

Informations

Mes coordonnées

Abdulrahman ALLOUCHE

Bâtiment Brillouin (2^{ème})
allouchea@univ-lyon1.fr

Enseignants

Cédric Ray ; cedric.ray@univ-lyon1.fr (3 premières séances de TP)

Abdulrahman Allouche ; allouchea@univ-lyon1.fr (CM+reste des TP)

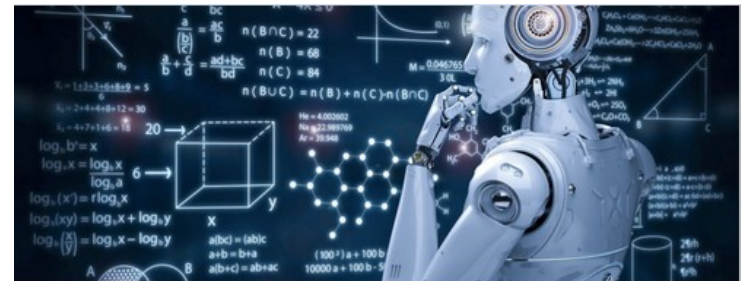
Supports

moodle.univ-lyon1.fr

Mes cours actuels :

Intelligence Artificielle pour la physique

Tableau de bord



Informations

Prérequis

- Quelques notions en algèbre linéaire
- Avoir déjà programmé

Modalité de contrôle de connaissance :

- **Contrôle partiel : Projet (binôme)** début vers 8 décembre: 2 semaines pour rendre le projet
- 1^{ère} session : **Examen individuel sur machine : 1h30**
- 2^{ème} session : **Examen individuel sur machine : 1h00**

Note Finale : 0,4 Projet + 0,6 Examen individuel

Logiciels nécessaires : Anaconda. 3 méthodes possibles :

- **jupyter.univ-lyon1.fr (serveur M2) :**
machine personnelle ou machine en salle des TP
- **Machine dans les salles des TP. Tout est déjà installé**
- **Machine personnelle en local :**
Installer Anaconda AVANT le début de TP
Voir le fichier **InstallAnaconda.pdf** sur moodle (Students/Courses/Tutorials)

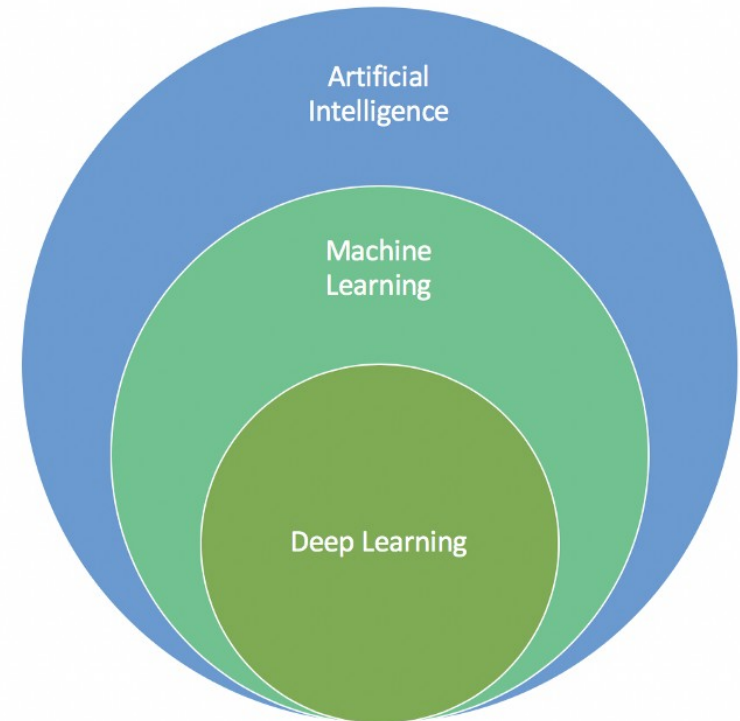
Some definitions

Artificial Intelligence (AI) :

A field of computer science that aims to make computers achieve human-style intelligence.

