# **Machine Learning**

## **Computer Engineering**

**Fabio Vandin** 

October 2<sup>nd</sup>, 2023

# **Machine Learning**

### 6 credits:

- 48 hours in class lectures
  - Some hours will be in a lab: more decisions to come later
- 102 hours individual study

Everything (lectures, exams, homeworks) in English!

Note: if you are enrolled in "ICT for Internet and Multimedia" or in "Control Systems Engineering", you should follow a different section!

### Course Website

Course website: from <a href="https://stem.elearning.unipd.it/">https://stem.elearning.unipd.it/</a>

Register today if not done yet!

## Lectures: When and Where

Monday 10:30-12:30pm, room Ae: How should we split the lecture?

— 10:30-12:00pm; no break ☺

Friday 10:30am-12:30pm, room Ae: How should we split the lecture?

— 10:30-12:00pm; no break ☺

## Lectures: When and Where

Lectures are not live streamed.

Material from analogous lectures from 2021/2022 are available on the course website.

See the material posted on the website to see if there are differences in the material covered.

## **Labs and Homeworks**

#### Labs:

 some (6?) lectures will be done in the lab, dates to be fixed

#### **Homeworks:**

- (up to) 3 homeworks
- will give up to 3 points for the final grade
- not compulsory but highly recommended
- will require to complete the code in Jupyter notebooks
- typical schedule:
  - day X: homework released
  - day X+14: deadline for homework submission



# **Grading**

#### Written test (**NEW FORMAT**):

- Part 1: multiple choice questions, pass/fail [30 mins]
- Part 2: questions and exercises [1.5-2 hours]
  - graded only if part 1 is passed
  - see sample tests from previous years in elearning
  - will be graded on a scale from 0 to 30L.

**Homeworks [not compulsory]**: some (3?) homeworks. Up to 3 points as a bonus on the written test grade.

**Final Grade** = grades written test + homeworks

#### **Example**

24.5 written test + 2.66 homeworks = 27 final grade

**Note**: there may be an oral exam just to confirm the grade of the written exam.

## Final Exam: dates

### 1. Tuesday, January 23<sup>rd</sup>, 2023

time: 2pm

rooms: De, Ke, Ve

### 2. Tuesday, February 13<sup>th</sup>, 2023

time: 9:30am

room: De, Ke, Ve

### 3. Tuesday, July 2<sup>nd</sup>, 2023

time: 9:30am

room: Ke

### 4. Tuesday, September 10<sup>th</sup>, 2023

time: 9:30am

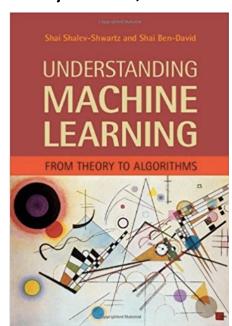
room: Le

## Material

#### **Main Book**

• [UML] Shalev-Shwartz, S. and Shai Ben-David. *Understanding machine learning: From theory to algorithms*. Cambridge University Press, 2014.

Material in class will be related to the book as much as possible



#### PDF available from the

**authors** <a href="http://www.cs.huji.ac.il/~shais/UnderstandingMachineLearning/copy.html">http://www.cs.huji.ac.il/~shais/UnderstandingMachineLearning/copy.html</a>

## Material (2)

#### **Other Books (NOT Mandatory)**

- M. Hardt, B. Recht, PATTERNS, PREDICTIONS, AND ACTIONS A story about machine learning. Princeton University Press, 2022 (available from the authors at <a href="https://mlstory.org/">https://mlstory.org/</a>).
- T. Hastie, R. Tibshirani, J. Friedman. *The Elements of Statistical Learning*. Springer, 2008.

#### PDF from authors online

- C. M. Bishop. Pattern Recognition and Machine Learning. Springer, 2006
- K.P. Murphy. *Machine Learning A Probabilistic Perspective*, MIT Press. 2012.
- Yaser S.Abu-Mostafa, M. Magdon-Ismail, H. Lin. Learning from Data. AMLBook, 2012.

Part of other books may be used: will provide handouts whenever possible...

#### **Additional material**: course website (*stem.elearning.unipd.it*)

- draft slides: provided sometime (...) before the lecture
- slides used in class: published after lectures
- links, etc.

If you are missing background notions (probability, algebra): ask me for material

# Programming and more

Language: python (https://www.python.org)

Some libraries: scikit-learn, numpy,...

