

Università di Verona

A.A. 2020-21

# Machine Learning & Artificial Intelligence

Notizie preliminari

Introduzione

Vittorio Murino

# About the teachers

## **Prof. Vittorio Murino**

Ca' Vignal 2, floor 1, room 1.60

email: [vittorio.murino@univr.it](mailto:vittorio.murino@univr.it)

Q&A: Monday, among  
or after the classes or on demand

<http://www.sci.univr.it/~swan/Teaching>

## **Dr. Geri Skenderi**

Ca' Vignal 2, floor -2, room 141

email: [geri.skenderi@univr.it](mailto:geri.skenderi@univr.it)

Q&A: on demand, ask by email

# Informazioni generali

- Corso teorico e pratico: 7 cfu (56h) di teoria e 2 cfu di laboratorio (24h), con Python.
- Propedeuticità: Probabilità e Statistica, non sarebbe male aver seguito anche corso di Elaborazione delle Immagini.
- Materiale didattico: lucidi del corso (italiano/inglese), libri suggeriti, articoli.
- Gli eventuali seminari di docenti esterni sono inclusi nel materiale didattico del corso.
- Laboratorio di riferimento: VIPS (Vision, Image Processing & Sound).

Info: <http://vips.sci.univr.it>

# About the course

## THEORY (32 hours)



## LABORATORY (24 hours)

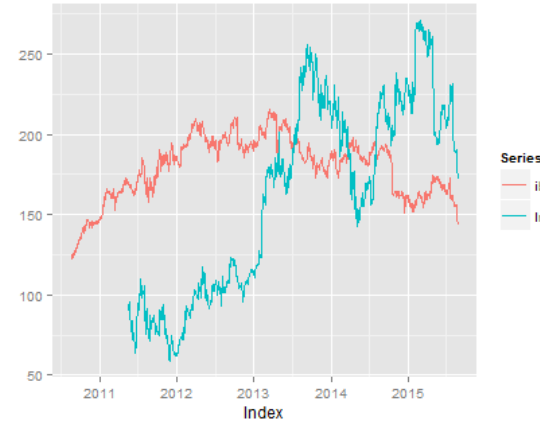


# Modalità di esame

- Esame: progetto, con relazione ed esposizione orale (con presentazione ppt)
  - 2 persone max, 3 persone per progetti più complessi (da concordare col docente)
  - L'esposizione del progetto include una valutazione della conoscenza dei contenuti del corso.
  - Possibile eventuale breve orale per arrotondamenti
- Possibilità di progetto congiunto con corsi affini, previo accordi coi docenti.
- Tesi di Laurea



# About the project and exam



# Course objectives

- The course aims to provide the **theoretical foundations** and describe the main **methodologies** related to **Machine Learning** and **Pattern Recognition** and, more generally, to **Artificial Intelligence**.
- In particular, the course will deal with the methods of **analysis**, **recognition** and **automatic classification** of data of any type, typically called *patterns*.
- These disciplines are at the basis, are used, and often complement many other disciplines and application areas of wide diffusion, such as **computational vision**, **robotics**, **image processing**, **data mining**, **analysis and interpretation of medical and biological data**, **bioinformatics**, **biometrics**, **video surveillance**, **speech and text recognition**, and many others.
- More precisely, the methodologies that will be introduced in the course are often an integral part of the aforementioned application areas, and constitute their "**intelligent**" part with the ultimate goal of understanding (classifying, recognizing, analyzing) the data from the process of interest (whether they are signals, images, strings, categorical, or other types of data).

# Course objectives

- Starting from the type of sensed data, the entire analysis pipeline will be considered such as
  - the extraction and selection of characteristics (*features*);
  - supervised and unsupervised learning methods,
  - parametric and non-parametric analysis techniques,
  - validation protocols,
  - recent deep learning techniques: basic notions, and some advanced topics with case studies.
- In conclusion, the course aims to provide the students with a set of **theoretical foundations** and **algorithmic tools** to address the problems that can be encountered in strategic and innovative industrial sectors such as those involving robotics, cyber physical systems, (big) data mining, digital manufacturing, visual inspection of products/production processes, and automation in general.