Machine Learning (ML) :

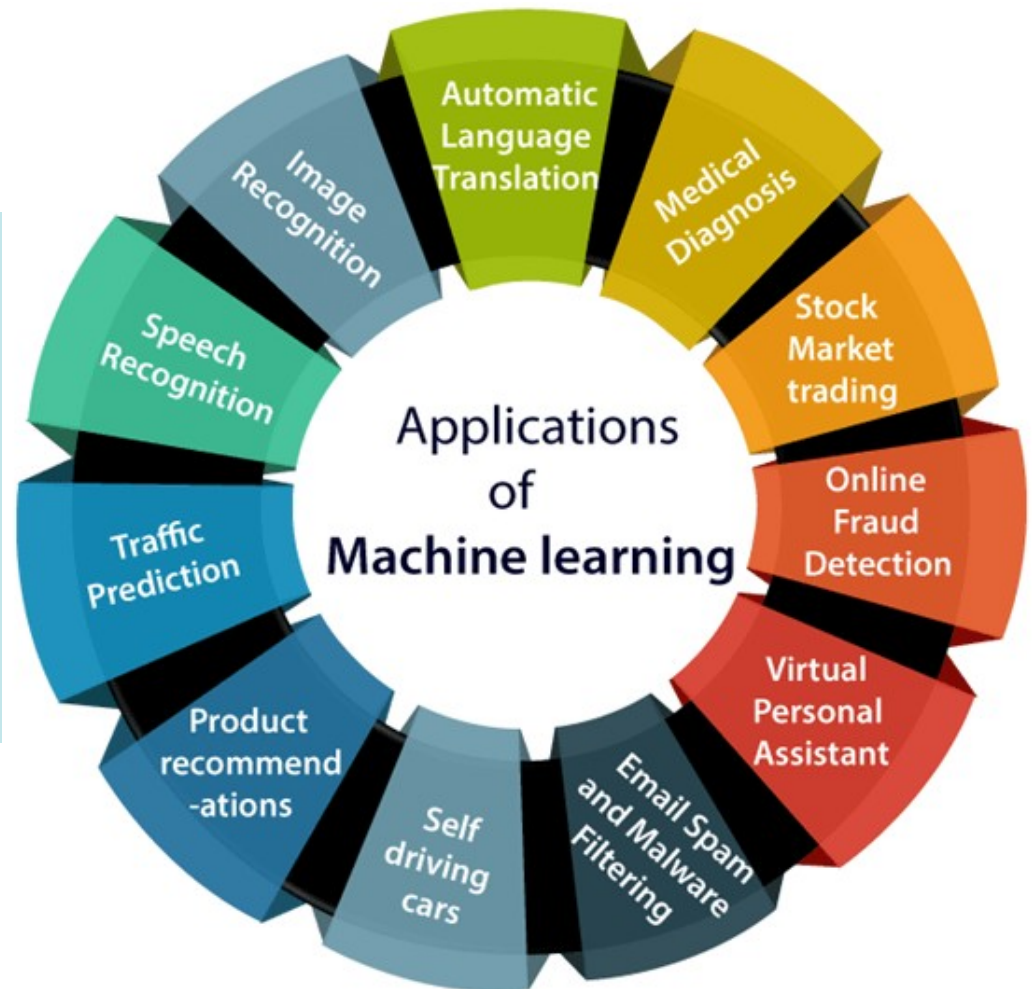
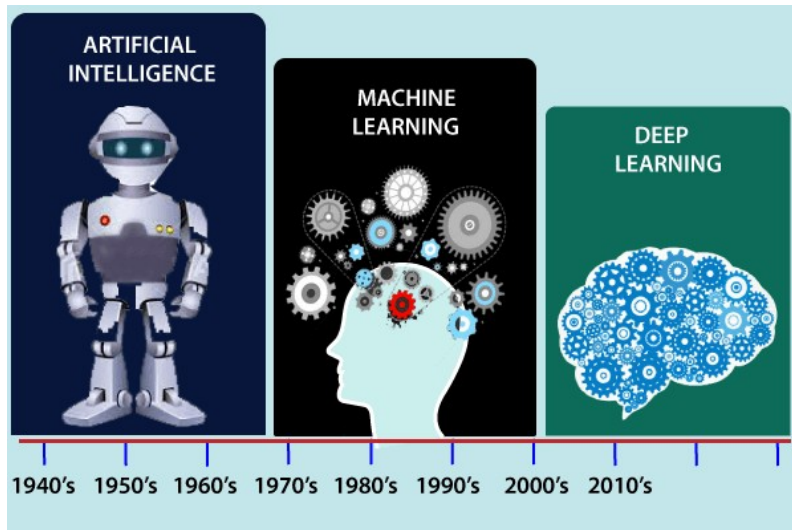
- Subset of AI.
- The science (art) of programming computers how to **learn from data**.
- ML algorithms are **not only "instructions"** to the computer but **dynamic models that analyze how well things are working**, using the information they get to improve. it changes and outputs an algorithm as well as an output



