

# **Machine learning introduction**

Simon Gay

# Introduction to machine learning

- **What is the purpose of this session ?**
  - Learning (and recognizing) the most common forms of AI
    - ... And the application domain of each form
  - Being able to implement basic mechanisms of a neural network
  - Being able to use high level tools to implement a deep neural network
- **What we will not do**
  - Creating a complete deep learning AI from scratch
  - Creating a AI to dominate the world

# Introduction to machine learning

People with no idea about AI  
saying it will take over the world:



My Neural Network:



# Introduction to machine learning

- **Introduction**
- **I - From neuron to multi-layer network**
  - The biological neuron
  - The formal neuron
  - Multi-layer networks
- **II - Deep learning**
  - Deep neural networks
  - Generative antagonist networks
  - Libraries for deep learning : TensorFlow and Keras
- **Github : <https://github.com/gaysimon/ARTISAN2022>**

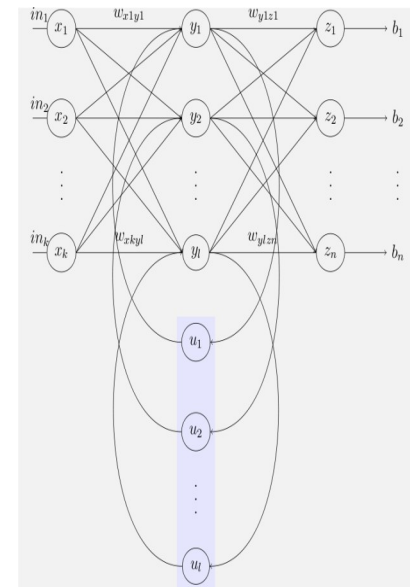
# Introduction to machine learning

- **What is (artificial) Intelligence ?**
  - Learning capacities (Beckmann, 2006)
  - Abilities to adapt to a context (Piaget, 1953)
  - Ability to solve problems using new solutions (Gardner, 1989)
- **Artificial Intelligence aims at reproducing these properties on an artificial system**
- **Several approaches were created to reach this goal**

# Introduction to machine learning

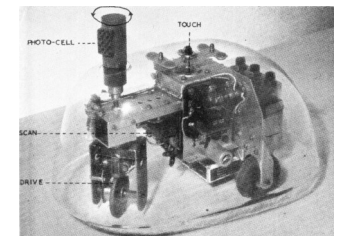
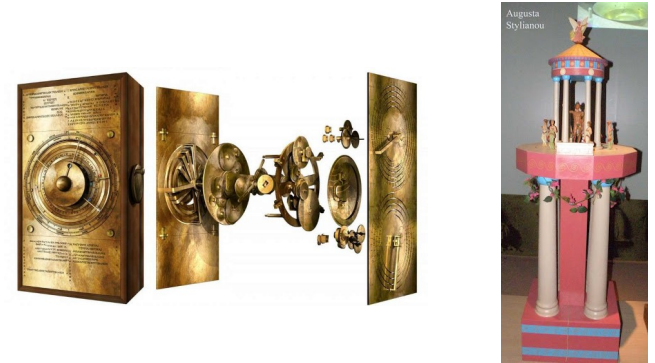
- Since the beginning of AI, two main approaches emerged :
  - **Symbolic AI:** algorithms manipulating symbols and logic rules (inference and deduction)
  - **Connectionist AI:** algorithms made of elements connected with values, and learning mechanisms modifying these values to converge to a solution

```
mother_child(trude, sally).  
  
father_child(tom, sally).  
father_child(tom, erica).  
father_child(mike, tom).  
  
sibling(X, Y)      :- parent_child(Z, X), parent_child(Z, Y).  
  
parent_child(X, Y) :- father_child(X, Y).  
parent_child(X, Y) :- mother_child(X, Y).  
  
?- sibling(sally, erica).  
Yes
```



