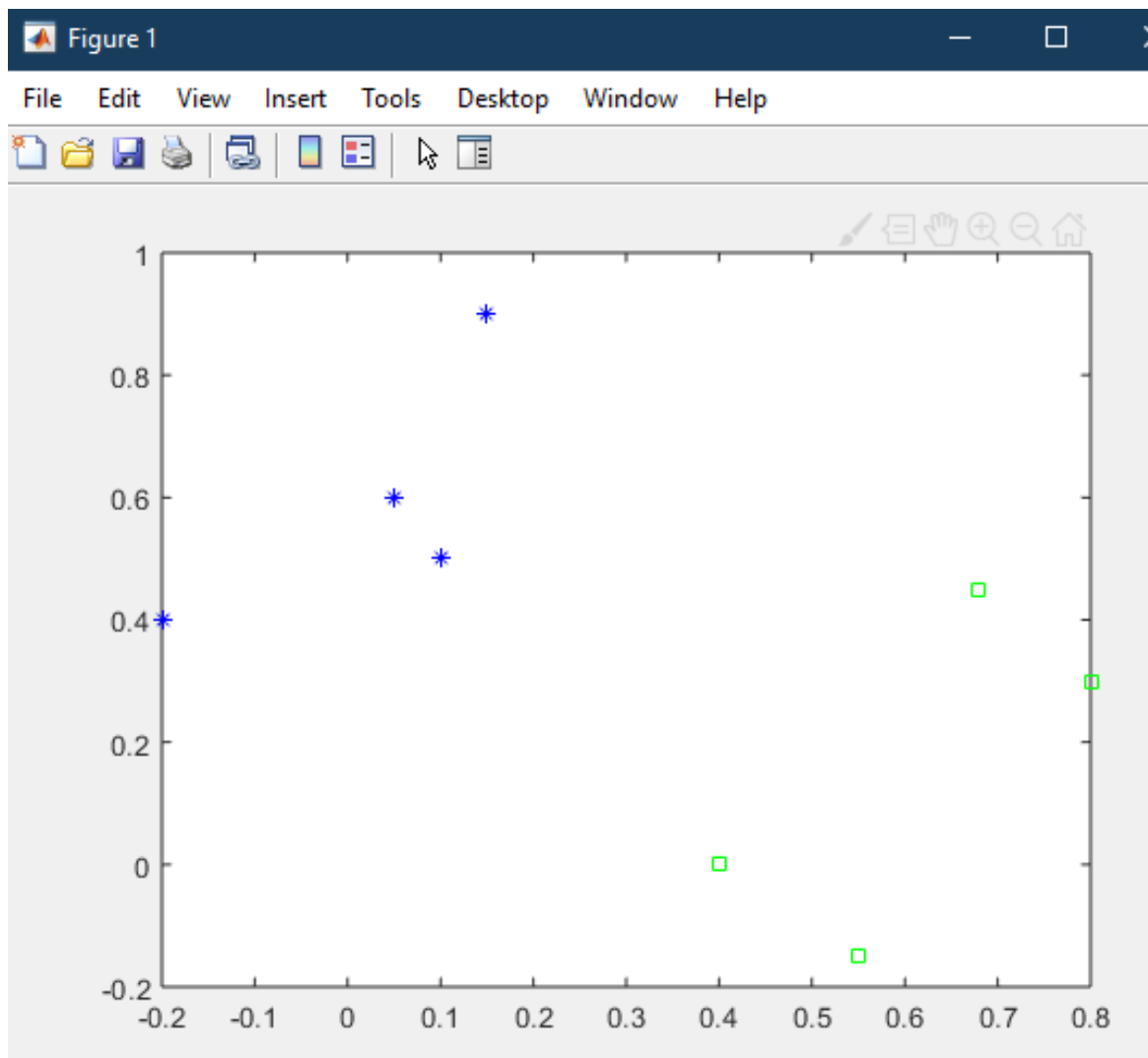


Figure 2 part a

Part b) using the batch version given (black line)