Deep Learning :

- Subset of Machine Learning
- Use artificial **neural networks** réseau de neurone
- Analyzes data with a **logical structure similar to how a human** would draw conclusions

Applications



Applications

iLM

- Machine learning interatomic potential for transformation-induced plasticity of zirconia
- Repulsive potentials for the DFTB method from neural networks
- Control of non-equilibrium systems
- Biomedical image analysis with CNNs
- Imagerie LIBS et Intelligence Artificielle : une belle combinaison !
- Prediction of laboratory quakes with machine learning
- Imagerie LIBS et Intelligence Artificielle : Etude des mortiers archéologique
- Laboratory evaluation of the scattering matrix of ragweed, ash, birch and pine pollens towards pollen classification
- Quantum Mechanics and Machine-Learning Calculations for Vibrational Frequencies
- Neural network approach for a rapid prediction of metal-supported borophene properties
-

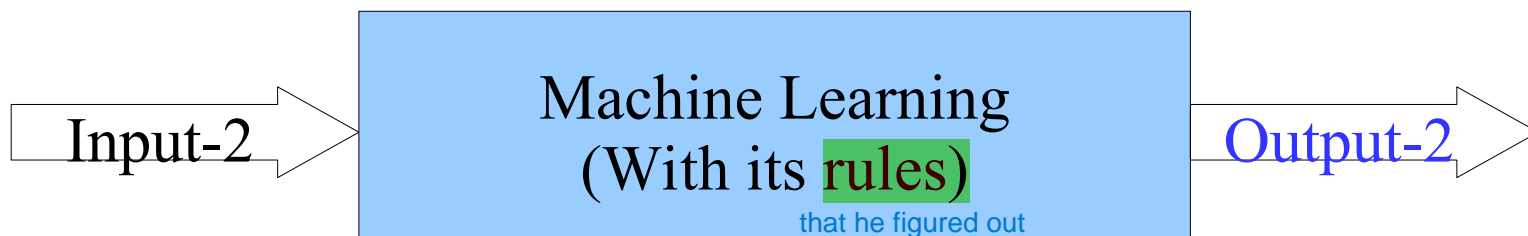
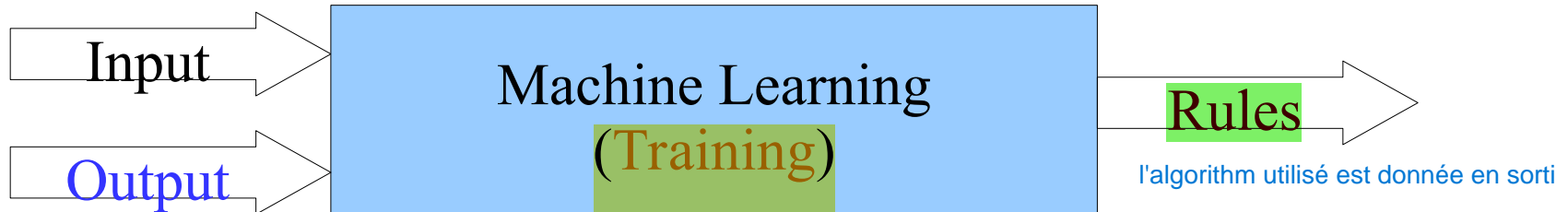
TP

