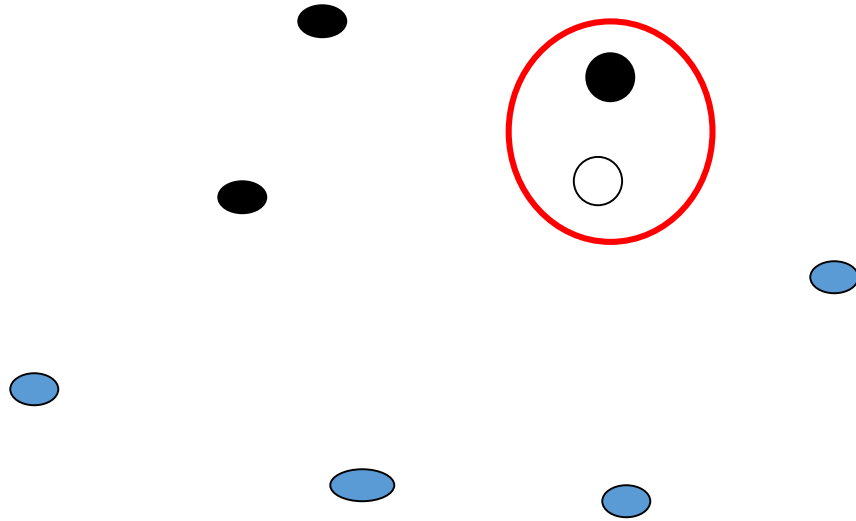


Machine Learning Course

Vahid Reza Khazaie

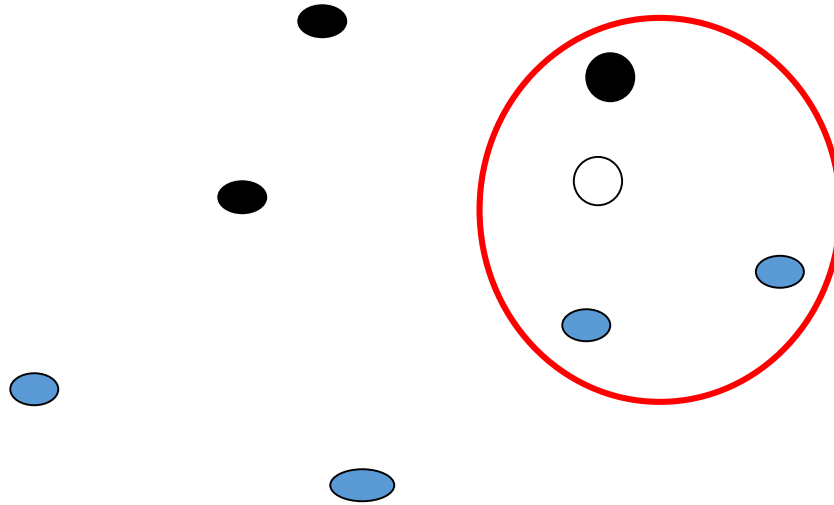
1-NN

- 1-Nearest Neighbor



3-NN

- 3-Nearest Neighbor

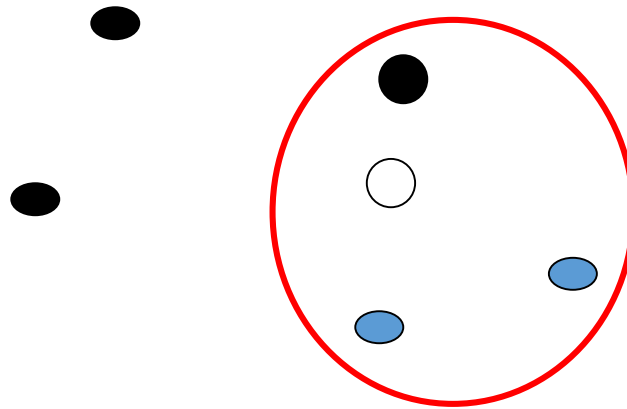


k-NN

- ▶ In pattern recognition, the k-nearest neighbors algorithm (k-NN) is a non-parametric method used for classification and regression. In both cases, the input consists of the k closest training examples in the feature space. The output depends on whether k-NN is used for classification or regression:
- ▶ In k-NN classification, the output is a class membership. An object is classified by a plurality vote of its neighbors, with the object being assigned to the class most common among its k nearest neighbors (k is a positive integer, typically small). If $k = 1$, then the object is simply assigned to the class of that single nearest neighbor.
- ▶ In k-NN regression, the output is the property value for the object. This value is the average of the values of k nearest neighbors.