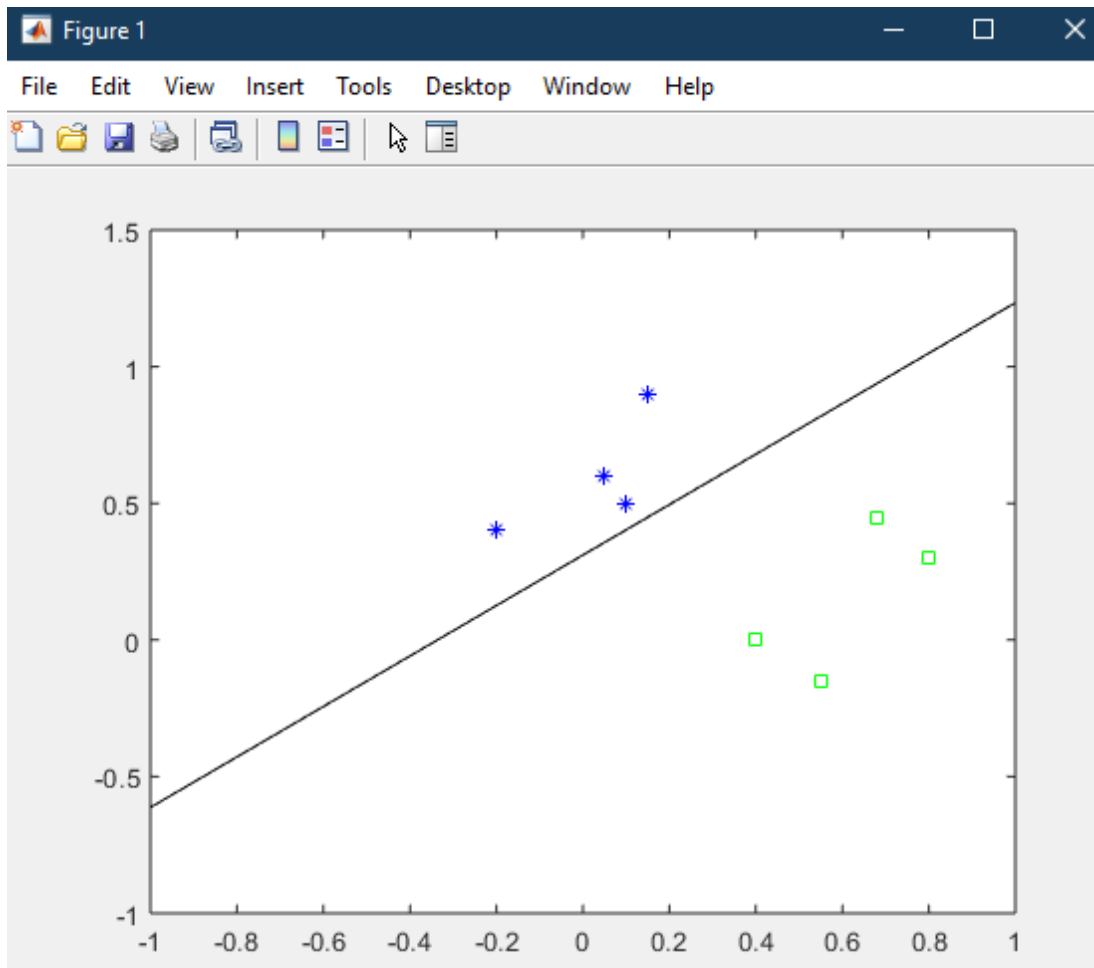


Figure 3 part b

Part c) using the online perceptron algorithm (magenta line)

```

Editor - D:\Pattern Recognition\Matlab files\HW3\perceptron_batch_online.m
HW3.m  perceptron_batch_online.m  perceptron_batch.m  sum_of_error.m  +
1  function W = perceptron_batch_online(X1, X2, ro, maxiter)
2  % function W = perceptron_batch(X1, X2, ro, maxiter) The function accepts %...%
11
12  % get number of samples from each class
13  [m1 n1] = size(X1);
14  [m2 n2] = size(X2);
15
16  % check feature vector dimensions
17  if n1 ~= n2
18      disp('ERROR! Size of feature vectors from both classes must be equal! Exiting...');
19  else
20      n = n1;
21      % initialize W0
22      W = rand(1,n+1);
23      % initialize termination conditions
24      found = 0;
25      t = 0;
26      while (found == 0 && t < maxiter)
27          t = t+1;
28          misclassified = 0;
29          for i=1:m1
30              if W*[X1(i,:), 1]' < 0
31                  misclassified = misclassified + 1;
32                  W = W + (ro * [X1(i,:), 1]);
33              end
34          end
35          for i=1:m2
36              if W*[X2(i,:), 1]' > 0
37                  misclassified = misclassified + 1;
38                  W = W - (ro * [X2(i,:), 1]);
39              end
40          end
41          % apply correction to current weights
42          if misclassified == 0;
43              found = 1;
44          end
45      end
46      if found == 0
47          disp('ERROR! Maximum number of iterations reached without finding')
48          disp('a solution. Returned W is the latest solution achieved.')
49      end
50  end
51