### Jupyter lab:

- https://jupyterlab.readthedocs.io/en/stable/getting\_st arted/overview.html
- Allows for a mix of text and code;
- Suggestion: install it through Anaconda as suggested on Jupyter website, you get a lot of other packages/libraries for ML, visualization, etc.

https://www.anaconda.com/download/

(That's what we are going to use for homeworks, etc.)

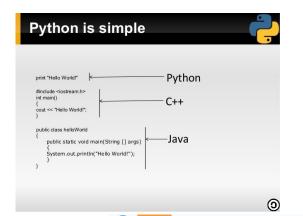
# Why Python

Install User Guide API Examples More >

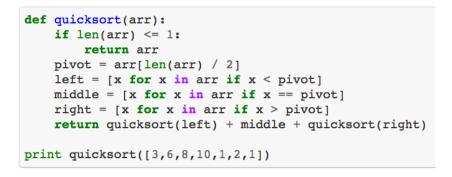
GitHub

Regression

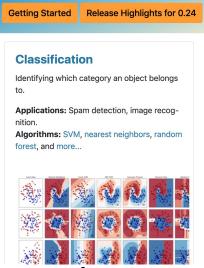
### Easy! (...)



learn



## A lot of support for ML!



scikit-learn

Machine Learning in Python

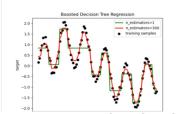
#### Accessible to everybody, and reusable in various contexts Built on NumPy, SciPy, and matplotlib

- Open source, commercially usable BSD license

Simple and efficient tools for predictive data analysis

Clustering





### Automatic grouping of similar objects into Applications: Customer segmentation, Grouping experiment outcomes Algorithms: k-Means, spectral clustering, mean-shift, and more...

Go

Used in industry, academia, research labs...

## Homework 0

- 1. Go home and install Anaconda (version with Python 3.11):
  - https://www.anaconda.com/download/
- Go through the following tutorial (it's for Python 3.7, but still useful):
  - http://cs231n.github.io/python-numpy-tutorial/
- 3. Read Chapter 1 and Chapter 2 from *Data Structures and Algorithms in Python* [Goodrich, Tamassia, Goldwasser]
  - available on course website
- 4. Get used to Jupyter lab/Jupyter notebooks
- 5. Go through the following tutorial running the Jupyter notebook and in script mode
  - https://github.com/marcc-hpc/python3tutorial/blob/master/python3-tutorial.ipynb

### Lecturer

Fabio Vandin, Professor, DEI (Department of Information Engineering)

Email: fabio.vandin@unipd.it

Website: <a href="www.dei.unipd.it/~vandinfa">www.dei.unipd.it/~vandinfa</a>

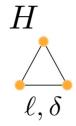
Office: 410 (4th floor, DEI/G), phone: 049-827-7946

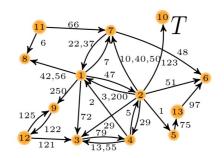
Office hours: Tuesday, 9-10am, by appointment (email before Monday at 5pm)

### Lecturer

#### **CV**

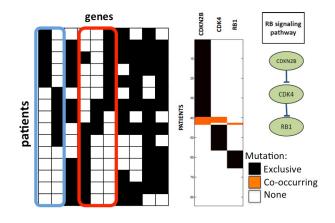
- Laurea Triennale in Computer Engineering (2004)
- Laurea Specialistica in Computer Engineering (2006)
- Ph.D. at DEI (2010)
- 2010-May/2015: Researcher/Assistant Prof.
  - Brown University (USA)
  - University of Southern Denmark
- 2015-2020: Associate Prof. at DEI
- February 2020-now: Prof. at DEI





#### **Research Interests:**

- Methods: algorithms for machine learning, data mining, and big data
- Applications: Biology, Medicine, Social Networks, ...