3-NN

- 3-Nearest Neighbor

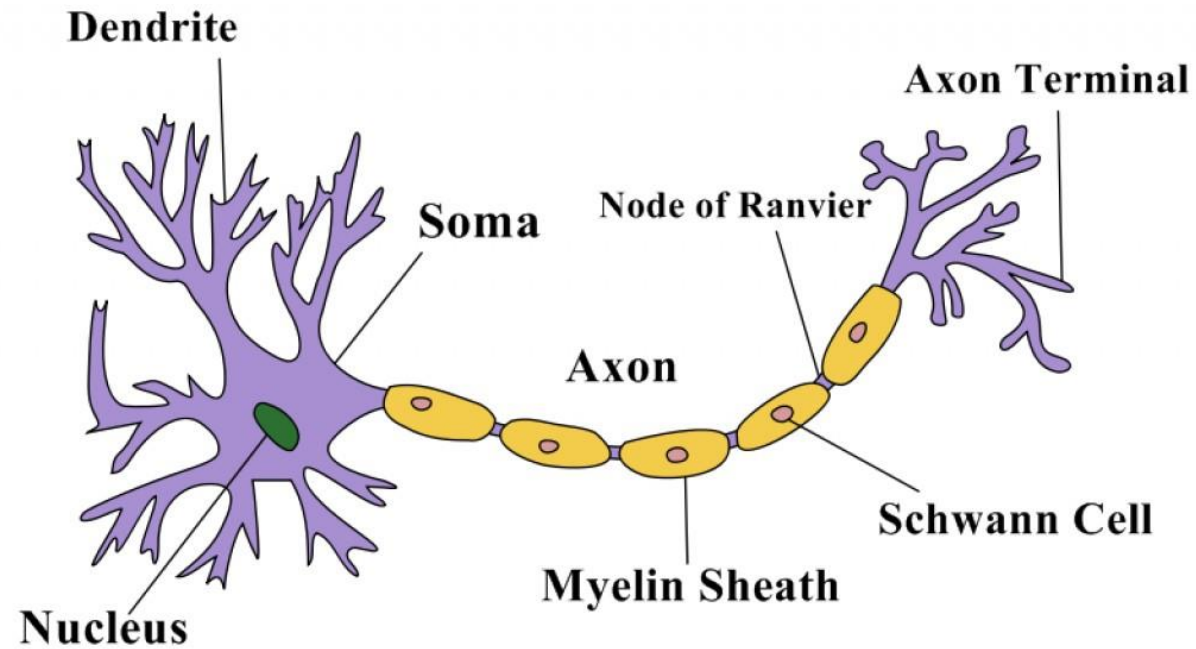


Introduction to the Perceptron

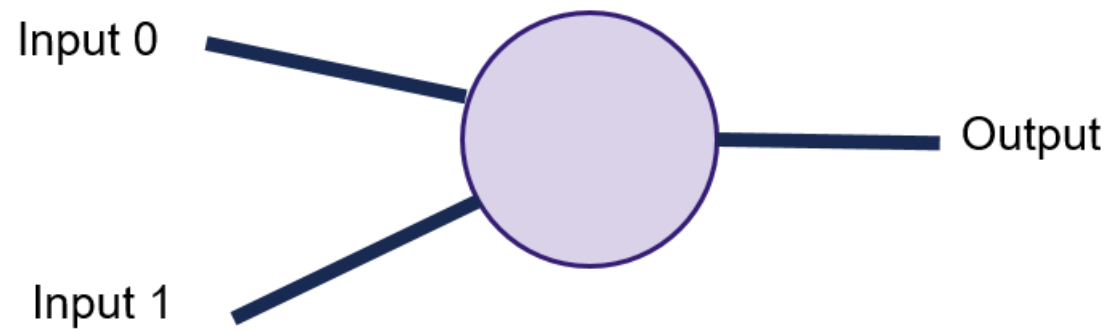
- ▶ In machine learning, the **perceptron** is an algorithm for supervised learning of binary classifiers. A binary classifier is a function which can decide whether or not an input, represented by a vector of numbers, belongs to some specific class.