- Project : Prediction of atomization energies of molecules. ML methods without Schrodinger eq will be used, multiple approaches
- COVID-19 can be detected from the Raman spectrum of serum using machine learning. an exam exercise example
- *Model capable of classifying images of moles into benign and malignant*
- *Recurrent Neural Networks, forecast weather based on input features (T , P , ...)* optional

Machine Learning vs. Traditional Programming

What we give at start

What we get at the end



Machine learning Life cycle

3. Build Model

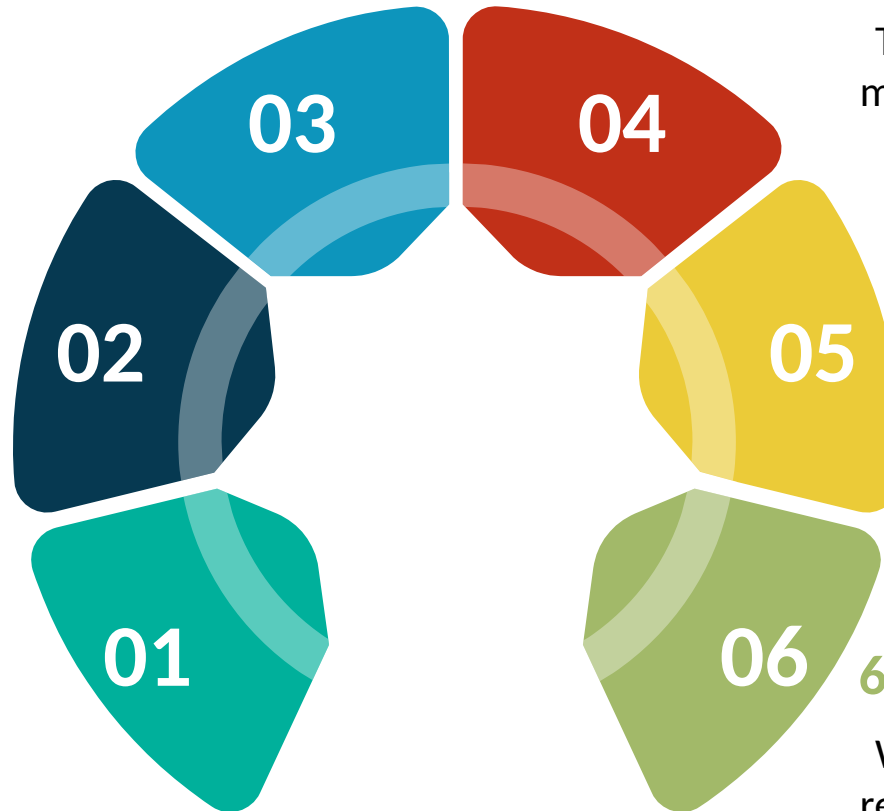
- Determination of the machine learning techniques
- Build the model

2. Data Preparation

- Data exploration: characteristics, format, quality
- Missing Values
- Duplicate data
- Invalid data
- Noise
- Data pre-processing: Standardization, Convert categorical data

1. Gathering Data

- Identify various data sources
- Collect data
- Integrate the data/different sources



4. Train Model

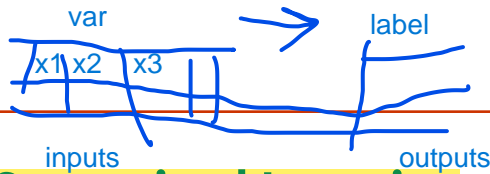
Train the model using various machine learning algorithms

5. Test Model

Check the accuracy of our model using a test dataset

6. Deployment

We deploy the model in the real-world system.



