



**Nombre:** Mauricio Josiel Villalobos Rodriguez

**Materia:** Inteligencia artificial 2

**Profesor:** CARLOS ALBERTO VILLASENOR PADILLA

**Fecha de entrega:** 28/11/2023

**Actividad:** A04 Perceptrón multicapa

## Red neuronal multicapa (MLP, Multilayer Perceptron)

```
1 class MLP:
2     def __init__(self, layers_dims,
3                   hidden_activation=tanh,
4                   output_activation=logistic):
5         # Attributes
6         self.L = len(layers_dims) - 1
7         self.w = [None] * (self.L + 1)
8         self.b = [None] * (self.L + 1)
9         self.f = [None] * (self.L + 1)
10
11     for l in range(1, self.L+1):
12         self.w[l] = -1 + 2 * np.random.rand(layers_dims[l],
13                                              layers_dims[l-1])
14         self.b[l] = -1 + 2 * np.random.rand(layers_dims[l], 1)
15         if l == self.L:
16             self.f[l] = output_activation
17         else:
18             self.f[l] = hidden_activation
19
20     def predict(self, X):
21         A = X
22         for l in range(1, self.L + 1):
23             Z = self.w[l] @ A + self.b[l]
24             A = self.f[l](Z)
25         return A
26
27     def fit(self, X, Y, epochs=500, lr=0.1):
28         p = X.shape[1]
29         for _ in range(epochs):
30             # Initialize containers
31             A = [None] * (self.L + 1)
32             dA = [None] * (self.L + 1)
33             lg = [None] * (self.L + 1)
34
35             # Propagation -----
36             A[0] = X
37             for l in range(1, self.L + 1):
38                 Z = self.w[l] @ A[l-1] + self.b[l]
39                 A[l], dA[l] = self.f[l](Z, derivative=True)
40
41             # Backpropagation -----
42             for l in range(self.L, 0, -1):
43                 if l == self.L:
44                     lg[l] = (Y - A[l]) * dA[l]
45                 else:
46                     lg[l] = (self.w[l+1].T @ lg[l+1]) * dA[l]
47
48             # Weight and bias update
49             for l in range(1, self.L + 1):
50                 self.w[l] += (lr/p) * (lg[l] @ A[l-1].T)
51                 self.b[l] += (lr/p) * np.sum(lg[l])
```

Draw function

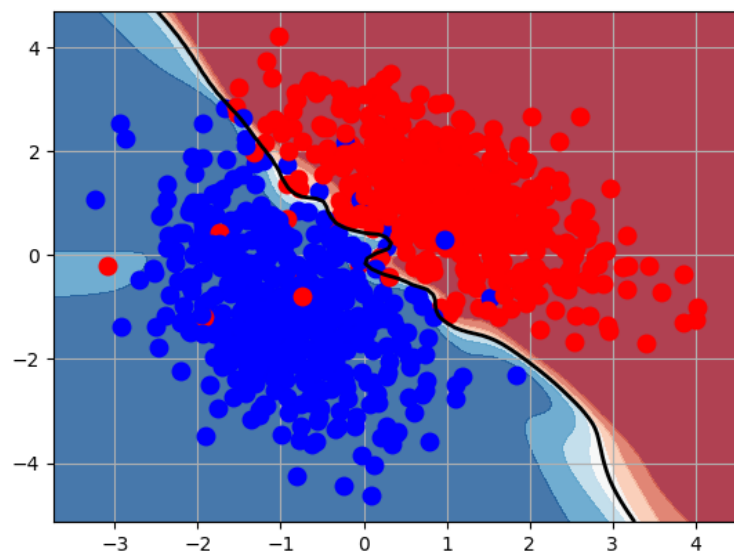
```
1 def MLP_binary_class_2d(X, Y, net):
2     plt.figure()
3     for i in range(X.shape[1]):
4         if Y[0, i] == 0:
5             plt.plot(X[0, i], X[1, i], 'ro', markersize=9)
6         else:
7             plt.plot(X[0, i], X[1, i], 'bo', markersize=9)
8     xmin, ymin = np.min(X[0, :]) - 0.5, np.min(X[1, :]) - 0.5
9     xmax, ymax = np.max(X[0, :]) + 0.5, np.max(X[1, :]) + 0.5
10    xx, yy = np.meshgrid(np.linspace(xmin, xmax, 100),
11                          np.linspace(ymin, ymax, 100))
12    data = [xx.ravel(), yy.ravel()]
13
14    zz = net.predict(data)
15    zz = zz.reshape(xx.shape)
16    plt.contour(xx,yy,zz, [0.5], colors='k', linestyle='--', linewidths=2)
17    plt.contourf(xx,yy,zz, alpha=0.8, cmap=plt.cm.RdBu)
18    plt.xlim([xmin, xmax])
19    plt.ylim([ymin, ymax])
20    plt.grid()
21    plt.show()
```

