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Machine Learning: II motore dell'Intelligenza Artificiale e dell'investigazione scientifica guidata dai Big Data.

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- RAK AK
- Theory of Computing/Algorithmic Theory
- Algorithms and Data Science

Algorithms and Uncertainty

Algorithmic Game Theory/Economics and Computation

- ERC Adv. Grant " Algorithmic and Mechanism Design

Research in Online MArkets" (AMDROMA) 2018-2023

- Supported by Google and Facebook since 2012

Attività Accademiche

- Dottorato di Ricerca in Data Science [2018]
- Laurea Magistrale in Data Science [2014]
- Laurea Triennale in "Scienze per l'Intelligenza Artificiale" [2022 -]
- Sapienza School of Advanced Studies (SSAS) [2012-2018]
- Cultura e Creatività Digitale presso Fondazione "I Lincei per la Scuola"
 [2018]
- Commissione Scuola Accademia dei Lincei [2020 -]
- Conferenza Lincea "Theory of Computing: A Multidisciplinary Perspective" in occasion of the 54° ACM Symposium on Theory of Computing, Rome 20 – 24 June 2022.



Machine Learning: il motore dell'Intelligenza Artificiale

Ringraziamenti: Fabrizio Silvestri (Sapienza) Fabio Petroni (Facebook) Tancredi Massimo Pentimalli (Berlin School of Integrative Oncology) Stefano Giacu (Sapienza)



Geogle Search

Fm Feeling Lucky























The importance of being on twitter

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TEXT PROMPT an armchair in the shape of an avocado....

AI-GENERATED IMAGES



Edit prompt or view more images↓



Cosa è il Machine Learning?

Unsupervised Machine Learning



What does (unsupervised) learning mean?





Supervised Machine Learning





What does (supervised) learning mean?



Optimization

- Common to all ML (and DL) techniques there is the optimization component
 O We want to find the parameters of a model that minimize/maximize some cost functions.
- An ML model is essentially a member of a family of functions that we call hypothesis
- A model depends generically on parameters $\theta \to P(X;\theta)$
- How do you find the best model?
 - O You pick the best parameters!







The central challenge in ML is that our algorithm must perform well on new, previously unseen inputs

Data Representation





Data Representation



EMBEDDING

Male-Female

Verb tense



The big data revolution





Language Modeling



• More formally: given a sequence of words $x^{(1)}, x^{(2)}, \dots, x^{(t)}$ compute the probability distribution of the next word $x^{(t+1)}$

$$P(\boldsymbol{x}^{(t+1)} | \boldsymbol{x}^{(t)}, \dots, \boldsymbol{x}^{(1)})$$

where $m{x}^{(t+1)}$ can be any word in the vocabulary $V = \{m{w}_1,...,m{w}_{|V|}\}$

• A system that does this is called a Language Model.



A RNN Language Model





Generating text with a RNN Language Model

•You can train a RNN-LM on any kind of text, then generate text in that style.

•RNN-LM trained on Obama speeches:



The United States will step up to the cost of a new challenges of the American people that will share the fact that we created the problem. They were attacked and so that they have to say that all the task of the final days of war that I will not be able to get this done.

https://medium.com/@samim/obama-rnn-machine-generated-political-speeches-c8abd18a2ea0



Transformers (e.g., BERT 2018)1. pretrain2

Huge corpora (B of words)



Unsupervised

This takes days/weeks on several GPUs / TPUs

Pre-trained models publicly available on the web!

2. fine-tune



Supervised

Fine-tune for a specific task on as much labeled data as you have.

Fast !



Reinforcement Learning





Credits Wikipedia

Alpha Zero (Deep Mind 2018)

- A single system that taught itself from scratch how to master the games of chess, <u>shogi</u> (Japanese chess), and <u>Go</u>, beating a world champion program in each case.
- To learn each game, an untrained neural network plays millions of games against itself via a process of trial and error called reinforcement learning.
- At first, it plays completely randomly, but over time the system learns from wins, losses, and draws to adjust the parameters of the neural network, making it more likely to choose advantageous moves in the future.



I can't disguise my satisfaction that it plays with a very dynamic style, much like my own!"

GARRY KASPAROV FORMER WORLD CHESS CHAMPION



Machine Learning e l'investigazione scientifica basata sui big data

Alpha Fold (Deep Mind 2018-2020)

- What a protein does largely depends on its unique 3D structure. Figuring out what shapes proteins fold into is known as the <u>"protein folding problem"</u>, and has stood as a grand challenge in biology for the past 50 years.
- The latest version of AlphaFold, used at CASP14, is an attention-based neural network system, trained end-to-end, that attempts to interpret the structure of the graph between residues in close proximity.
- The system is trained on publicly available data consisting of ~170,000 protein structures from the <u>protein data</u> <u>bank</u> together with <u>large</u> <u>databases</u> containing protein sequences of unknown structure.



