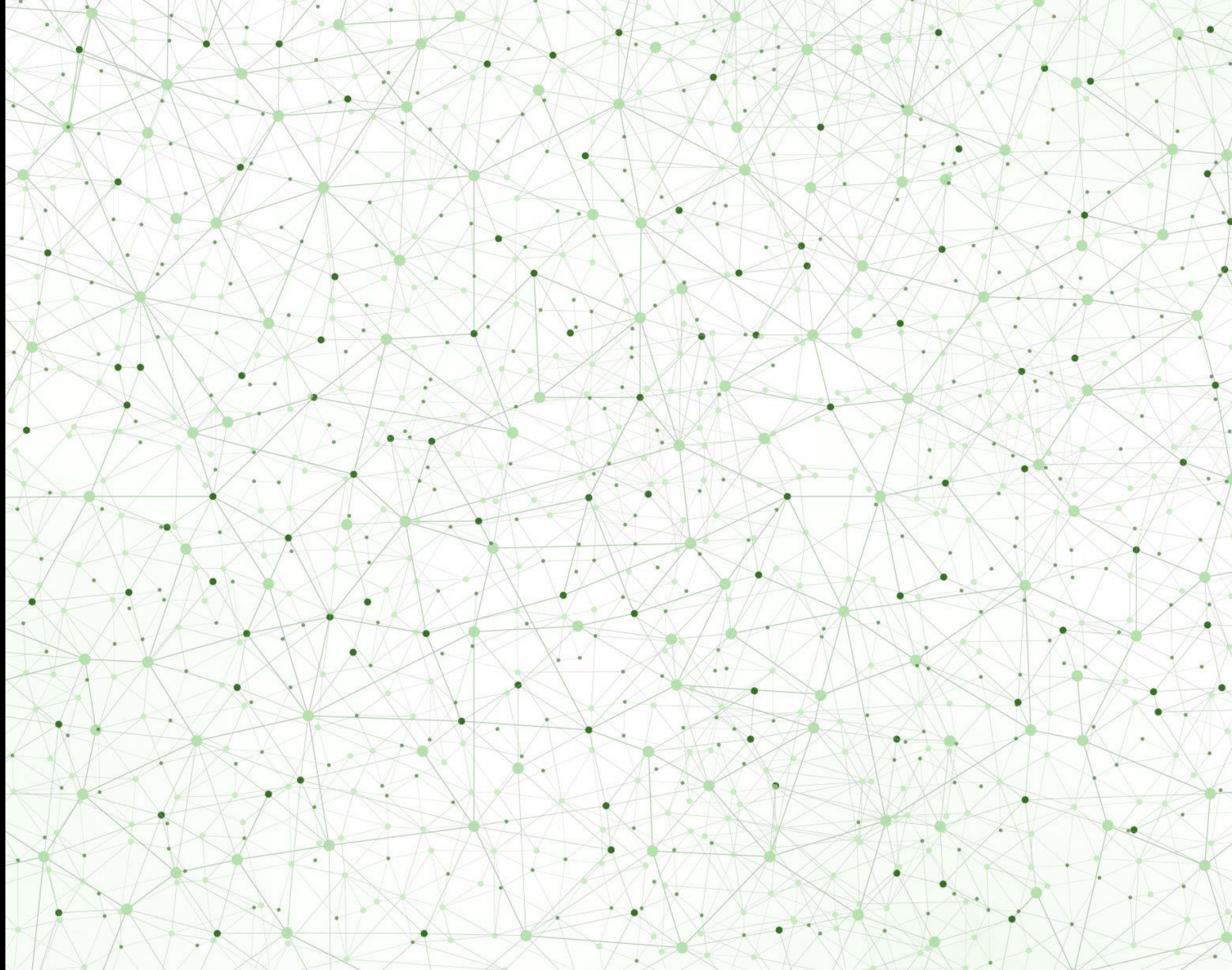


ML2: Perceptron

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AP 186



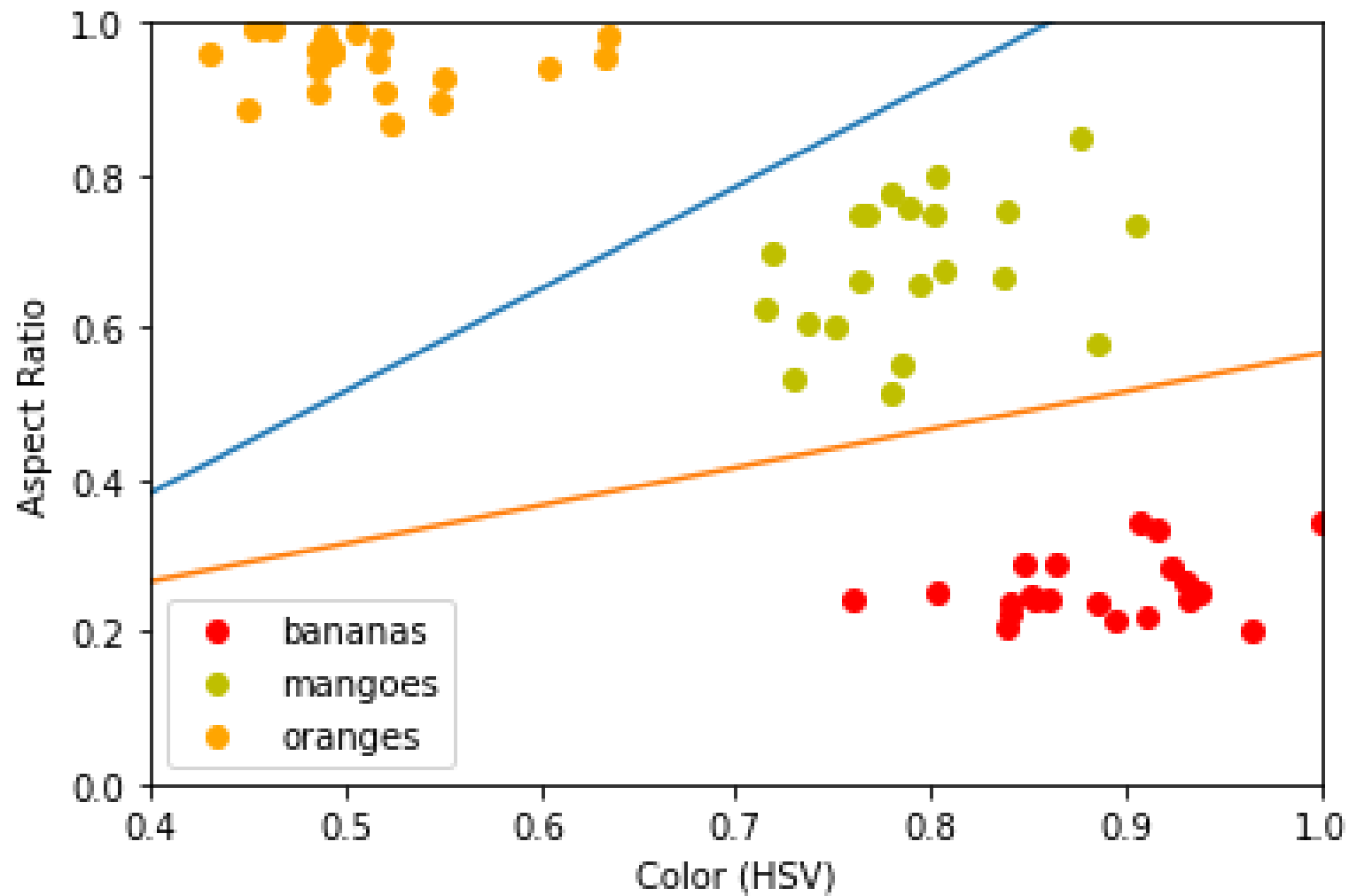
Objectives:

The main objective of this activity is to compute the decision surface between the two classes using the **perceptron algorithm**.

```
def function(a):  
    if a >= 0:  
        return 1  
    else:  
        return -1  
  
for i in range(100):  
    for j in range(42):  
        if j < 21:  
            d = 1  
        else:  
            d = -1  
        a = weights1[0] + x[j]*weights1[1] + y[j]*weights1[2]  
        z = function(a)  
  
        weights1[0] += rate*(d-z)  
        weights1[1] += rate*(d-z)*x[j]  
        weights1[2] += rate*(d-z)*y[j]
```

Results:
Perceptron
Algorithm

Figure 2: Simple Perceptron Algorithm in Python



Blue decision line (mango vs orange):

$$y = 1.3451233019364752x + 0.1571489100747912$$

Orange decision line (mango vs bananas):

$$y = 0.5001057568800451x + 0.06555828479948116$$

Figure 2: Decision lines for the three fruit classes

Discussion

The decision lines for each pair of classes were successfully obtained and plotted as shown in Figure 2. Blue decision line separates the mangoes from the oranges while the orange decision line separates the mangoes from the bananas.

The **initial weights** that I used for both decisions lines are **$w_0=0.1$, $w_1=0.2$, $w_2=-0.2$** .

The **final weights** for orange vs mango decision line are