```

Figure 4 part c full code

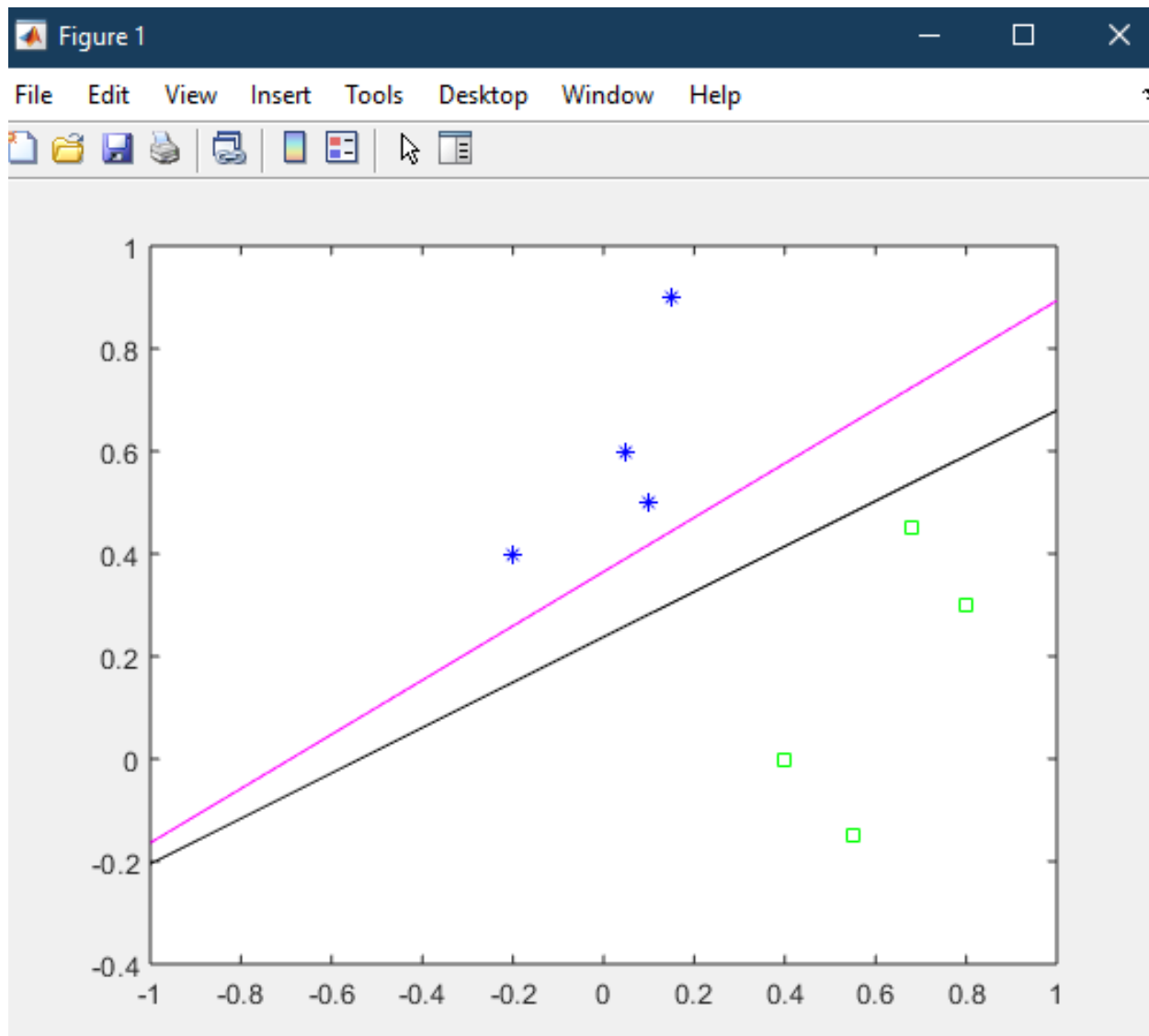


Figure 5 part c plot

Part d) using the sum of errors algorithm (green line)

```

1  function w = sum_of_error(X1, X2)
2  length = size(X1) + size(X2);
3  for i=1:length(2)
4      y(i) = 1;
5  end
6  for i=length(2)+1:length(1)
7      y(i) = -1;
8  end
9  x = [X1;X2];
10 one = (ones(8,1));
11 x = [x one];
12 w = inv(x'*x)*x'*y;
13 end
14