# Syllabus

*Theoretical foundations and main methods devoted to data analysis, not necessarily images – Statistical Pattern Recognition*

*Provide a basis for the recent Deep Learning techniques*

- Introduction to the course: pattern recognition, machine learning and artificial intelligence. Systems and applications.
- Bayes decision theory
- Parameter estimation and nonparametric methods
- Linear, nonlinear classifiers and discriminant functions
- Linear transformations, Fisher method, feature extraction and selection, Principal Component Analysis
- Kernel methods and Support Vector Machines

# Syllabus

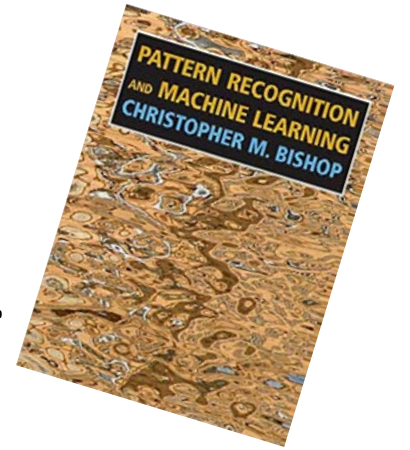
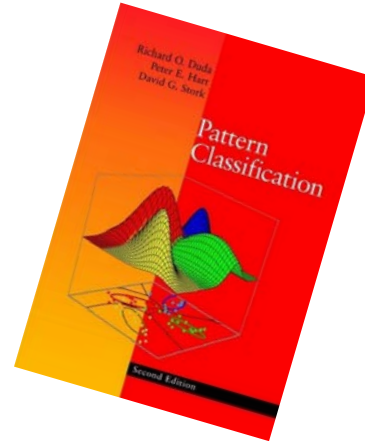
*Theoretical foundations and main methods devoted to data analysis, not necessarily images – Statistical Pattern Recognition*

*Provide a basis for the recent Deep Learning techniques*

- Artificial neural networks
- Unsupervised classification methods, clustering, mixtures models and Expectation-Maximization algorithm
- Analysis of sequential data and Hidden Markov Models
- Model ensemble and fusion
- Deep learning: basic notions, advanced topics and case studies.
- Selected application examples