Introduction to the Perceptron

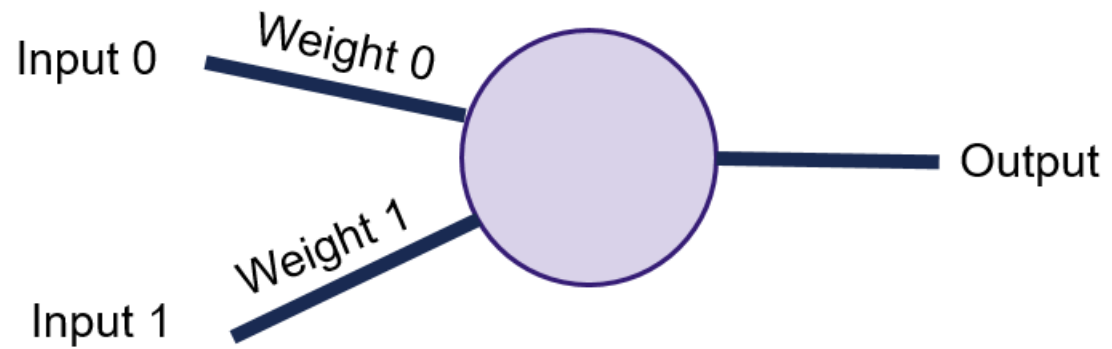
► Biological Neuron



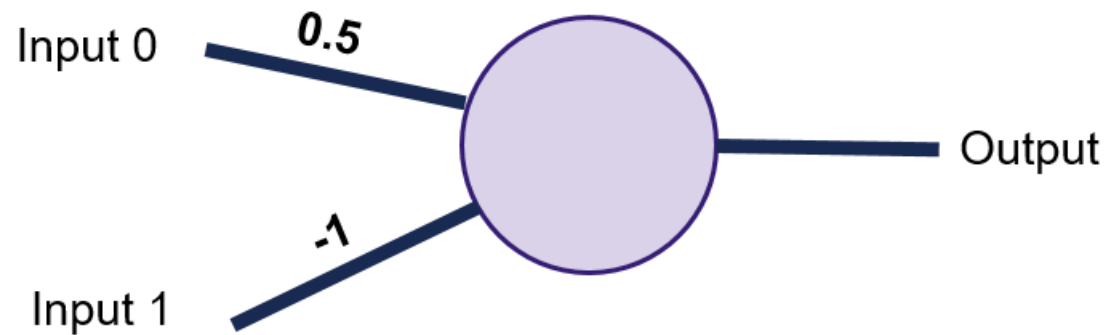
Perceptron



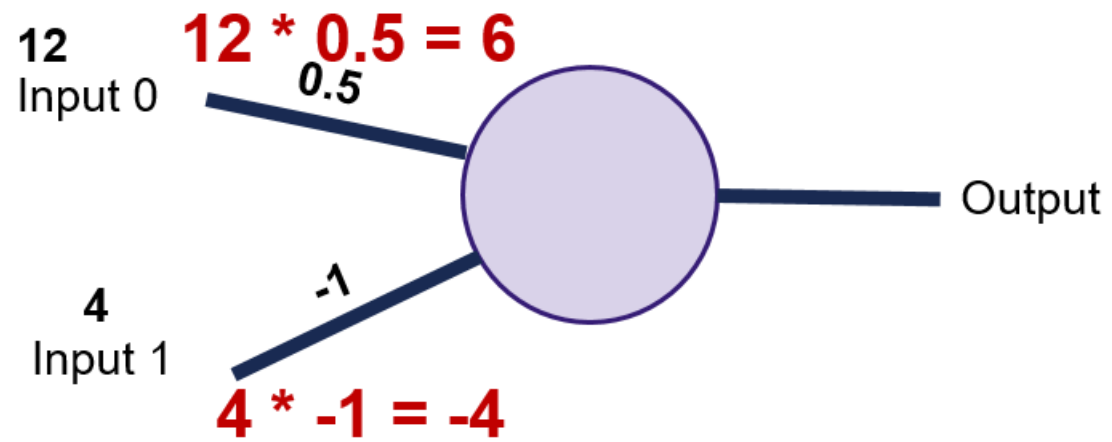
Perceptron



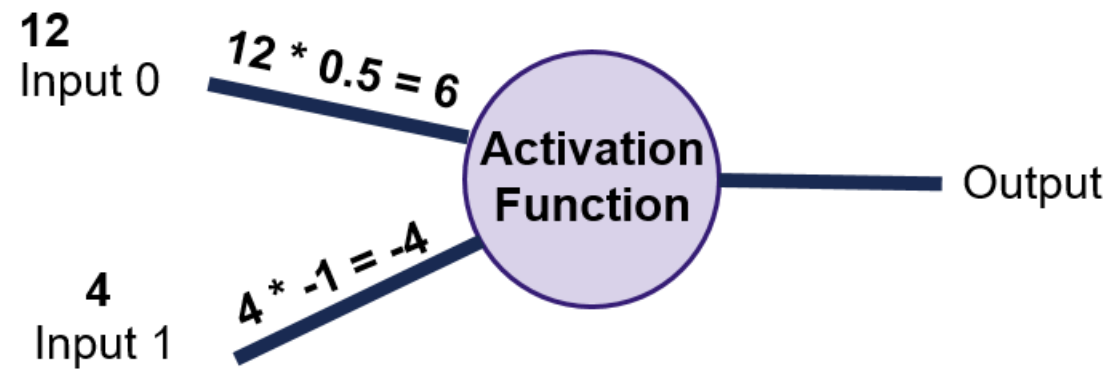
Perceptron



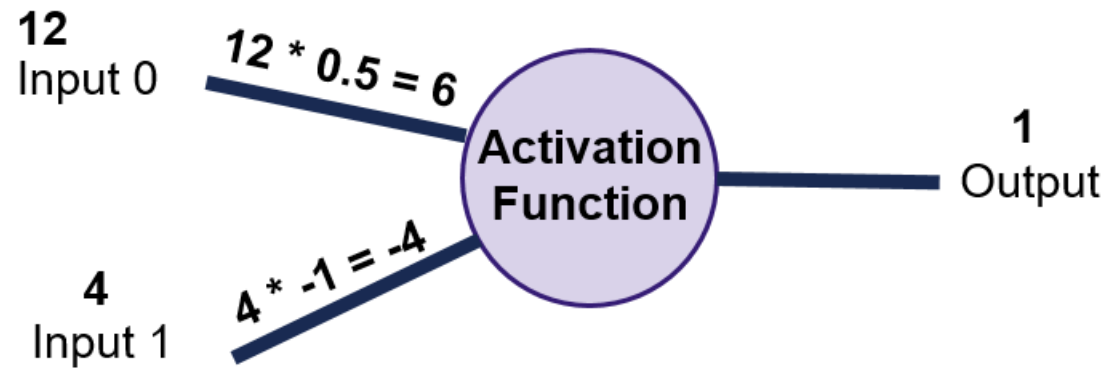
Perceptron



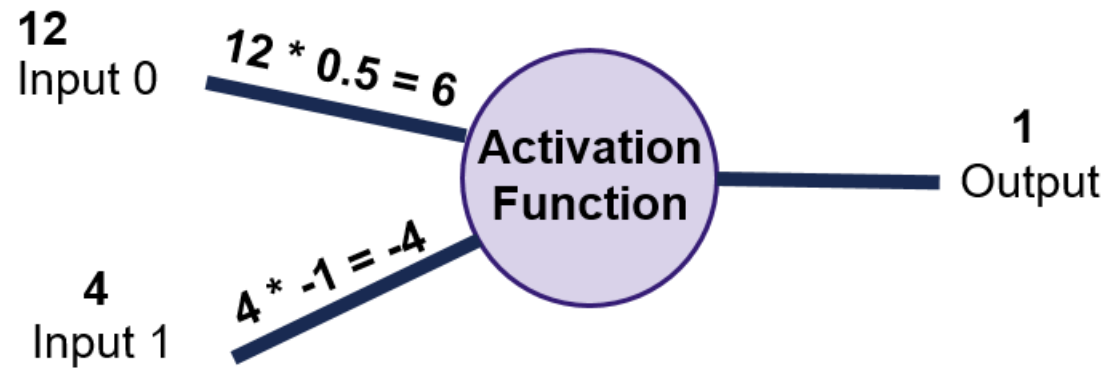
Perceptron



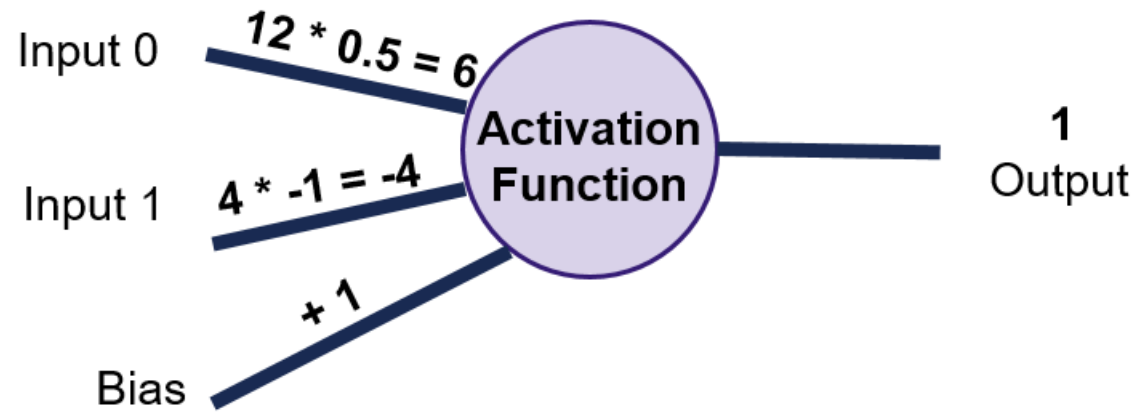
Perceptron



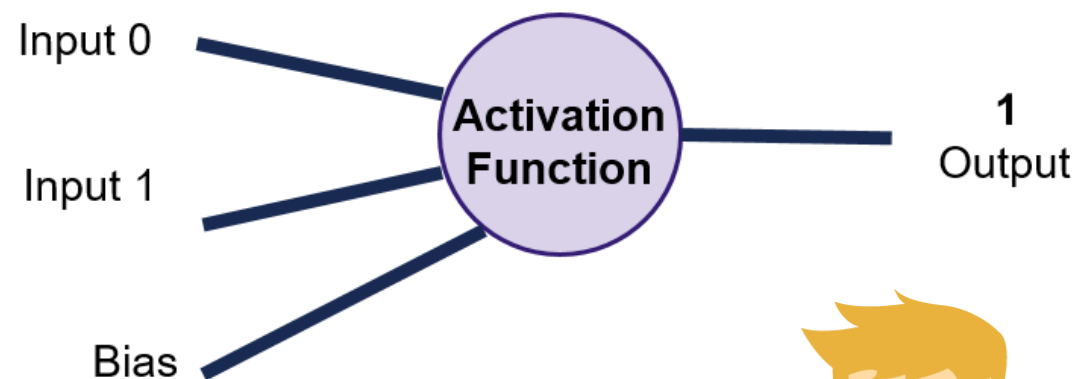