```

Figure 6 part d full code

```

w =

    -2.2209
     1.3275
     0.2046

```

Figure 7 equation of the line that separates the two batches

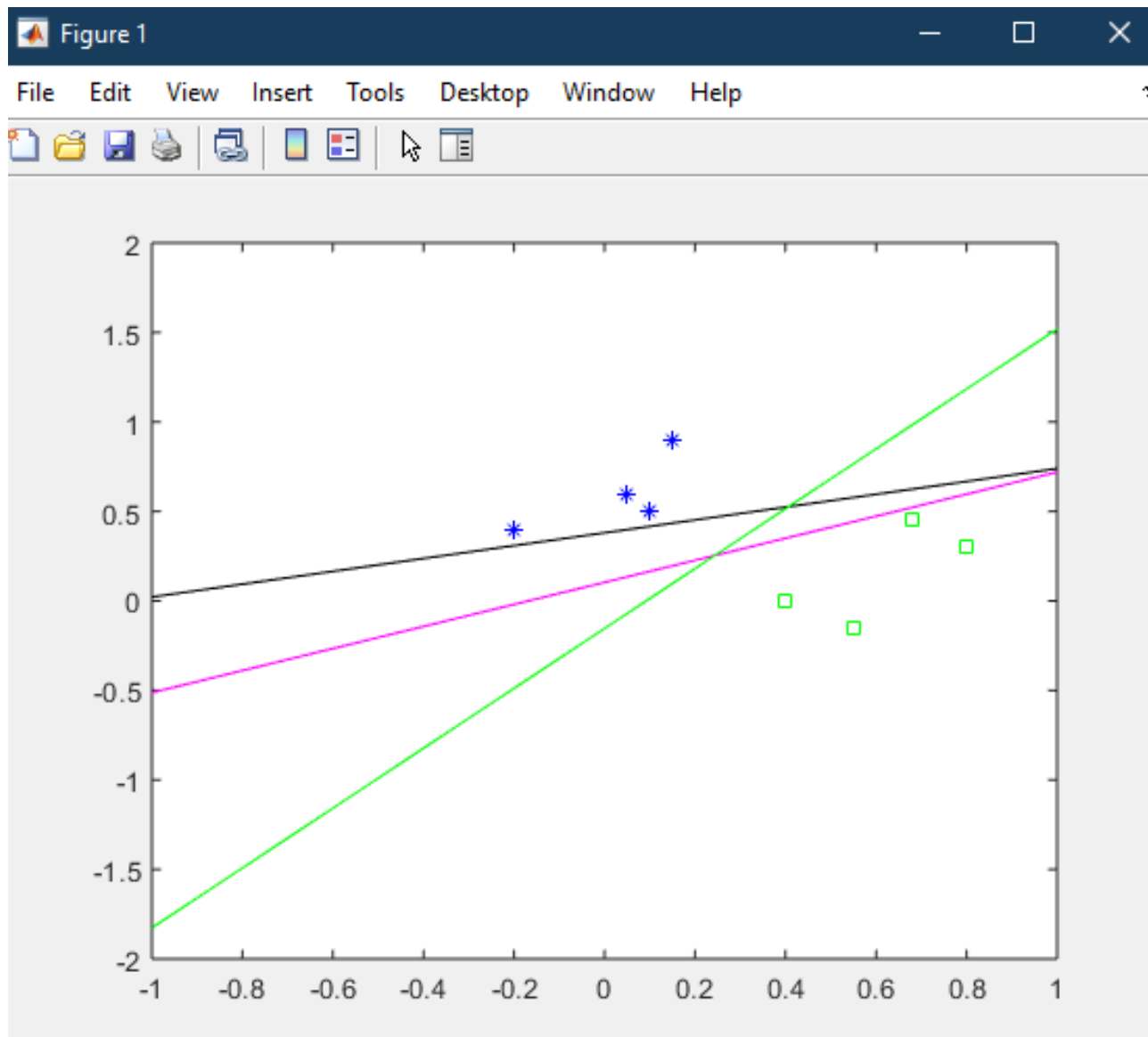


Figure 8 plot of all three different lines to separate batches

All the lines did separate the classes like they were programmed to do; however, the one best solution is the sum of errors line. This line is the best one because it created a line that shot straight down the middle of both batches giving our future features a greater chance to be correct.