# Books

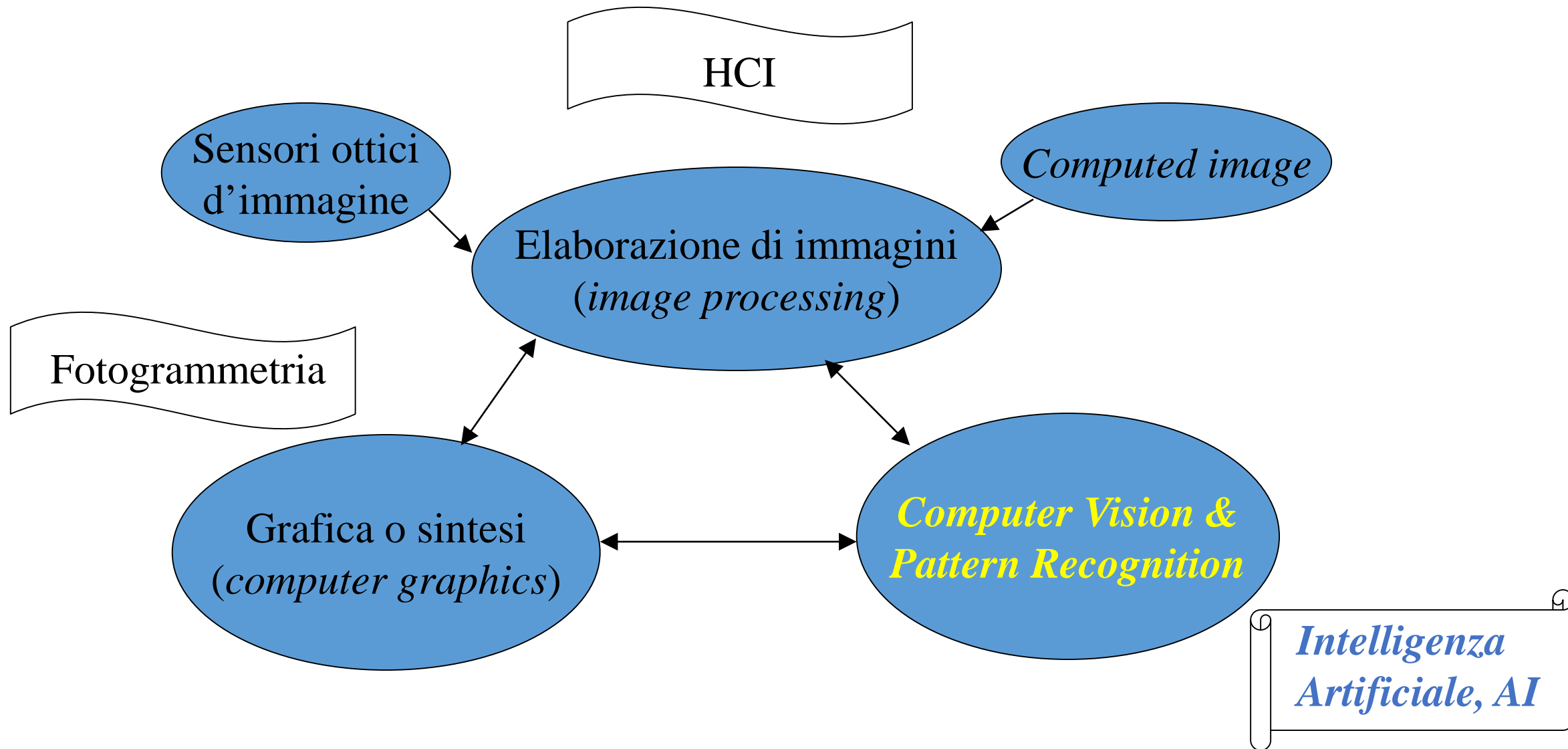
- R. Duda, P. Hart, D. Stork  
*Pattern Classification*, Wiley, 2001.
- C.M. Bishop  
*Pattern Recognition and Machine Learning*, Springer, 2006.
- S. Theodoridis, K. Koutroumbas  
*Pattern Recognition*, Academic Press, 1998.
- C.M. Bishop  
*Neural Networks for Pattern Recognition*, Oxford University Press, 1995.



## Other books/material

- T. Hastie, R. Tibshirani, J. Friedman  
*The Elements of Statistical Learning*. Springer, 2001
- J.T. Tou, R.C.Gonzales  
*Pattern Recognition Principles*. Addison-Wesley, Publishing Co., Reading Mass., 1974.
- K. Fukunaga  
*Introduction to Statistical Pattern Recognition*. Academic Press Inc. 1990.
- Articoli monografici, rassegne  
*... in Internet*

# Le immagini al computer



# Elaborazione di immagini

- **Manipolazione** di una immagine al fine di produrre una nuova versione di essa.
- Immagine originale acquisita da un sensore fisico o da *file*.
- Comprende l'**analisi di immagine**: eseguire calcoli riguardanti aspetti specifici dell'immagine stessa.
- Ricadute importanti per le scienze biomedicali e non solo: ricostruzione da img MRI, SPECT, TAC, ecografi, etc., analisi di misure, ausilio alla diagnosi (con AI, PR), filtraggio e rinforzo, etc.



# Considerazioni sull'elaborazione di immagini

- È la disciplina più vecchia: albori durante la II guerra mondiale per quantificare gli effetti dei bombardamenti
- Nel 1960 molto è già sviluppato, per migliorare le immagini da satellite
- Oggi, maggiore migrazione verso nuove applicazioni che sviluppo di metodi innovativi
- L'avvento di PC e monitor potenti ed economici la rende di dominio pubblico
- È spesso usata al servizio della *computer vision* e della grafica (contro *aliasing* e *sampling*)