Perceptron



Perceptron



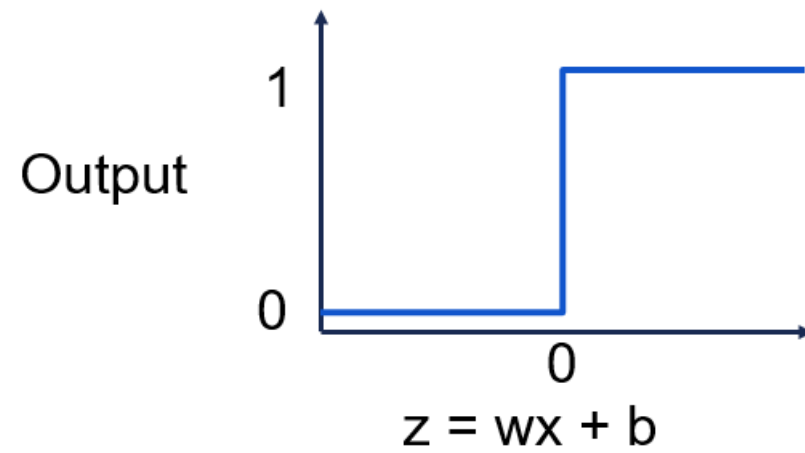
Perceptron



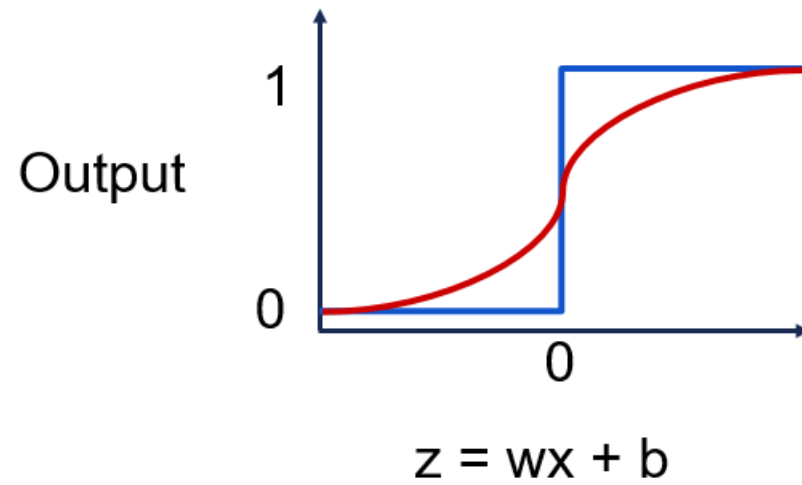
$$\sum_{i=0}^n w_i x_i + b$$



Activation Function

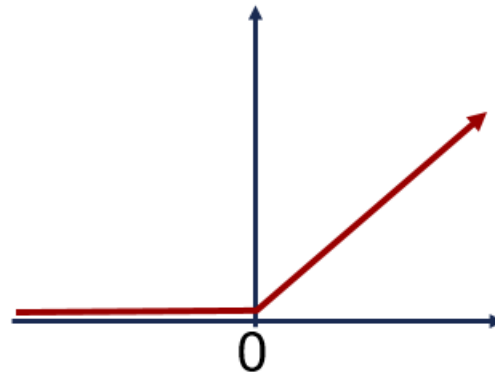


Activation Function



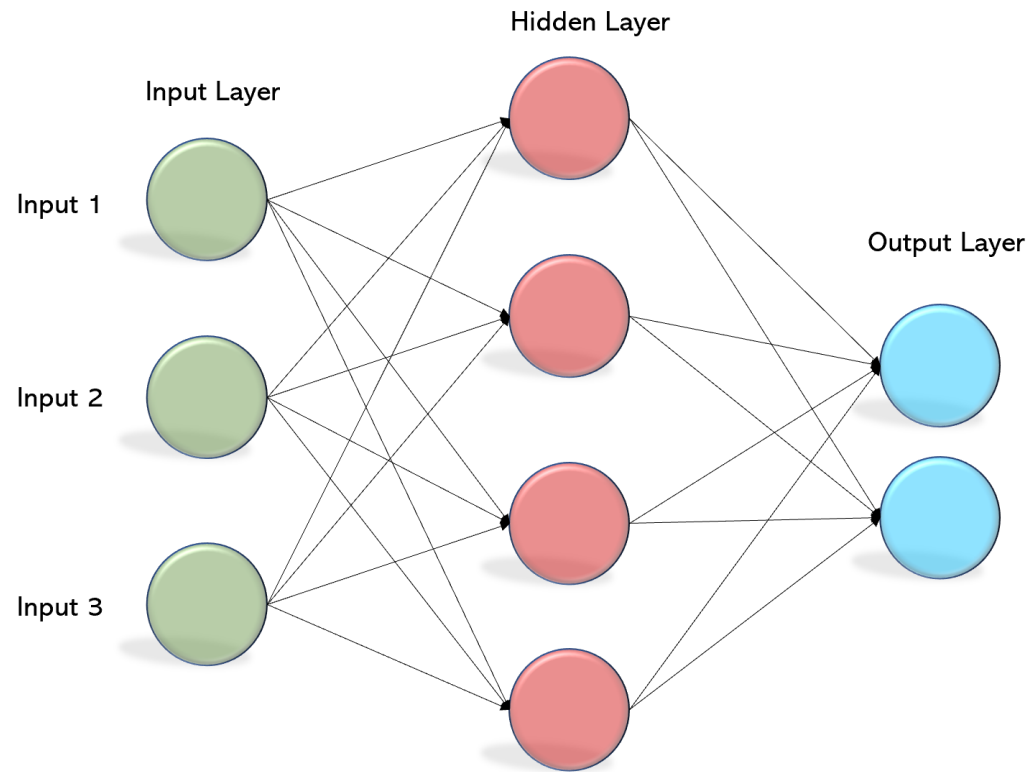
Activation Function

Output



$$z = wx + b$$

Introduction to Neural Networks



Cost Functions

▶ $C = \sum (y - \hat{y})^2 / n$