0

CASP7

2006

CASP8

2008

CASP9

2010

CASPIO

CASP

2012

CASP11

2014

CASP12 CASP13

2018

2016

CASP14 2020



Alpha Fold (Deep Mind 2018-2020)

This computational work represents a stunning advance on the protein-folding problem, a 50-year-old grand challenge in biology. It has occurred decades before many people in the field would have predicted. It will be exciting to see the many ways in which it will fundamentally change biological research.

PROFESSOR VENKI RAMAKRISHNAN NOBEL LAUREATE AND PRESIDENT OF THE ROYAL SOCIETY We have been stuck on this one problem – how do proteins fold up – for nearly 50 years. To see DeepMind produce a solution for this, having worked personally on this problem for so long and after so many stops and starts, wondering if we'd ever get there, is a very special moment.

PROFESSOR JOHN MOULT CO-FOUNDER AND CHAIR OF CASP, UNIVERSITY OF MARYLAND

In partnership with EMBL-EBI, it has been launched the AlphaFold Protein Structure Database.



ML and disease classification: brain tumors

Problem:

- How many subtypes of brain tumors exist?
- How to identify (i.e. classify and Dx) them?
- Tumors are rare and Dx is not standardized

Relevance:

- Completely changes prognosis and treatment
- A process combining an advanced imaging technology and artificial intelligence (AI) can accurately diagnose brain tumors in fewer than 3 minutes during surgery.

Avaliable data:

 Tumor methylation profiles (Epigenetic marks of active and inactive DNA) (What are the tumor cells doing? Which genes are active?)



CNNs and digital pathology: Al-assisted Dx

Problem:

- How to reliably identify skin cancer?
- Clinicians have different levels of experience

Relevance:

- Skin lesions extremely frequent (most common cancer!)
- Early detection dramatically changes treatment (i.e simple surgical excision)
- Low-cost, widespread access (e.g. smartphones!)



- CNN end-to-end training Google's Inception CNN 130'000 images for training In: pixels and labels (biopsy results, 2'000 diseases) Out: Benign vs Malignant
- Compared to 21 dermatologists, CNN outperformed humans!

https://mashable.com/article/ibm-research-melanoma-testing-with-phone-camera

Math Conjectures (Deep Mind 2021)

- Use machine learning to make significant new discoveries in pure mathematics
 - Conjecture a new connection between the algebraic and geometric structure of knots, and a candidate algorithm predicted by the combinatorial invariance conjecture for symmetric groups
- A supervised learning model was able to detect the existence of a pattern between a large set of geometric invariants and the signature $\sigma(K)$ of a knot that was not previously known to be related to the hyperbolic geometry. <u>Nature</u> volume 600, pages70–74 (2021)

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Neural networks pre-trained on text and fine tuned on code perfectly solves university-level problems.

Drori et al (PNAS 2022)



Spin Glass Theory

- Finding the ground states of spin glasses is an essential computationally hard problem for the understanding of the nature of disordered magnetic and other physical systems. DIRAC Fan et al [2021] uses Deep Neural Networks trained purely on small-scale spin glass instances and then applied to arbitrarily large ones.
- Use methods developed in statistical physics of glassy systems to analyze numerically the training dynamics of deep neural networks.
 Baity-Jesi et al. [ICML 2018]





LEARNING TO SIMULATE COMPLEX PHYSICS WITH GRAPH NETWORKS

arXiv:2002.09405 [cs.LG]

- Graph Network-based Simulators that can learn to simulate challenging physical domains: fluids, rigid solids, deformable materials interacting with one another.
- State of a physical system represented with particles, expressed as nodes in a graph.
- Computes dynamics via learned messagepassing.
- Shown that it can generalize from singletimestep predictions with thousands of particles during training, to different initial conditions, thousands of timesteps, and at least an order of magnitude more particles at test time.



Forecast Earthquakes on Tecnonic Faults



Precursors: Consistent changes prior to failure (elastic wave speed, amplitude, fault dilation, fluid pressure/chemistry, etc.) https://www.lanl.gov/discover/news-releasearchive/2018/December/1217-machine-learning.php



In Codice Ratio – ML based transcription of the 85 km shelves of the Vatican Secret Archive

Automatically transcribe the contents of the manuscripts. Follow a novel approach, based on character segmentation.

Govern imprecise character segmentation by considering that correct segments are those that give rise to a sequence of characters that more likely compose a Latin word.

Designed a principled solution that relies on **convolutional neural networks** and **statistical language models**.

High School students will be involved in collecting and labelling training data through the Program of digital of Fondazione "I Lincei per La Scuola" e Ministero dell'Istruzione Transcription

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Output: text





http://www.inf.uniroma3.it/db/icr/

"I don't know how long it's going to be before the name of our field is changed from computer science to machine learning..."

Donald Knuth on Machine Learning and the Meaning of Life (2021)