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# Machine Learning Formulas\*

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May 28, 2017

## Abstract

[TODO ]

## 1 Introduction

[TODO ]

## 2 Supervised Learning

[TODO ]

**Error Measures** [TODO ]

- **Error Rate:** [TODO ]
- **Resubstitution error:** [TODO ]
- **Generalization Error:** [TODO ]

**Experimental Strategies** [TODO ]

- **Holdout:** [TODO ]
- **Repeated Holdout:** [TODO ]
- **Cross Validation:** [TODO ]
- **Repeated Cross Validation:** [TODO ]

### 2.1 Decision Trees

[TODO ]

#### 2.1.1 Information Theory

[TODO ]

#### 2.1.2 ID3

[TODO ]

#### 2.1.3 C4.5

[TODO ]

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\*This document is being maintained in <https://github.com/garciparedes/machine-learning-formulas>

## **2.2 Rule Based Systems**

[TODO ]

### **2.2.1 1R**

[TODO ]

### **2.2.2 PRISM**

[TODO ]

### **2.2.3 IREP**

[TODO ]

### **2.2.4 RIPPER**

[TODO ]

### **2.2.5 PART**

[TODO ]

## **2.3 Instance Based Learning**

[TODO ]

### **2.3.1 K - Nearest Neighbors**

[TODO ]

### **2.3.2 IB3**

[TODO ]

## **2.4 Bayes Learning**

[TODO ]

### **2.4.1 Naive Bayes**

[TODO ]

### **2.4.2 K2**

[TODO ]

### **2.4.3 TAN**

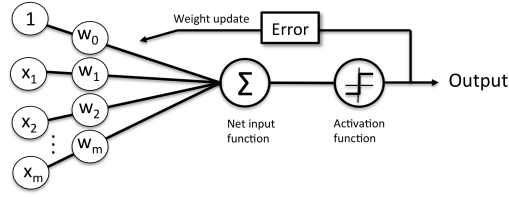
[TODO ]

## **2.5 Linear Classifiers**

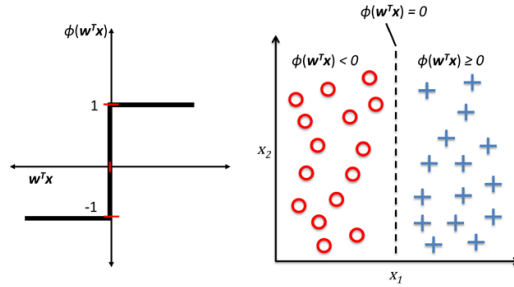
[TODO ]

### **2.5.1 Linear Regression**

[TODO ]



**Figure 1:** General concept of perceptron



**Figure 2:** Activation function

## 2.5.2 Logistic Regression

[TODO ]

## 2.5.3 Support Vector Machines

[TODO ]

## 2.6 Neural Networks

### 2.6.1 Perceptron model

Perceptron's structures

$$w = \begin{bmatrix} w_1 \\ \vdots \\ w_m \end{bmatrix}, x = \begin{bmatrix} x_1 \\ \vdots \\ x_m \end{bmatrix} \quad (1)$$

Output equation

$$z = w_1 x_1 + \cdots + w_m x_m = \mathbf{w}^T \mathbf{x} \quad (2)$$

Activation function

$$\phi(z) = \begin{cases} 1 & \text{if } z \geq \theta \\ -1 & \text{otherwise} \end{cases} \quad (3)$$

Update of weight vector

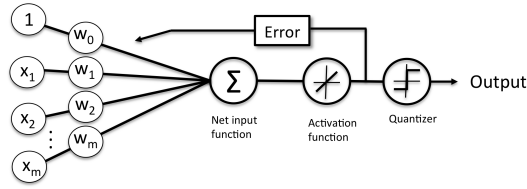
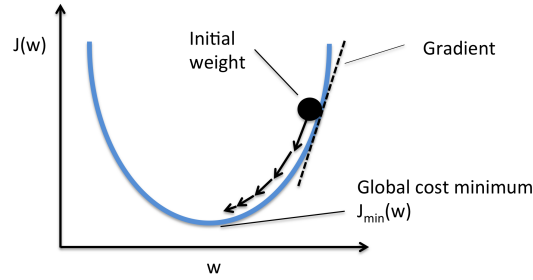
$$w_j := w_j + \Delta w_j \quad (4)$$

$$\Delta w_j = \eta(y^{(i)} - \hat{y}^{(i)})x_j^{(i)} \quad (5)$$

### 2.6.2 Adaptive linear neurons (ADALINE)

Cost function

$$J(w) = \frac{1}{2} \sum_i (y^{(i)} - \phi(z^{(i)}))^2 \quad (6)$$

**Figure 3:** *General concept of adaline perceptron***Figure 4:** *Gradient descent***Update of weight vector**

$$w_j := w_j + \Delta w_j \quad (7)$$

$$\Delta w_j = -\eta \nabla J(w) = \eta \sum_i (y^{(i)} - \phi(z^{(i)})) x_j^{(i)} \quad (8)$$

**Features standarization**

$$x'_j = \frac{x_j - \mu_j}{\sigma_j} \quad (9)$$

**2.6.3 Single Layer Neural Networks****Simple Perceptron** [TODO ]**ADALINE** [TODO ]**2.6.4 Multi Layer Perceptron**

[TODO ]

**2.6.5 Radial Basis Functions**

[TODO ]

**2.6.6 Convolutional Neural Networks**

[TODO ]

**2.6.7 Recurrent Neural Networks**

[TODO ]

**3 Unsupervised Learning**

[TODO ]

## References

- [CCAG17] Teodoro Calonge Cano and Carlos Javier Alonso González. Técnicas de Aprendizaje Automático, 2016/17.
- [Ols16] Dr. Randal S. Olson. Python Machine Learning, 2016.