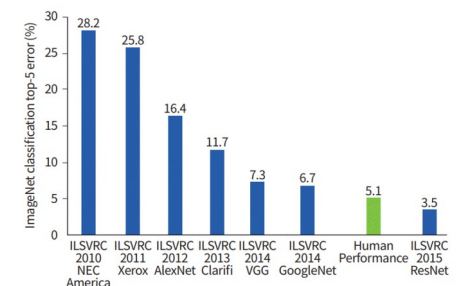
# Introduction to machine learning

- **History of artificial Intelligence**
- **Antiquity: myths of artificial persons...  
... but also first automatons**
- **Middle-age and renaissance**
  - Mechanical knight (Leonardo da Vinci)
  - Vaucanson's automatons (18<sup>th</sup>)
  - The first robot ? Tea serving doll automaton (17<sup>th</sup>)
- **20th century**
  - First electromechanical mobile systems (1912)
  - Karel Čapek : Rossum's Universal **Robots** (1921)
  - emergence of *cybernetics* en 1942
  - First forms of Machine Learning (1951)
    - Logistic Regression (1944?), K-nearest neighbors
- **1956: Artificial Intelligence officially becomes a research domain (Dartmouth conference)**



# Introduction to machine learning

- **1956 : Effervescence and optimism.**
  - First neural networks (1957), Logic programming (symbolic AI, 1958), K-means (1967)...
- **1974 : crash of AI**
  - Facing the lack of results, funds are canceled in laboratories all around the world
- **Beginning of 80s : renewed interest due to expert systems**
  - Total reign of symbolic approaches
  - Reinforcement Learning (1988), Multi-Agent Systems (middle of 80s)
  - First multi-layer neural systems (but nobody cares)
- **Middle of 90s : lack of progress in expert systems → Second crash of AI**
- **1997 : Chess computer Deep Blue defeats world champion Kasparov** : Strongly renewed interest in research and industry
  - Developmental robotics (middle 90s) and intelligence (middle 2000s)
  - Emergence of neural networks with high number of layers (~2010) (but nobody believes in it)
- **2012 : A deep learning model breaks all records at ILSVRC**
  - Total reign of connectionism approaches





# Introduction to machine learning

- **Different types of AI**

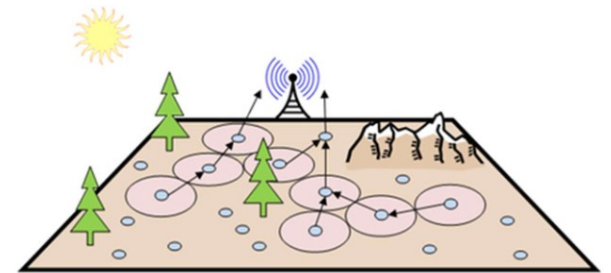
**!! there is no official classification**

- **Symbolic: inference and deduction from symbols and rules**
  - Logic programming
  - Expert Systems
- **Connectionism: adjustment of numerical values and connections between entities**
  - Neural networks
  - Logistic regression
  - Genetic algorithms
  - Reinforcement learning
  - ....
- **Emergence: solving problems through interaction between simple entities**
  - Multi-agent systems

# Introduction to machine learning

- **Emergence: multi-agent systems**

- Partially inspired by social insects (ants, bees...)
- Systems based on interacting simple entities
- Complex global behavior emerges from simple individual behaviors of entities (*The whole is more than the sum of parts*)



- **Applications**

- Simulation of complex phenomena (crowd, fire propagation, ...)
- Optimization of systems, message routing
- Cinema and video games (simulation of large armies, e.g. The Lord of the Rings)



# Introduction to machine learning

- Different types of AI

- **Symbolic: inference and deduction from symbols and rules**

- Logic programming
- Expert Systems

- **Connectionism: adjustment of numerical values and connections between entities**

- Neural networks
- Logistic regression
- Genetic algorithms
- Reinforcement learning
- ....

