

Assignment10

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1. Machine learning method work well because of human-designed representation and input features. And ML just optimizes weights to best make a final prediction.

Deep learning is a learning subfield of learning representations of data, which is effective at learning patterns. DL algorithms learn representation by using a hierarchy or multiple layers. It doesn't need people to design a method to extract features.

DL is useful because of the flow properties:

- (a) Manually designed features are often over-specified, incomplete and take a long time to design and validate.
- (b) Learned Features are easy to adapt, fast to learn.
- (c) Deep learning provides an almost very flexible, universal, learnable framework for representing world, visual and linguistic information.
- (d) Can learn both unsupervised and supervised.
- (e) Effective end-to-end joint system learning.
- (f) Utilize large amounts of training data.

2.

- (a) $y = \sigma(w_1x_1 + w_2x_2 + w_3x_3 + b)$, in which σ is sigmoid function, therefore,

$$y = \frac{1}{1 + e^{-(w_1x_1 + w_2x_2 + w_3x_3 + b)}}$$

$$(b) x_{11} = \sigma(2 * (-1) + 1 * 1 + 2) = \sigma(1) = \frac{1}{1 + e^{-1}} = 0.73$$

$$x_{12} = \sigma(2 * 1 + (-1) * (-1) - 4) = \sigma(-1) = \frac{1}{1 + e^1} = 0.27$$

$$y = \sigma(1 * 0.73 + 1 * 0.27 + 1) = \sigma(2) = \frac{1}{1 + e^{-2}} = 0.88$$

3.

- (a) During training:

$$x_{11} = \text{ReLU}((-1) * 1 + 0 * 2 + 0 * 1) = \sigma(-1) = 0$$

$$x_{12} = \text{ReLU}(2 * 1 + 1 * 2 + 0 * 1) = \text{ReLU}(4) = 4$$

$$x_{14} = \text{ReLU}(0 * 1 + 0 * 2 + 1 * 1) = \text{ReLU}(1) = 1$$

$$y_1 = \text{ReLU}((-1) * 0 + 2 * 4 + (-4) * 1) = \text{ReLU}(4) = 4$$

$$y_2 = \text{ReLU}(1 * 0 + 0 * 4 + (-2) * 1) = 0$$

(b) During testing:

$$x_{11} = \text{ReLU}(0.75 * ((-1) * 1 + 2.5 * 2 + 0 * 2 + 0 * 1)) = \text{ReLU}(3) = 3$$

$$x_{12} = \text{ReLU}(0.75 * (2 * 1 + 0 * 2 + 1 * 2 + 0 * 1)) = \text{ReLU}(3) = 3$$

$$x_{13} = \text{ReLU}(0.75 * (3 * 1 + (-1) * 2 + 0 * 2 + (-2) * 1)) = \text{ReLU}(-0.75) = 0$$

$$x_{14} = \text{ReLU}(0.75 * (0 * 1 + 0 * 2 + 0 * 2 + 1 * 1)) = \text{ReLU}(0.75) = 0.75$$

$$y_1 = \text{ReLU}(0.75 * ((-1) * 3 + 2 * 3 + 0 * 0 + (-4) * 0.75)) = \text{ReLU}(0) = 0$$

$$y_2 = \text{ReLU}(0.75 * (1 * 3 + 0 * 3 + (-1) * 0 + (-2) * 0.75)) = \text{ReLU}(1.125) = 1.125$$

4.

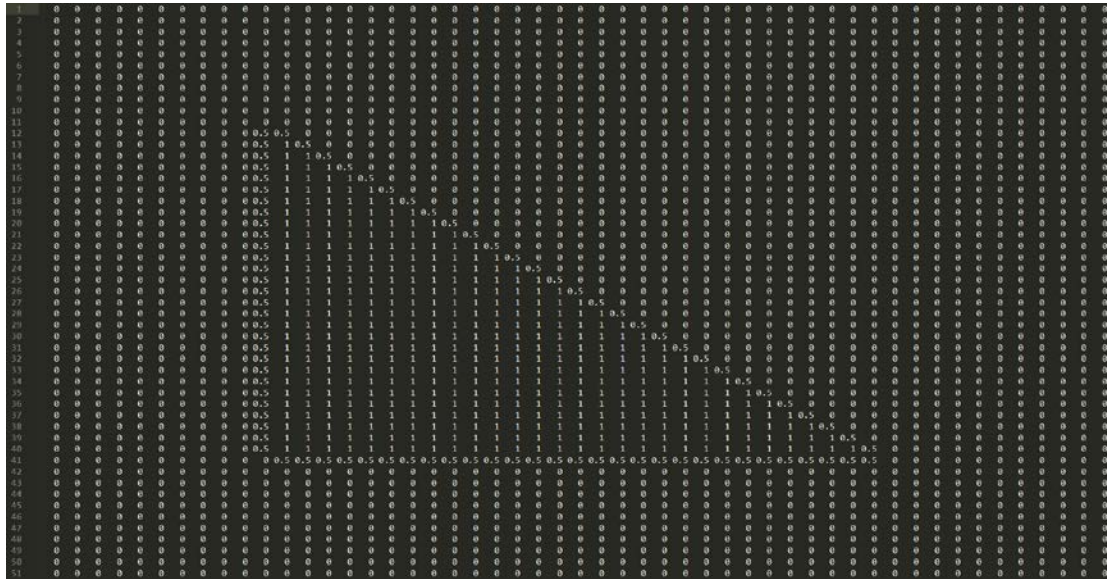
(a) I calculate the values by a simple program, which is as following:

```
1  #include <iostream>
2  #include <cmath>
3  #include <fstream>
4  using namespace std;
5
6  const double a[]={10.,0.,-10.};
7  const double b[]={0.,10.,-10.};
8  const double e=2.7182818;
9  const double c[]={40.,40.,40.};
10
11 double sigmoid(double input)
12 {
13     double output=1./(1+pow(e,-input));
14     if(output<0.001) return 0;
15     if(output>0.999) return 1;
16     return output;
17 }
18
19
20 int main() {
21     fstream result;
22     result.open("result.txt");
23     if(!result.is_open())
24     {
25         cerr<<"something wrong";
26         exit(0);
27     }
28     for(int i=0;i<3;i++)
29     {
30         double x1,x2;
31         cin>>x1>>x2;
32         double y1,y2,y3;
33         y1=a[0]*x1+b[0]*x2;
34         y2=a[1]*x1+b[1]*x2;
35         y3=a[2]*x1+b[2]*x2+300;
36         y1=sigmoid(y1);
37         y2=sigmoid(y2);
38         y3=sigmoid(y3);
39         double output=sigmoid(c[0]*y1+c[1]*y2+c[2]*y3-100);
40         result<<x1<<' '<<x2<<': '<<output<<endl;
41     }
42     result.close();
43     return 0;
44 }
```

And the results is:

output(A)=1; output(B)=0; output(C)=0.

(b) Just as (a) below, I use a simple program to calculate the values of output that input is $[-10,40] \times [-10,40]$. The result is as following:



So the decision boundary can be displayed as:

$$\begin{cases} x_1 = 0 \\ x_2 = 0 \\ x_1 + x_2 = 30 \end{cases}$$