# *Computer vision*

- **Insieme di tecniche computazionali per stimare le proprietà geometriche e dinamiche del mondo 3D da una o più immagini.**
- **In senso lato, estrarre informazioni** da un'immagine per produrre una **rappresentazione o descrizione** della scena
- Ha finalità inverse alla grafica.
- Più ambiziosa dell'analisi d'immagine. Vorrebbe emulare il sistema visivo umano
- Ricavare informazioni 3-D da immagini 2-D, assegnare etichette, stimolare/guidare azioni

# *Pattern Recognition*

- Difficile da identificare, molte sovrapposizioni con CV
- Descrizione ed analisi delle misure fatte da processi fisici o mentali
  - richiede una fase di pre-elaborazione per ridurre rumore e ridondanza delle misure
  - uso della conoscenza disponibile sulle proprietà statistiche e strutturali delle misure
- Spesso, si identifica con la classificazione (di dati numerici e simbolici)
- Nuove discipline e applicazioni ne fanno uso: data mining, web semantico, intrusion detection

# Pattern Recognition

- *Cluster analysis*

- analisi dei dati per trovare inter-relazioni e discriminarli in gruppi (senza conoscenza a priori)

- Estrazione e selezione delle *feature*

- ridurre dimensionalità dei *pattern* (insiemi di misure)

