Machine learning introduction

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What is the purpose of this session?

- Learning (and recognizing) the most common forms of Al
 - ... 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
- I From neuron to multi-layer network
 - The biological neuron
 - The formal neuron
 - Multi-layer networks
- II Deep learning
 - Deep neural networks
 - Libraries for deep learning : TensorFlow and Keras
 - Generative antagonist networks
 - Text-2-Image

Github: https://github.com/gaysimon/ARTISAN2024

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

- History of artificial Intelligence
- Antiquity: myths of artificial persons...
 - ... but also first automatons



- Mechanical knight (Leonardo da Vinci)
- Vaucanson's automatons (18th)
- The first robot ? Tea serving doll automaton (17th)



- First electromecanical 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)

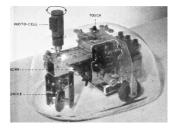






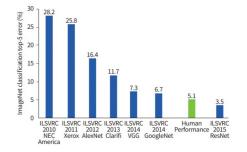






- 1956: Effervescence and optimism.
 - First neural networks (1957), Logic programming (symbolic AI, 1958), K-means (1967)...
- 1974 : crash of Al
 - Facing the lack of results, funds are canceled in laboratories all around the world
- Beginning of 80s: renewed interest due to expert systems
 - Reinforcement 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 Al
- 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 neuronal network breaks all records at ILSVRC
 - Emergence of deep learning





Different types of Al

/!\ there is no official classification

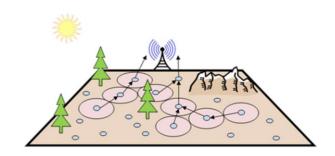
- Symbolic: inference and deduction from symbols and rules
 - Logic programming
 - Expert Systems
- Connectionism: adjustment of numerical values and connections between entities
 - Neural networks
 - Logistic regression
 - Genetic algorithms
 - Reinforcement learning
 -
- Emergence: solving problems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



- 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)



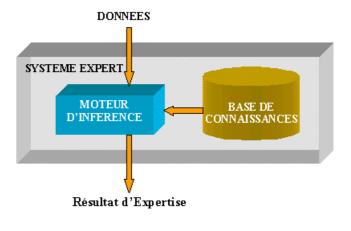


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Symbolic AI

- 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)



```
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
```

mother child(trude, sally).

Applications

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

Different types of AI

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Connectionist AI: supervised learning

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

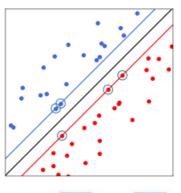


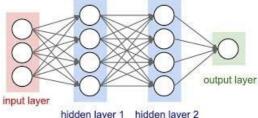
- K-NN
- Logistic regression
- Decision trees
- SVM
- Biologically inspired models:
 - Artificial neuron networks

Applications

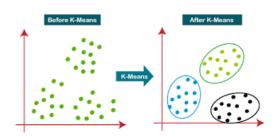
- Classification, recognition and generation of data, images, text, sound...
- Translation, decision assistance...
- ... as soon as a sufficient training dataset is available
 - Note: even a deep learning model cannot 'understand' these data

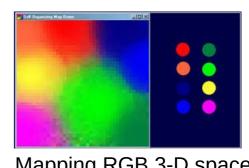






- Connectionist AI: unsupervised learning
 - Learning on data with unknown results
 - → These approaches are used to organize data
 - → The system converges to an optimal organization (or clustering) of data
 - Mathematical approaches:
 - K-means
 - Hierarchical clustering
 - Biologically inspired approaches:
 - Unsupervised neural networks
 - Hopfield networks: associative memory
 - Self-organizing maps (or Kohonen maps)
- Applications: complex data analysis and organization





Mapping RGB 3-D space on a 2-surface

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

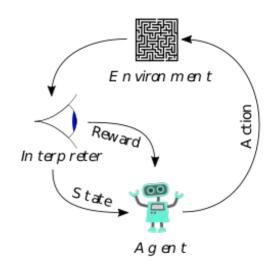
- Reinforcement Learning (1988)
- Q-Learning (1989)
- Behavior oriented approaches :

Used to study living beings development and behavior emergence

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

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 known how to reach a state)

Applications :

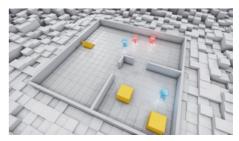
Robot control, trading, virtual bots, diagnostic (medical fields)...



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

Reinforcement learning (RL): behavior oriented approaches

Developmental robotics

• Fundamental research domain: understanding sensorimotor development stages on humans (and sometime animals)

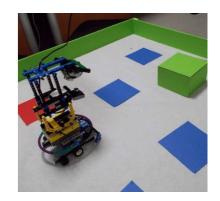
e.g. crawling, hand-eye coordination, standing up, grab an object...





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)



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