# **Example: Ad Click Prediction**

**Goal**: predict if a user will click on a given ad if the ad is shown to the user



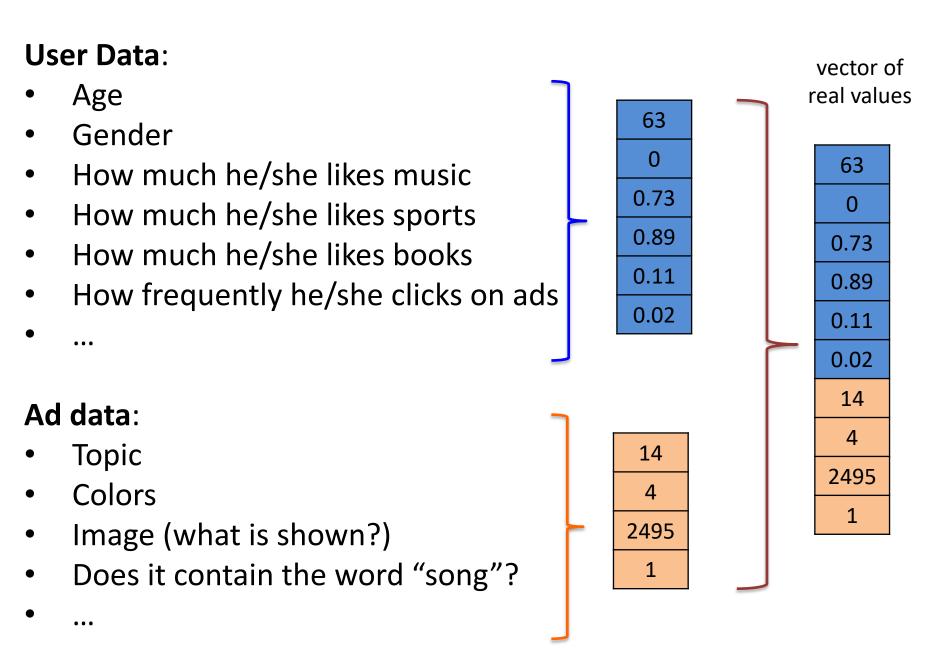
### Why is it important?

Social networks...

How would you solve the problem?

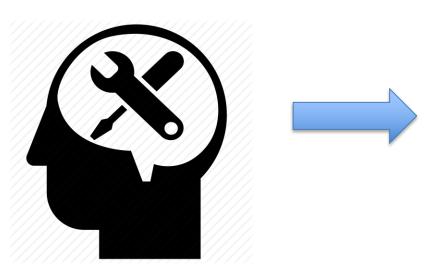
You have data about the user and about the ad

## Ad Click Prediction: Data



## A Solution that is NOT Machine Learning

Somebody (an "expert") manually builds a model/formula/algorithm that decides if the user will click on the ad or not

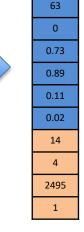


"If the age is <25
and
the gender is male
and
the ad is about soccer
then
he will click on the ad"

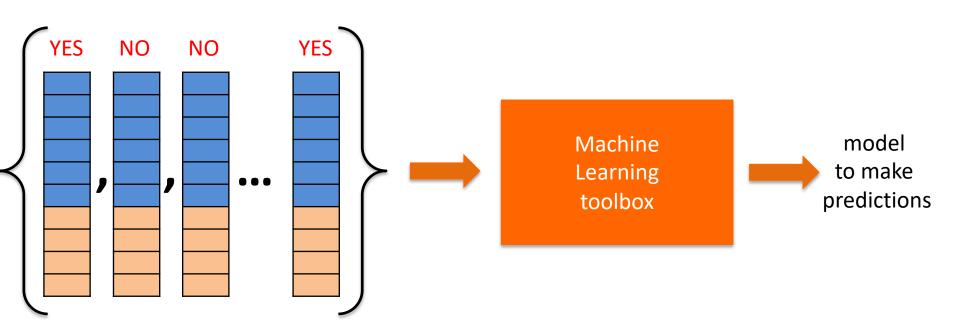
# A Machine Learning Solution

You use previous data about:

- user and ad that was displayed to user
- whether the user clicked on the ad (label: YES/NO)

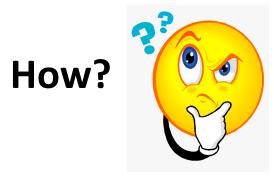


You give a lot of such pairs (vector, label) to a machine learning system which then produces in output a *model* to make the prediction



# Learning: A Difficult Problem

**Note**: coming up with a model that performs very well on the data I already have is very easy!



The difficulty is about making predictions for data we have not seen!



# Questions?

# Example

Want to predict whether a student that just graduated from Computer Engineering will have a fun job or not based on some features:

- age at graduation
- LM final grade
- Machine Learning (ML) grade
- height

See Jupyter notebook.