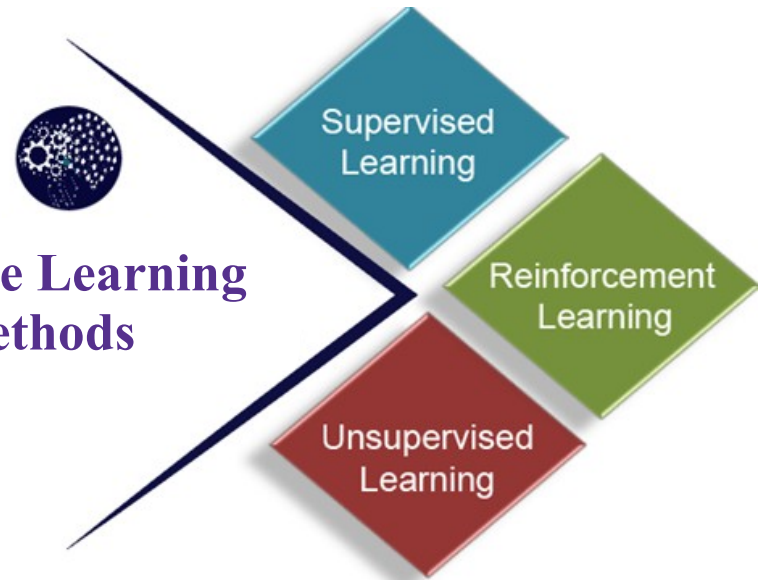
Supervised Learning and Unsupervised Learning will be treated

Machine Learning Methods

Supervised Learning

- Map input data with the output data.
- Supervision, as student learns in the supervision of the teacher.
- Example : spam filtering
- Can be grouped in two categories :
 - Classification *un entier/ true false compte aussi*
 - Regression *les labels sont des reels ex spectre d'energie*

Machine Learning Methods



Unsupervised Learning

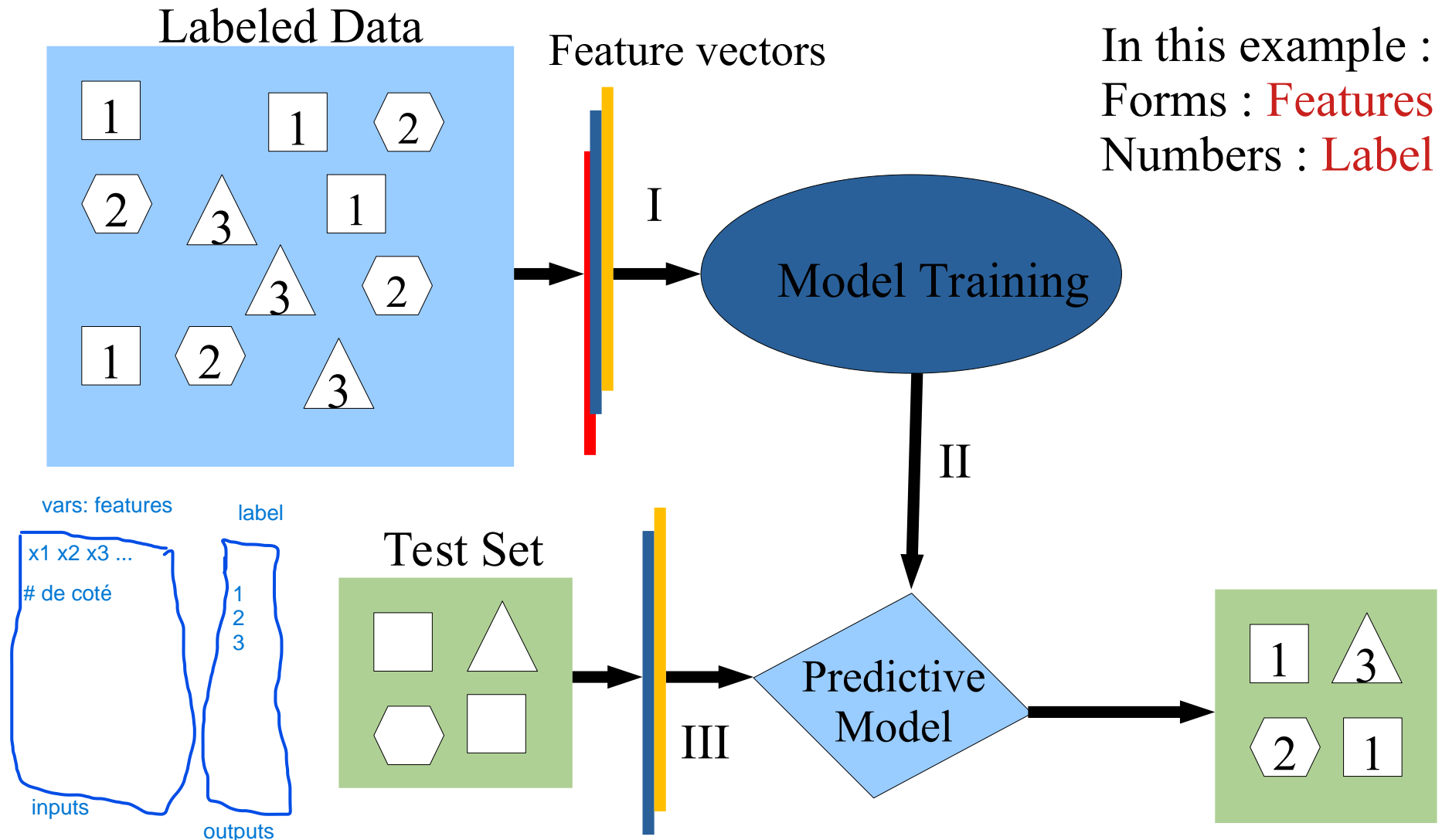
- Data **not labeled**, classified, or categorized
- The algorithm needs to act on that data without any supervision.
- Can be grouped in two categories :
 - Clustering
 - Dimensionality Reduction
diminuer le nombre de variable en eliminant des colonnes de var redondant ou tres correlée