▶ $C = (-1/n) \sum (y \cdot \ln(\hat{y}) + (1 - y) \cdot \ln(1 - \hat{y}))$

Learning

- ▶ Neurons
- ▶ Cost Function
- ▶ Gradient Decent and Backpropagation

Learning

► Gradient Decent

Repeat until convergence {

$$\theta_j \leftarrow \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta)$$

}

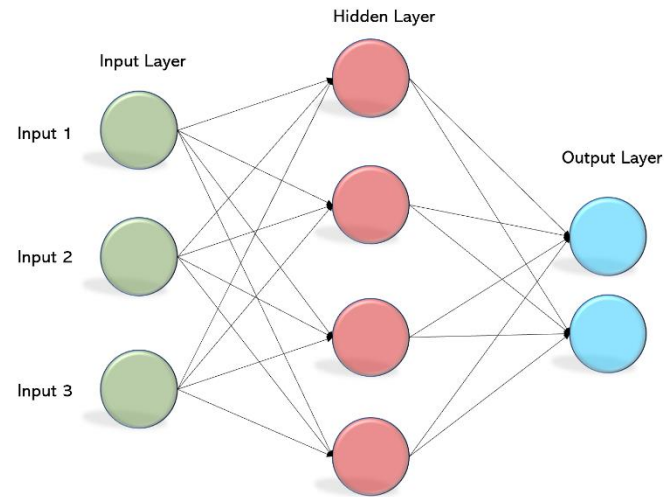
Learning

► Backpropagation

Repeat until convergence {

$$\theta_j \leftarrow \theta_j - \alpha \frac{\partial}{\partial \theta_j} J(\theta)$$

