## 1 Supervised learning

Method where you train the program by feeding the learning algorithm with a mapping of inputs to correct outputs.

### 1.1 Regression

Regression is curve fitting: learn a continuous input  $\rightarrow$  output mapping from a set of examples.

#### 1.2 Classification

Outputs are discrete variables (category labels). Learn a decision boundary that separates one class from the other. Generally, a confidence is also desired, i.e., how sure are we that the input belongs to the chosen category.

#### 1.3 Training set

The training set is a set of m(X, y) pairs, where:

 $X \in \mathbb{R}^d$  models the input.  $y \in \{0,1\}$  models the output.

#### 1.4 Error function

The error function for a model  $f: X \mapsto y$  parameterized by W applied to a dataset  $\{(X, y)\}$  of size m is:

$$\min_{W} \sum_{i=1}^{m} \left( f_{W}(X_{i}) - y_{i} \right)^{2}$$

### 1.5 Perceptron

Perceptron is the trivial neural network. The model for a parameter  $W = (\text{threshold}, w_1, \ldots, w_d)$  and inputs of the form  $(1, x_1, \ldots, x_d)$  is given by

$$f_W(X) = \operatorname{sign}(W^\top X)$$

If  $x_i$  is evidence for approval, then  $w_i$  should be high.

If  $x_i$  is evidence for denial, then  $w_i$  should be low.

#### 1.5.1 Learning algorithm

The learning algorithm of the Perceptron is quite simple. For a training set  $S = \{(X_1, y_1), (X_1, y_1), \ldots\}$ 

- Show each sample in sequence repetitively.
- If the output is correct, do nothing.
- If the produced output is negative, and the correct output is positive, increase/decrease the weights whose inputs are positive/negative.
- If the produced output is positive, and the correct output is negative, decrease/increase the weights whose inputs are positive/negative.

# 2 Reinforcement learning

Method where you train the program by rewarding the learning algorithm positively or negatively according to the produced results. This method is similar to how we teach animals.

## 3 Unsupervised learning

Given only inputs as training, find a pattern: discover clusters, manifolds, embedding.