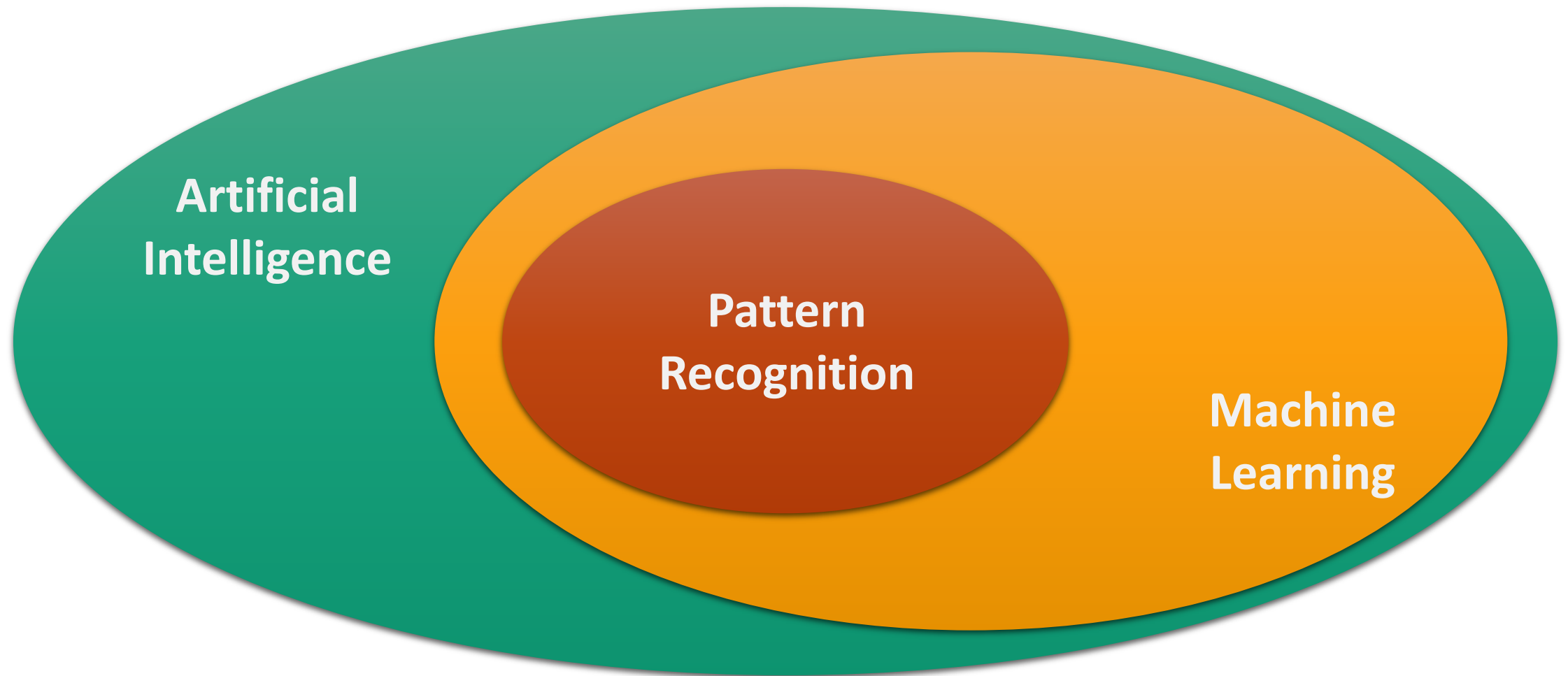
- Classificazione

- Strutturale (sintattica):
  - ogni pattern è espresso come composizione di primitive e si determina un'analogia tra la struttura del pattern e la sintassi di un linguaggio  $\Rightarrow$  riconoscimento visto come *parsing* (secondo le regole di sintassi)
- Statistica

# Considerazioni su CV & PR

- Coetanea dell'elaborazione d'immagine e più vecchia della grafica
- In senso lato include la classificazione (*pattern recognition*)
- Difficoltà e assenza di soluzioni in molte applicazioni che ne necessitano
- È spesso collegata ad una funzione robotica ed è volta ad imitare la visione umana
- Ad oggi è un problema aperto. Nei casi non vincolati, i sistemi esistenti non sono soddisfacenti
- La **conoscenza** e l'**esperienza** sono fattori predominanti nella visione umana

Where we are...





*Machine Learning*  
~~Pattern Recognition~~

# Pattern Recognition vs. Machine Learning

*Pattern Recognition has its origins in engineering, whereas Machine Learning grew out of computer science. However, these activities can be viewed as two facets of the same field, ...*

*Christopher M. Bishop*



# What is Machine Learning?

*I think that Machine Learning is about this broader notion of building computational artifacts that **learn over time based on experience**.*



*You have data, you do analysis on that data, you try to glean things from them using various kinds of computational structures: it's **computational statistics!***



# What does “learning” mean?

*A computer program is said to learn from experience  $E$  with respect to a task  $T$  and performance measure  $P$  if its performance at task  $T$ , as measured by  $P$ , improves with experience  $E$ .*