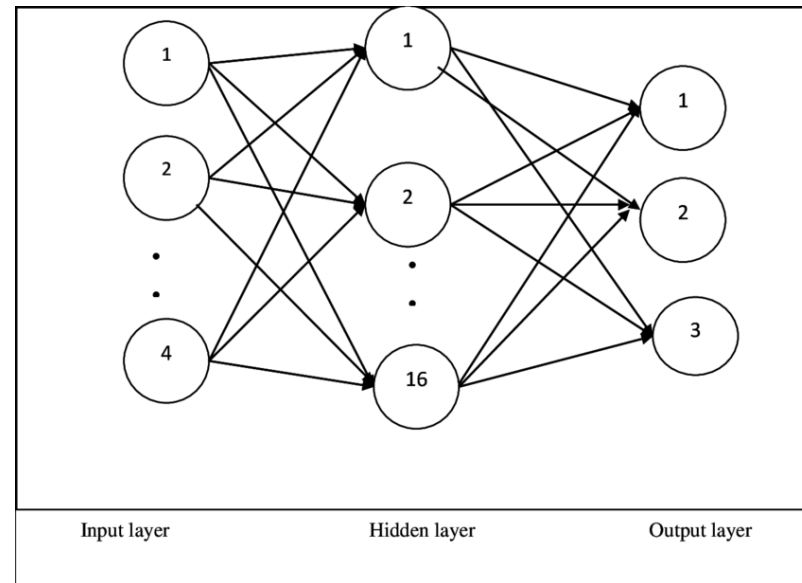
}



Learning

► Softmax

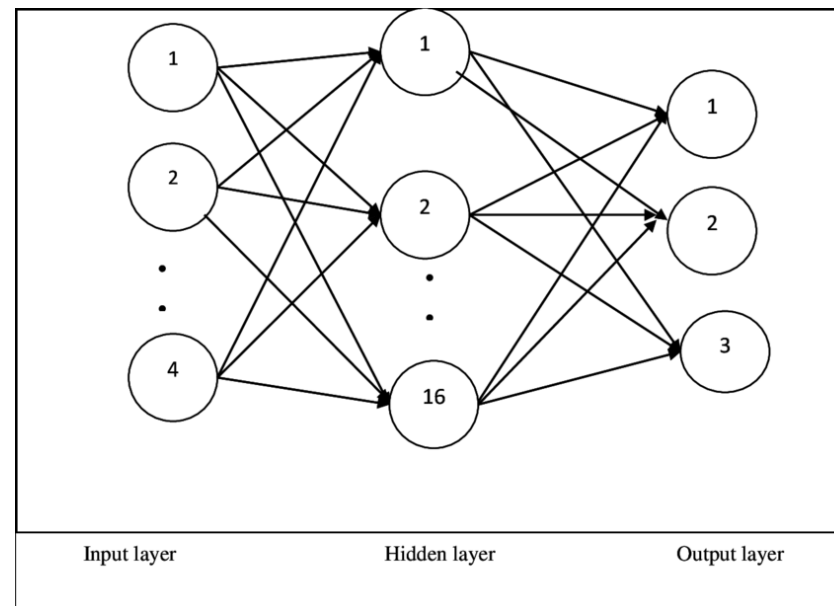
$$\sigma(\mathbf{z})_i = \frac{e^{z_i}}{\sum_{j=1}^K e^{z_j}} \text{ for } i = 1, \dots, K \text{ and } \mathbf{z} = (z_1, \dots, z_K) \in \mathbb{R}^K$$



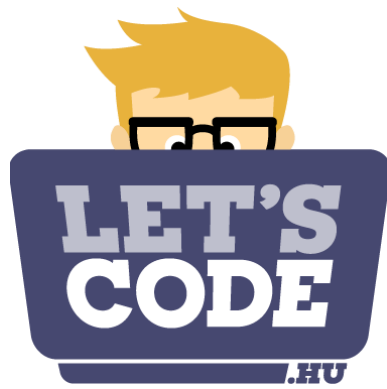
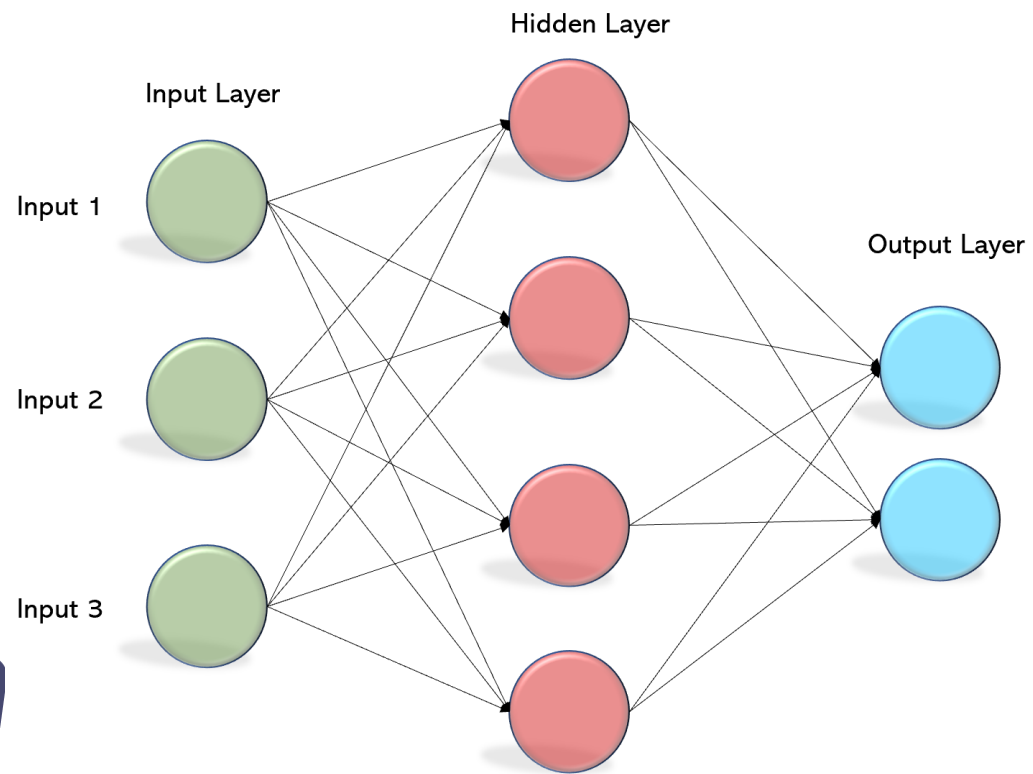
MLP

