

# Hands On Machine Learning Notes

Notes based on the Book “Hands-On Machine Learning with Scikit-Learn & TensorFlow” by Aurelien Geron

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## Contents

1. Types of Machine Learning systems .....	1
1.1. (Un)Supervised Learning .....	1
1.2. Semisupervised Learning .....	2
1.3. Reinforcement Learning .....	2
1.4. Batch Learning .....	2

## 1. Types of Machine Learning systems

There are many different Types of ML systems that it is often useful to classify them in broad categories based on :

- Whether or not they are trained with human supervision (supervised, unsupervised, reinforcement learning)
- Whether or not they can learn incrementally on the fly (online vs batch learning)
- Whether they work by simply comparing new data points to old data points, or instead detect patterns in training data

### 1.1. (Un)Supervised Learning

- Machine learning systems can be classified according to the amount and type of supervision they get during training. There are four major categories: **supervised learning, unsupervised learning, semi-supervised learning, and reinforcement learning**

#### Definition 1.1.1

In **supervised learning**, the training data you feed to the algorithm includes the desired solution, called **labels**

A typical supervised learning task is **classification**. The spam filter is an example of this, where models are filled with lots of emails with their class(label) of spam/normal emails.

Another task can be predict a target numeric value, such as the price of a car, known as *predictors*. This task is often called **regression**.

- One example of regression is actually **logistic regression**, which is used for classification. Basically, it outputs a certain probability than an object belongs to a class.

#### Example 1.1.1

Important Supervised Learning Algorithms:

- k-Nearest Neighbors
- Linear Regression
- Logistic Regression
- Support Vector Machines (SVM)
- Decision Trees and Random Forests
- Neural Networks (*can also be unsupervised*)

## Definition 1.1.2

In **unsupervised learning** the training data is unlabeled. In this case, the system tries to learn without having a teacher.

## Example 1.1.2

## Important Unsupervised Learning Algorithms

- Clustering
  - K Means
  - DBSCAN
  - Hierarchical Cluster Analysis (HCA)
- Anomaly detection and novelty detection
  - One class SVM
  - Isolation Forests
- Visualization and dimensionality reduction
  - Principal Component Analysis (PCA)
  - Kernel PCA
  - Local Linear Embedding
  - t-distributed Stochastic Neighbor Embedding
- Association rule learning
  - Apriori
  - Eclat

For example, say you have a lot of data about your blog's visitors. You can run an unsupervised model to group them without ever giving info about what groups they are in.

## 1.2. Semisupervised Learning

## Definition 1.2.1

Some algorithms can deal with partially labeled training data, usually a lot of unlabeled data and a little bit of labeled data. A good example is google photos when it detects some of your pictures faces to get the rest.

## 1.3. Reinforcement Learning

## Definition 1.3.1

**Reinforcement Learning** has a learning system called the *agent* that observes the environment, selects and performs actions, and gets rewards/penalties based on its actions. It then find its best strategy, called its *policy*, to get the most reward.

## 1.4. Batch Learning

## Definition 1.4.1

In **batch learning**, the system is incapable of learning incrementally: it must be trained using all the available data. This often takes a huge amount of time and lots of computational resources, so it is often done offline, then once fully done it can be used, This is called *offline learning*.

While simple, this can take many hours of training and is not often ideal, also requiring many computing hours.

## Definition 1.4.2

In **online learning**, you train the system incrementally by feeding it data instances sequentially, in “mini batches”. This is especially useful when you have a continuous flow of new information.

These systems are especially good because they don’t take much computation power since you don’t need to retrain many times.

## Note 1.4.1

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