*Tom M. Mitchell*



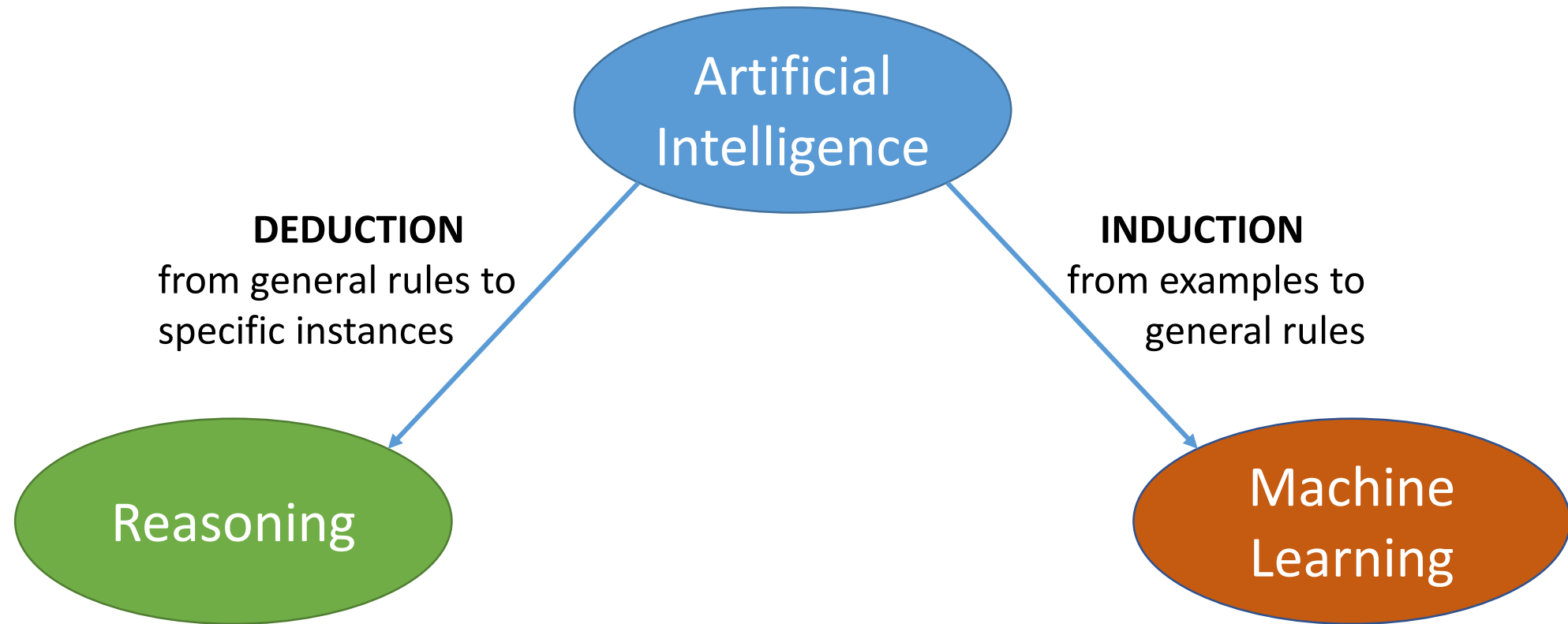
# Types of learning

**Supervised learning:** given a set of pairs  $[\mathbf{x}, y]$  ruled by  $y = f(\mathbf{x})$ , the goal is to find an approximate function  $\hat{f}(\mathbf{x})$  so that given a new  $\mathbf{x}$  we can guess  $y = \hat{f}(\mathbf{x})$ .

**Unsupervised learning:** given a set of data  $[\mathbf{x}]$ , the goal is to find a function  $f(\mathbf{x})$  that gives a compact representation of the given set.

**Reinforcement learning:** given a set of triplets  $[\mathbf{x}, a, r]$  ruled by  $r = f(a|\mathbf{x})$ , the goal is to find a mapping  $\mathbf{x} \rightarrow a$  such that the reward  $r$  is maximum.

# Artificial Intelligence vs. Machine Learning



Indeed, AI is covering all aspects related to intelligence, from perception to action, passing through learning and reasoning



# Artificial Intelligence

- The field of **Artificial Intelligence**, or AI, goes further still: it attempts not just to understand but also to *build* intelligent entities.
- Work started in earnest soon after World War II, and the name itself was coined in 1956.
- **AI** currently encompasses a huge variety of subfields, ranging from the general (*learning and perception*) to the specific, such as playing chess, proving mathematical theorems, writing poetry, driving a car on a crowded street, and diagnosing diseases.
- AI is relevant to any intellectual task; it is truly a universal field.

# Artificial Intelligence

<b>Thinking Humanly</b> “The exciting new effort to make computers think . . . <i>machines with minds</i> , in the full and literal sense.” (Haugeland, 1985) “[The automation of] activities that we associate with human thinking, activities such as decision-making, problem solving, learning . . .” (Bellman, 1978)	<b>Thinking Rationally</b> “The study of mental faculties through the use of computational models.” (Charniak and McDermott, 1985) “The study of the computations that make it possible to perceive, reason, and act.” (Winston, 1992)
<b>Acting Humanly</b> “The art of creating machines that perform functions that require intelligence when performed by people.” (Kurzweil, 1990) “The study of how to make computers do things at which, at the moment, people are better.” (Rich and Knight, 1991)	<b>Acting Rationally</b> “Computational Intelligence is the study of the design of intelligent agents.” (Poole <i>et al.</i> , 1998) “AI . . . is concerned with intelligent behavior in artifacts.” (Nilsson, 1998)

# Artificial Intelligence

- **Thinking humanly:** cognitive science, information processing psychology, requires (neuro)scientific theories of internal activities of the brain, from behaviour to neural correlates and vice versa
- **Thinking rationally:** laws of thoughts, normative rather than descriptive, notation and rules of derivation for thoughts, from math and philosophy to modern AI
- **Acting rationally:** rational behaviour, doing the right thing, maximising goal achievement, not necessarily involve thinking (e.g., blinking reflex)
  - *Aristotele*: every art and every enquiry, and similarly every action and pursuit, is thought to aim at some good