► Visualization

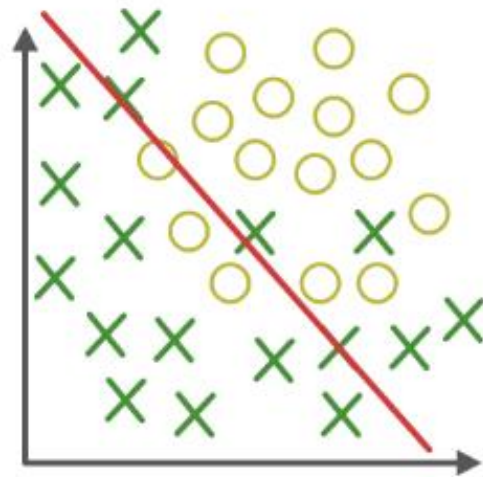
playground.tensorflow.org



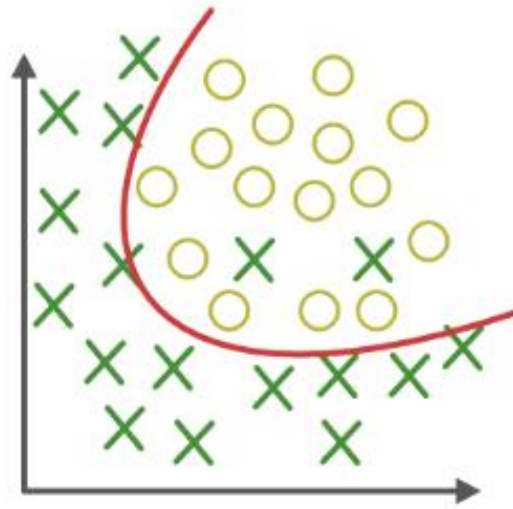
MLP



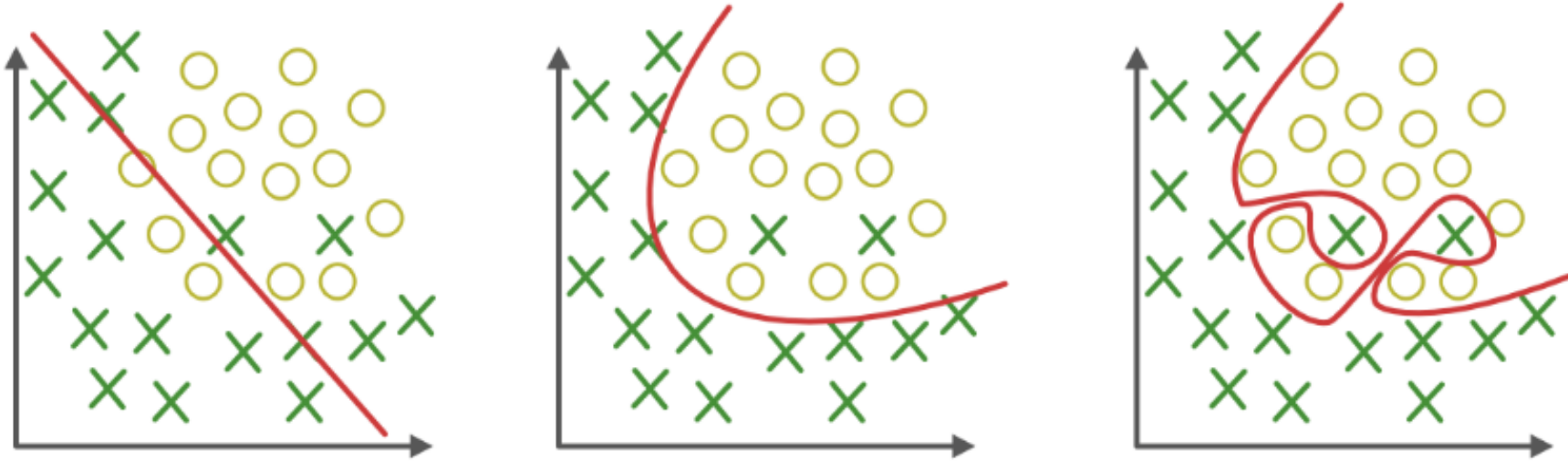
Overfitting, Underfitting



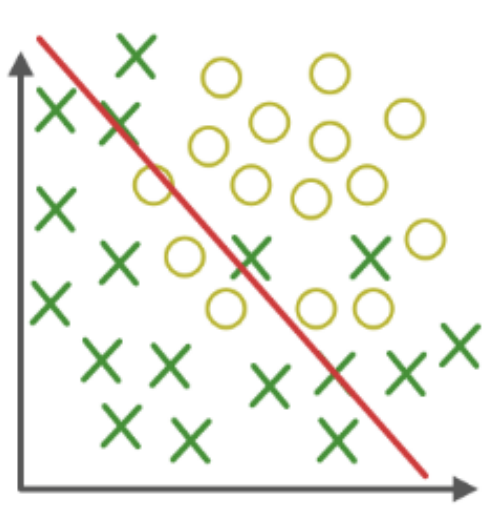
Overfitting, Underfitting



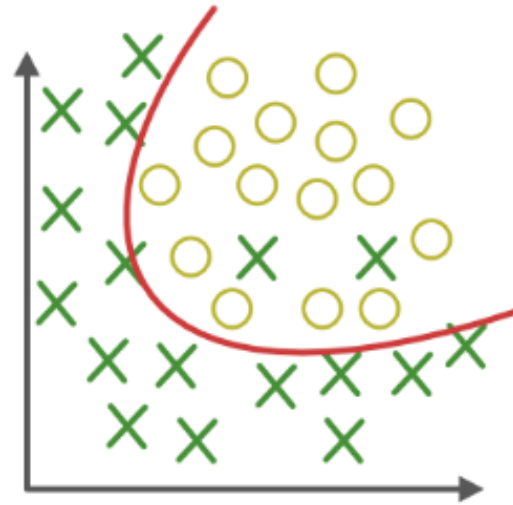
Overfitting, Underfitting



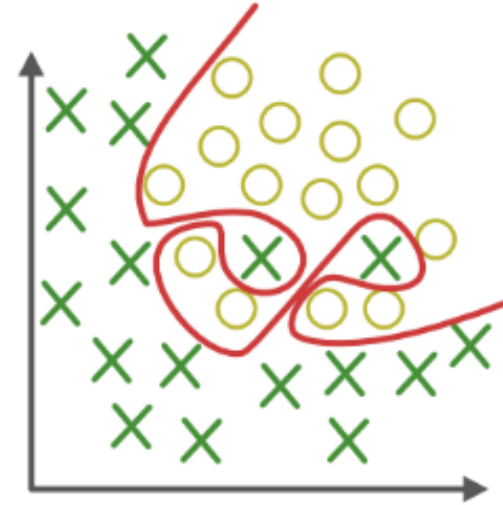
Overfitting, Underfitting



Under-fitting

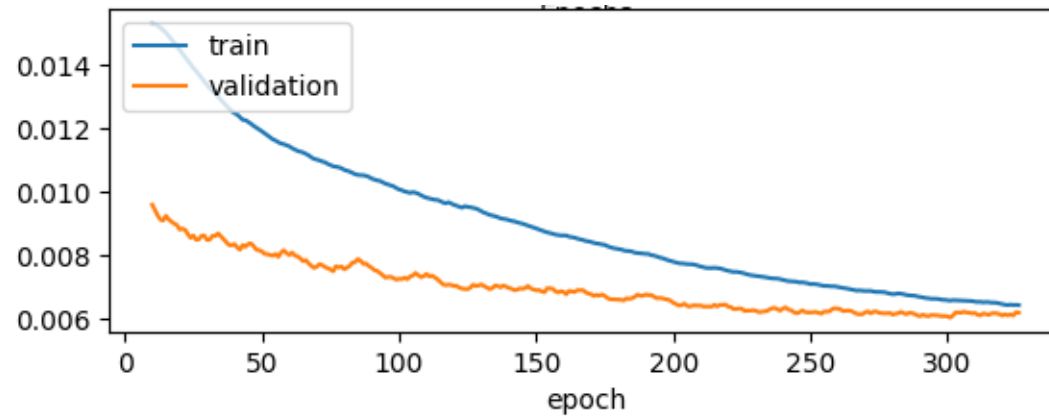