$w_0=0.3854543025001047$, $w_1=-3.299313777472603$, and $w_2=2.4527965374793697$.

The **final weights** for mango vs banana decision line are

$w_0=-0.26170127453084074$, $w_1=-1.9963657434911168$, and $w_2=3.991887147921721$.

Testing the perceptron

Mango vs Orange

$W_0 = -0.3854543025001047$

$W_1 = -3.299313777472603$

$W_2 = 2.4527965374793697$

If $x_1 = 0.48$ and $x_2 = 0.9$ (features of an orange), the feature point is equal to

0.23873 (above the blue decision line!)

Mango vs Banana

$W_0 = -0.26170127453084074$

$W_1 = -1.9963657434911168$

$W_2 = 3.991887147921721$

If $x_1 = 0.84$ and $x_2 = 0.2$ (features of a banana), the feature point is equal to

-1.14 (below the orange decision line!)

Analysis

Upon testing the perceptron more, the algorithm accurately classified the fruits from each other.

To produce a decision line/separation that is somewhat more equidistant from the different clusters, we can increase the iterations so that $(d-z)$ is really small.

Reflection

Rating 10/10

I was able to finish this activity successfully! I am quite sure with my algorithm since I've checked it for numerous values of x_1 and x_2 . This activity was quite intimidating but it was actually pretty easy.

References

- Soriano, M. (2020). ML2 – Perceptron.
- Python Machine Learning Tutorial. [Machine Learning with Python: Separating Classes with Dividing Lines \(python-course.eu\)](https://python-course.eu/machine-learning-with-python-separating-classes-with-dividing-lines)