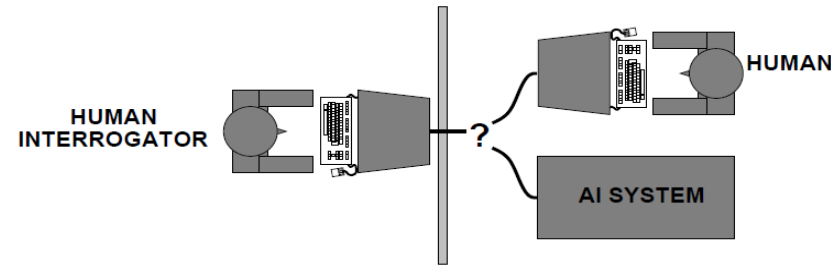
# Artificial Intelligence

## Acting humanly: Turing Test



Turing (1950) “Computing machinery and intelligence”:

- ◇ “Can machines think?” → “Can machines behave intelligently?”
- ◇ Operational test for intelligent behavior: the Imitation Game



- ◇ Predicted that by 2000, a machine might have a 30% chance of fooling a lay person for 5 minutes
- ◇ Anticipated all major arguments against AI in following 50 years
- ◇ Suggested major components of AI: knowledge, reasoning, language understanding, learning

Problem: Turing test is not reproducible, constructive, or amenable to mathematical analysis

# Artificial Intelligence

- A computer would need to possess the following capabilities:
    - **natural language processing** to enable it to communicate successfully in English;
    - **knowledge representation** to store what it knows or hears;
    - **automated reasoning** to use the stored information to answer questions and to draw new conclusions;
    - **machine learning** to adapt to new circumstances and to detect and extrapolate patterns.
- Total Turing Test {
- **computer vision** for understanding visual scenes by image/videos
  - **robotics**, to manipulate objects

# Artificial Intelligence

- Rational agents

An agent is an entity that perceives and acts

Abstractly, an agent is a function from percept histories to actions:

$$f : \mathcal{P}^* \rightarrow \mathcal{A}$$

For any given class of environments and tasks, we seek the agent (or class of agents) with the best performance

Caveat: computational limitations make perfect rationality unachievable  
→ design best program for given machine resources

# Artificial Intelligence: AI prehistory

Philosophy	logic, methods of reasoning mind as physical system foundations of learning, language, rationality
Mathematics	formal representation and proof algorithms computation, (un)decidability, (in)tractability probability
Psychology	adaptation phenomena of perception and motor control experimental techniques (psychophysics, etc.)
Linguistics	knowledge representation grammar
Neuroscience	physical substrate for mental activity
Control theory	homeostatic systems, stability simple optimal agent designs

# Artificial Intelligence: a little history

- 1943 McCulloch & Pitts: Boolean circuit model of brain
- 1950 Turing's "Computing Machinery and Intelligence"
- 1952–69 Look, Ma, no hands!
- 1950s Early AI programs, including Samuel's checkers program, Newell & Simon's Logic Theorist, Gelernter's Geometry Engine
- 1956 Dartmouth meeting: "Artificial Intelligence" adopted
- 1965 Robinson's complete algorithm for logical reasoning
- 1966–74 AI discovers computational complexity  
Neural network research almost disappears
- 1969–79 Early development of knowledge-based systems
- 1980–88 Expert systems industry booms
- 1988–93 Expert systems industry busts: "AI Winter"
- 1985–95 Neural networks return to popularity
- 1988– Resurgence of probabilistic and decision-theoretic methods  
Rapid increase in technical depth of mainstream AI  
"Nouvelle AI": ALife, GAs, soft computing



# Artificial Intelligence: a little history

## Deep Learning

- Convolutional Neural Networks (CNN)
  - LeCun, Bottou, Bengio, Haffner, *Proceedings of the IEEE*, 1998
- Deep Belief Nets
  - Hinton et al., *Neural Computation*, 2006
- Auto-encoder
  - Hinton & Salakhutdinov, *Science*, 2006