- **Emergence: solving problems through interaction between simple entities**

- Multi-agent systems

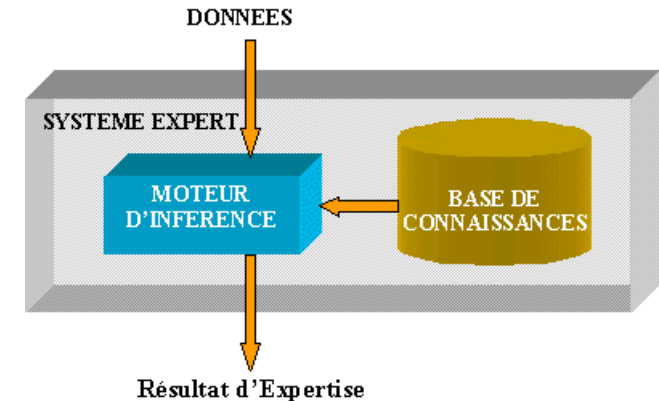
# Introduction to machine learning

- **Symbolic AI**

- Set of rules provided by the conceptor
- Relies on the manipulation of symbols
- Most known model: The expert system
  - Aims at reproducing the knowledge of an expert human in a specific domain
- Logic programming languages
  - PROLOG
- No learning (in most models)

- **Applications**

- Widely used to assist decision making in specific domains (e.g. medical fields, diagnostic...)



# Introduction to machine learning

- **Different types of AI**
- **Symbolic: inference and deduction from symbols and rules**
  - Logic programming
  - Expert Systems
- **Connectionism: adjustment of numerical values and connections between entities**
  - Neural networks
  - Logistic regression
  - Genetic algorithms
  - Reinforcement learning
  - ....
- **Emergence: solving problems through interaction between simple entities**
  - Multi-agent systems

# Introduction to machine learning

- **Connectionist AI: a whole family**

- **Supervised learning:** the model is trained on a dataset with known results, then exploited on data with unknown results.
- **Unsupervised learning:** models that can be used with datasets with unknown results. The model must converge toward an optimal solution
- **Reinforcement learning:** The system can interact with its environment and is actively involved in the learning process: the system (or *agent*) performs action and learns from obtained results.

# Introduction to machine learning

- **Connectionist AI: supervised learning**

- Learns from a dataset with known results
  - Requires to first create a dataset of labeled data

- Mathematical approaches:

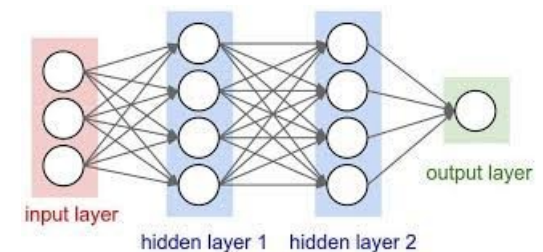
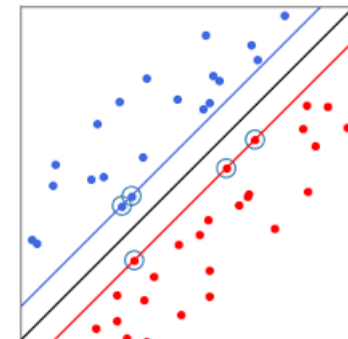
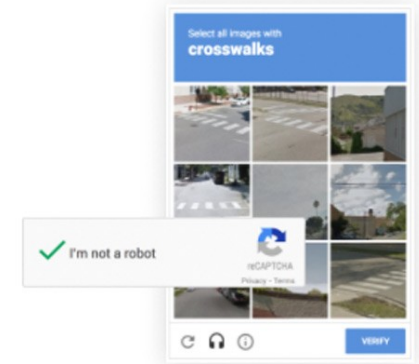
- K-NN
- Logistic regression
- Decision trees
- SVM

- Biologically inspired models:

- Artificial neuron networks

- **Applications**

- Classification and recognition of data, images, text, sound...
- Translation, decision assistance...
  - Note : even a deep learning model cannot 'understand' these data
- ... as soon as a sufficient training dataset is available



# Introduction to machine learning

- **Connectionist AI: a whole family**

- **Supervised learning:** the model is trained on a dataset with known results, then exploited on data with unknown results.

- **Unsupervised learning:** models that can be used with datasets with unknown results. The model must converge toward an optimal solution

- **Reinforcement learning:** The system can interact with its environment and is actively involved in the learning process: the system (or *agent*) performs action and learns from obtained results.