Reinforcement Learning

- Feedback-based learning method
- Agent gets a reward for each right action and gets a penalty for each wrong action.
- Example : The robotic dog, which automatically learns the movement of his arms,

Exmple: consider vars of forms as labeled data set

Supervised learning

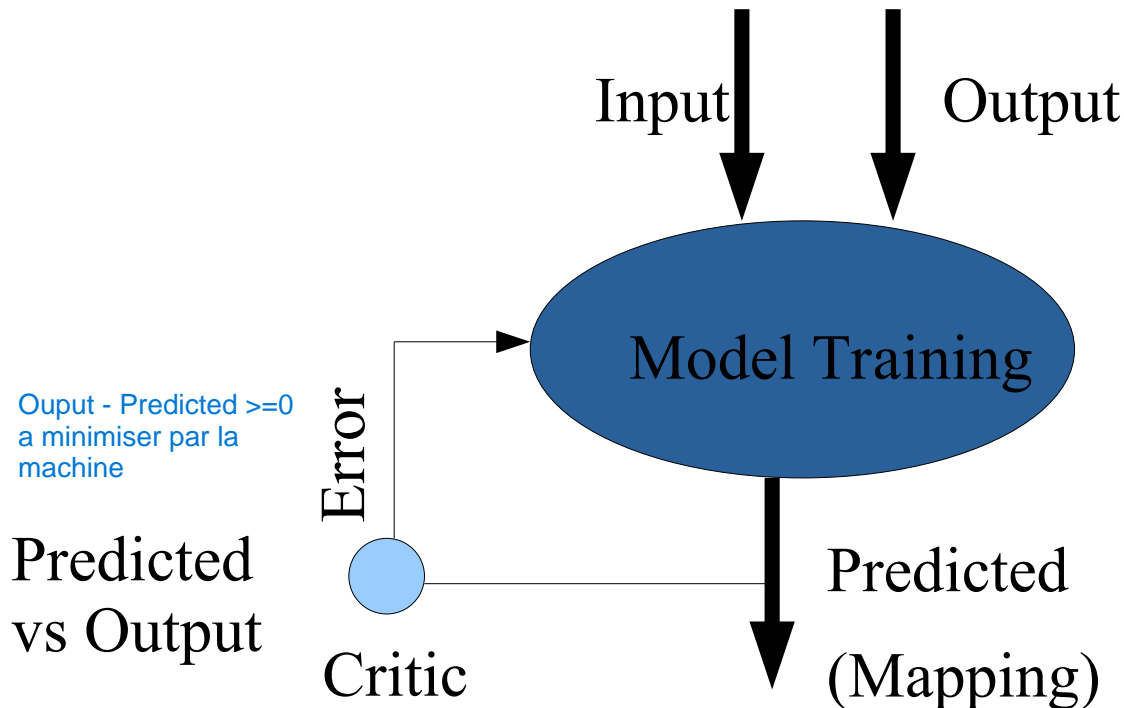


Supervised learning

- You have **input variables** (X) and an **output** variable (Y) and you use model to learn the **mapping function from the input to the output**.

$$Y = f(X)$$

- When you have new input data (X) that you can **predict** the output variables (Y) for that data.



Model : Given algorithm
(the form of the function)

on donne la forme de la fonction avec des parametre aleatoire initial que la machine va optimiser et trouver

Train : Search the optimal
parameters of the function

There are two types of predetermined functions that are used in Supervised Learning

Supervised learning

Supervised Learning

```
graph TD; A[Supervised Learning] --> B[Regression]; A --> C[Classification];
```

Regression

The **output variable is a real** value such as “dollars” or “weight”

Linear regression

Polynomial Regression

Ridge Regression

Lasso Regression

Support vector Regression

Trees Regression

k-Nearest Neighbors

Random Forest Regression

Neural Networks

Classification

The **output variable is an integer** or a category, such as “red” or “blue”