Appropriate-fitting

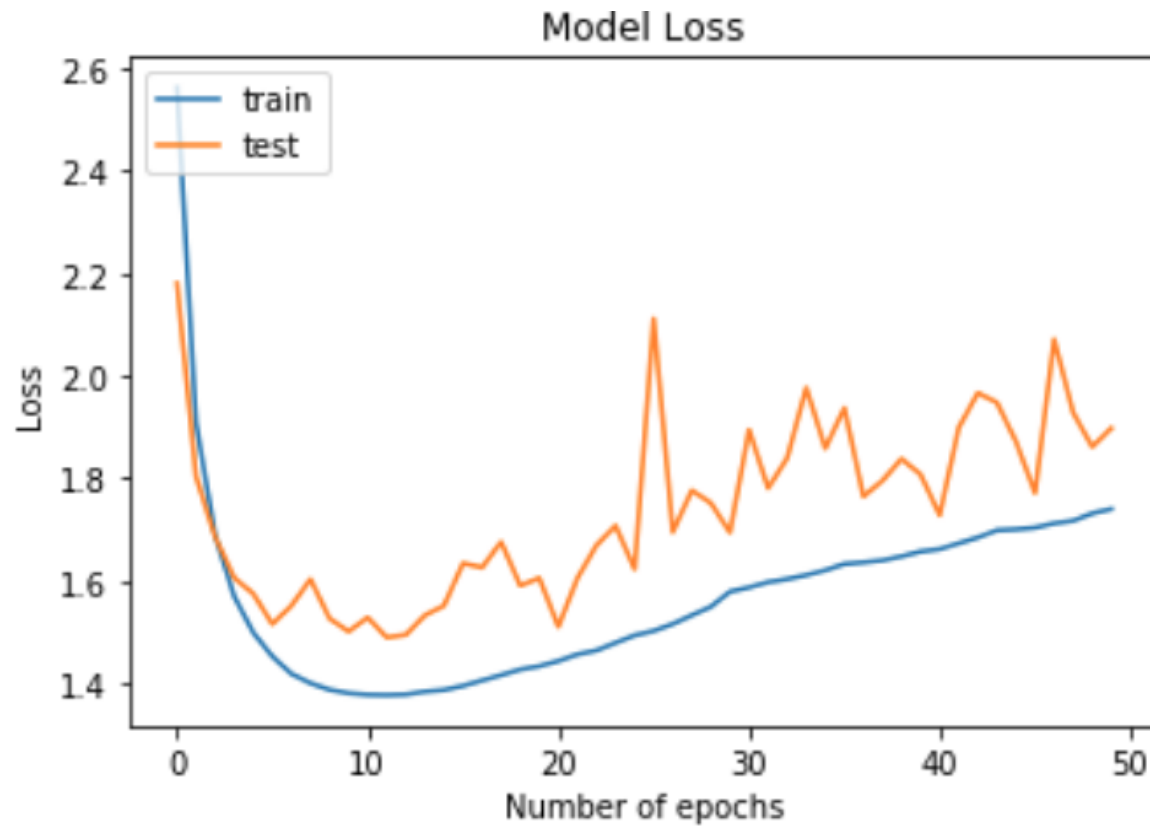


Over-fitting

Overfitting, Underfitting



Overfitting, Underfitting

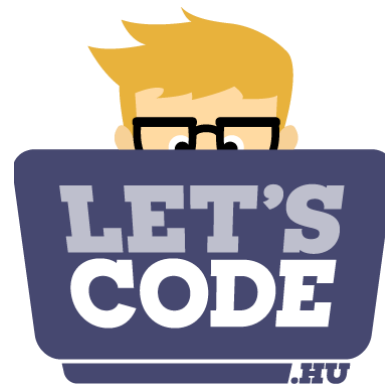


Avoid Overfitting

- ▶ Weight Regularization
- ▶ Dropout
- ▶ More Data
- ▶ Cross-validation

Performance Evaluation with Resampling

- ▶ Train/Validation/Test Sets
- ▶ K-fold Cross Validation
- ▶ Leave-one-out Cross Validation
- ▶ Bootstrapping



[13 resampling.ipynb](#)