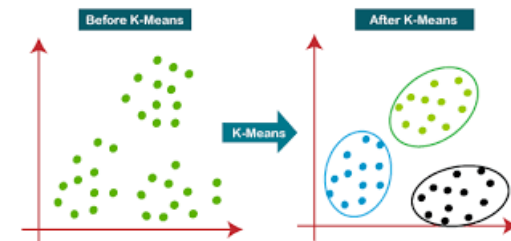
# Introduction to machine learning

- **Connectionist AI: unsupervised learning**

- Learning on data with unknown results
  - These approaches are used to organize data
  - The system converges to an optimal organization (or clustering) of data

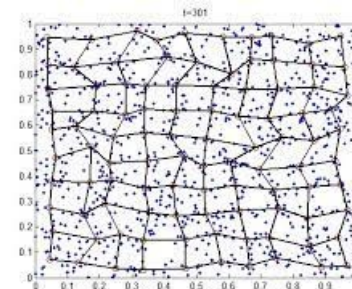
- Mathematical approaches:

- K-means
- Hierarchical clustering



- Biologically inspired approaches:

- Unsupervised neural networks

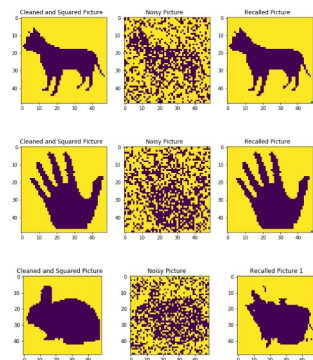
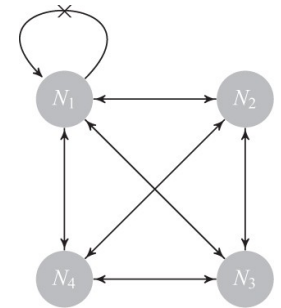


# Introduction to machine learning

- **Some well-known unsupervised neural networks:**

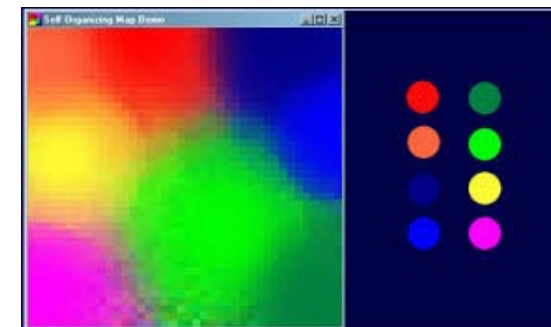
- **Hopfield networks: associative memory**

- Recursive and asynchronous network
- Trained on input data: the network will define a stable configuration for each data
- When presenting an incomplete or noisy data, the network will converge to the closest configuration, reconstructing the original data
- Used to study biological memory principles



- **Self-organizing maps (or Kohonen maps)**

- Inspired by brain's cortex principles
- 2-dimensions networks
- The self organizing map can remap data from a N-dimension space to a 2-D map, while respecting (as much as possible) distance between data.
- Used for complex data analysis and organization



Mapping RGB 3-D space  
on a 2-surface

# Introduction to machine learning

- **Connectionist AI: a whole family**
  - **Supervised learning:** the model is trained on a dataset with known results, then exploited on data with unknown results.
  - **Unsupervised learning:** models that can be used with datasets with unknown results. The model must converge toward an optimal solution
  - **Reinforcement learning:** The system can interact with its environment and is actively involved in the learning process: the system (or *agent*) performs action and learns from obtained results.

# Introduction to machine learning

- **Connectionist AI: Reinforcement learning**

- Embodied intelligence: an artificial agent interacts with an environment
- The agent perceives its environment, acts and analyzes changes
  - The learning process is supervised by the agent itself
- Goal oriented approaches

A predefined reward is attributed when performing a specific task or action

- Reinforcement Learning (1988)
- Q-Learning (1989)

- Behavior oriented approaches :

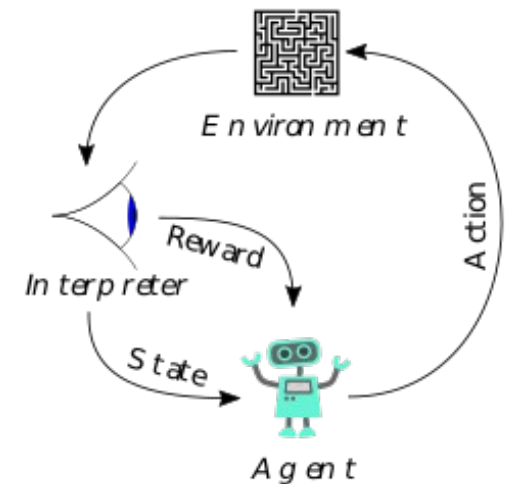
Used to study living beings development and behavior emergence