Logistic Regression

Support vector Machines

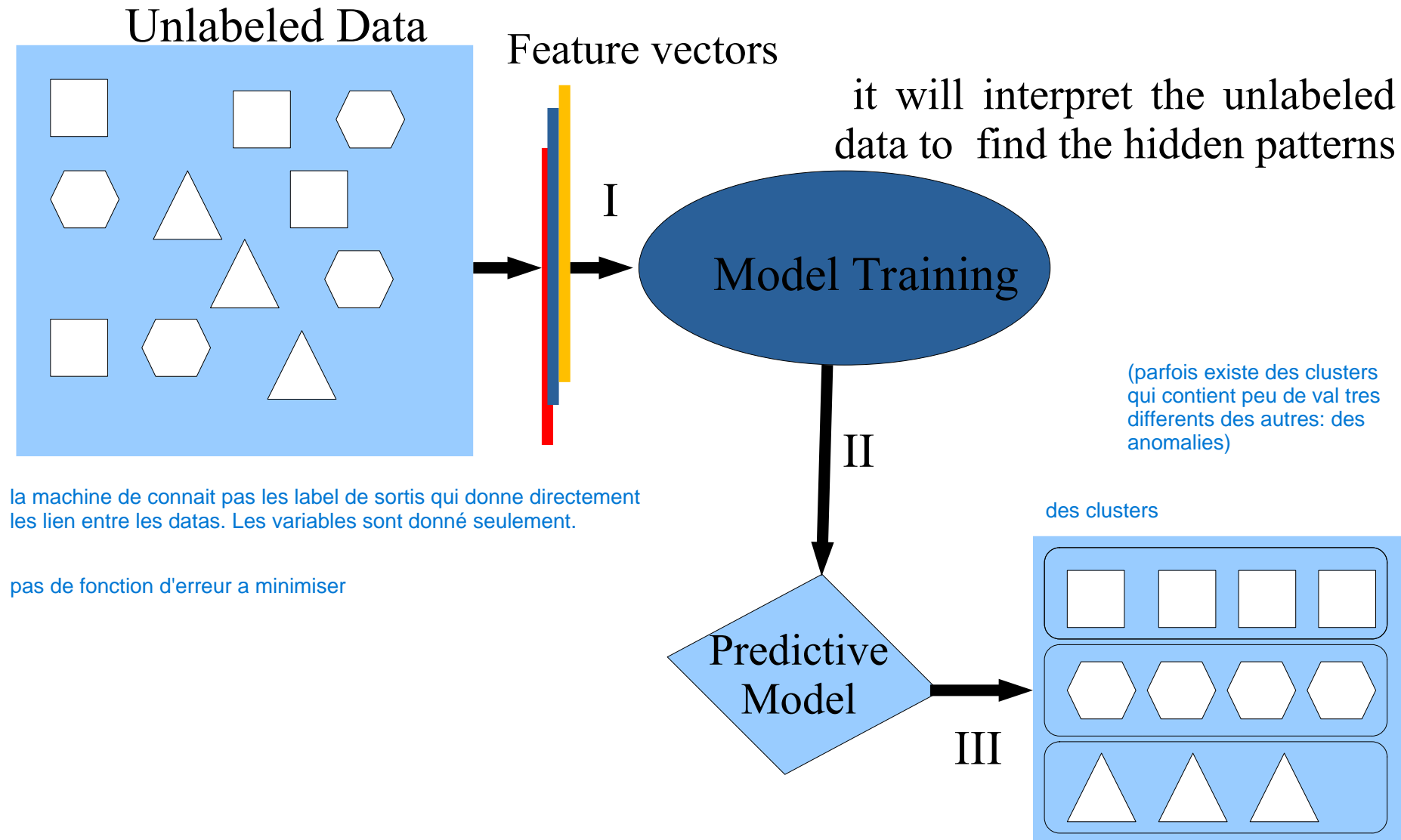
Decision Trees

k-Nearest Neighbors

Random Forest

Neural Networks

Unsupervised learning



Unsupervised learning methods

Unsupervised Learning

```
graph TD; A[Unsupervised Learning] --> B[Dimensionality Reduction]; A --> C[Clustering];
```

Dimensionality Reduction

Conducted to reduce the variable space before analysis

- PCA : Principal Component Analysis
la (+) utilisé, pour # de var grd
- t-NSE : t-distributed Stochastic Neighbor Embedding
1, 2, 3 dimensions, pour la visualisation
- Neural Networks
Pour de très grande base de données

Clustering

Discover groupings inside the input data

- K-means clustering
- DBSCAN : Density-based spatial clustering of applications with noise
(+) de tps de calcul
- Isolation Forest (Anomaly)
efficace mais cherche seulement des anomalies
ex: spam d'e-mail
- Neural Networks

Tools needed to make Machine Learning

- **Anaconda** : Environment management for any language : Python, R, C/++, Fortran,
- **Python** : The reference language for Machine Learning.
- **NumPy** stands for Numerical Python
- **Matplotlib** is a comprehensive library for visualizations in Python.
- **Seaborn** is a Python data visualization library based on matplotlib. It provides a high-level interface for **drawing attractive and informative statistical graphics**.