Antes de poder procesar nuestro set tenemos que preprocesarlo para poder meterlo a nuestro modelo

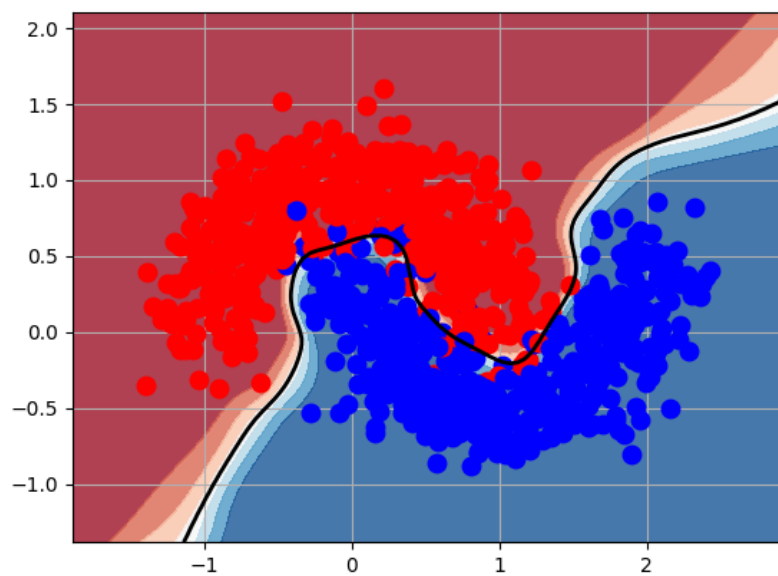
Blobs example:

```
1 dataset = pd.read_csv('blobs.csv', header = None)
2
3 nombres_columnas = ['x1', 'x2', 'y'] # Sustituye con tus nombres reales
4
5 # Asigna los nombres de las columnas al dataset
6 dataset.columns = nombres_columnas
7
8 print(dataset.info())
9 X = np.stack((dataset['x1'],dataset['x2']), axis=0)
10 Y = np.array([dataset['y']], dtype=int)
11
12 net = MLP((2,100,50,30,1))
13 net.fit(X, Y)
14 print(net.predict(X))
15 MLP_binary_class_2d(X, Y, net)
```

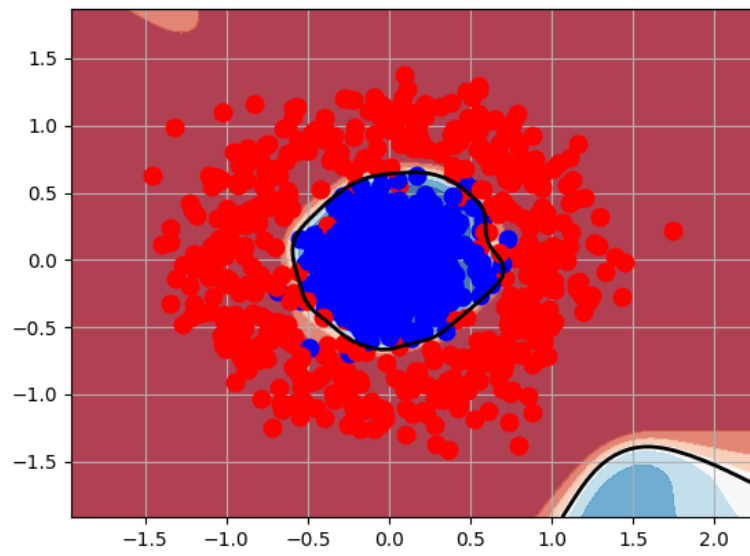
Blobs



Moons



circles



XOR