- Developmental robotics (middle of 90s)
- Developmental learning (middle of 2000s)

# Introduction to machine learning

- **Reinforcement learning (RL): goal oriented approaches**

- The agent can perform a set of actions and perceive the current state of the environment
- Rewards are given when reaching a certain state or performing a specific task
- The agent learns to link states and to define best states to cover to maximize future rewards



- **Q-Learning: a variant of reinforcement learning**

- Consider the best actions to perform instead of state to reach (useful when we do not know how to reach a state)

- **Applications :**

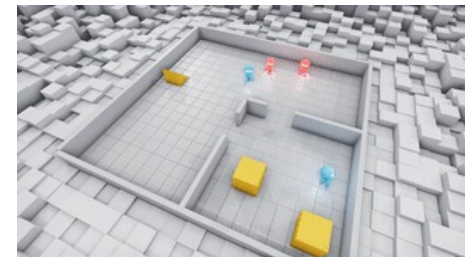
- Robot control, trading, virtual bots, diagnostic (medical fields)...



Learning to Walk via Deep Reinforcement Learning, T. Haarnoja et al., 2019



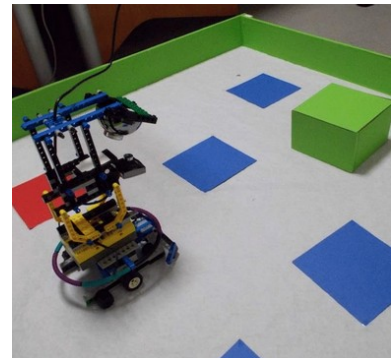
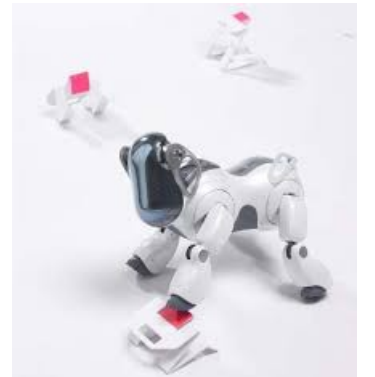
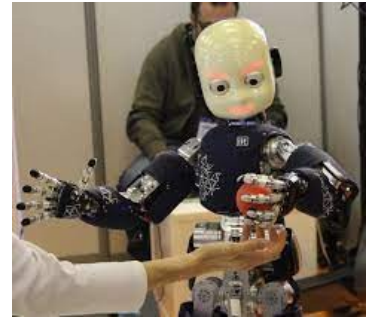
Control of a Quadrotor with Reinforcement Learning, J. Hwangbo et al., 2017



Emergent Tool Use From Multi-Agent Autocurricula, B. Baker et al., 2019

# Introduction to machine learning

- **Reinforcement learning (RL): behavior oriented approaches**
  - **Developmental robotics**
    - Fundamental research domain: understanding sensorimotor development stages on humans (and sometime animals)  
e.g. crawling, hand-eye coordination, standing up, grab an object...
  - **Developmental learning**
    - Fundamental domain research: understanding the emergence of environment interpretation and behavior emergence on more simple agents through their whole 'life'
    - No external reward, agents are fully autonomous and self-motivated
    - Inspired by psychology theories of cognition (ex Piaget, Gibson)



# Introduction to machine learning

- **Reinforcement learning (RL)**
  - **A peculiar model: genetic algorithms**
    - Inspired by species evolution
    - “learning” (optimization) through generations of a population of genes
  - **Principle :**
    - The problem is encoded as a set of parameters: genomes
    - A population is created with random genomes
    - Genomes are evaluated, best genomes are combined to create next generation
    - The genomes are optimized at each new generation
  - **Applications :**
    - Optimization, routing, object design and architecture, electronic circuits, finance and economics...



Spacecraft antenna

# Introduction to machine learning