- **Pandas** is a fast, powerful, flexible and easy to use open source **data analysis** and manipulation tool, built on top of the Python programming language.
- **Scikit-learn** (formerly scikits.learn and also known as sklearn) is a free software machine learning library for the Python programming language. It features **various** classification, regression and clustering **algorithms**
- **Tensorflow** : TensorFlow is the Google open source platform for machine learning. We use it for deep learning models.

See **InstallAnaconda.pdf** document to install all these tools (under anaconda)
(~ 10 GB without Tensorflow)

Course agenda

CM (9h00)

- General Introduction, 1h00
- Python Tutorial , 1h30
- NumPy Tutorial, 30 mn
- Pandas Tutorial, 30 mn
- Data Visualization Tutorial, 30 mn
- Supervised Learning Methods, 2h00
- Unsupervised Learning Methods, 1h00
- Introduction to Deep Methods, 2h00

Practical (13h30)

- Python, 1h30
- NumPy, 30mn
- Pandas & Visualization, 2h00
- Supervised & Unsupervised Methods, 3h30
- Project using ML Standard Methods, 1h30
- Deep Learning Methods, 3h00
- Project using Deep Learning, 1h30

To do at home

Before the first Practical session :

- Read [InstallAnaconda.pdf](#)
- [Install Anaconda](#) on your machine
- Read [JupyterTutorial.pdf](#)

• Before the Visualization Tutorial session :

- Read [CrashCoursesMathPlotLib.pdf](#)