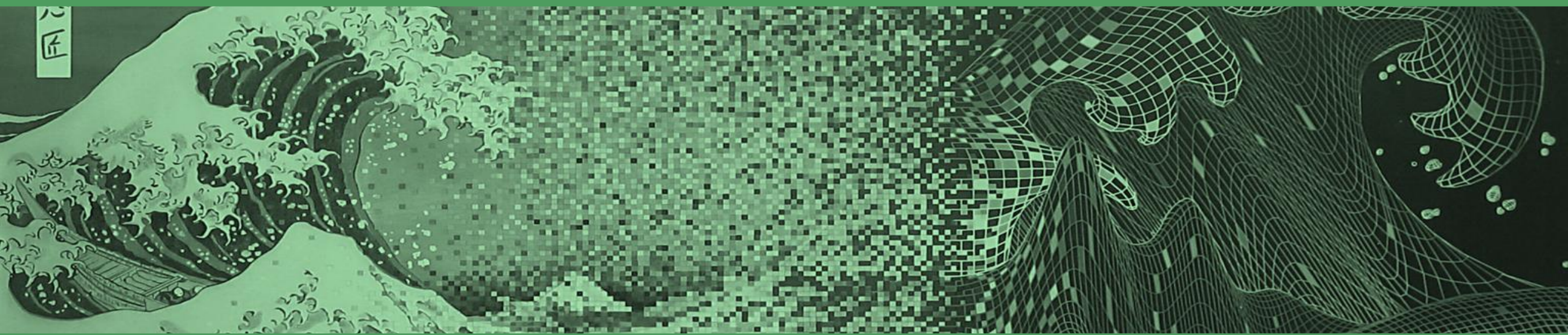
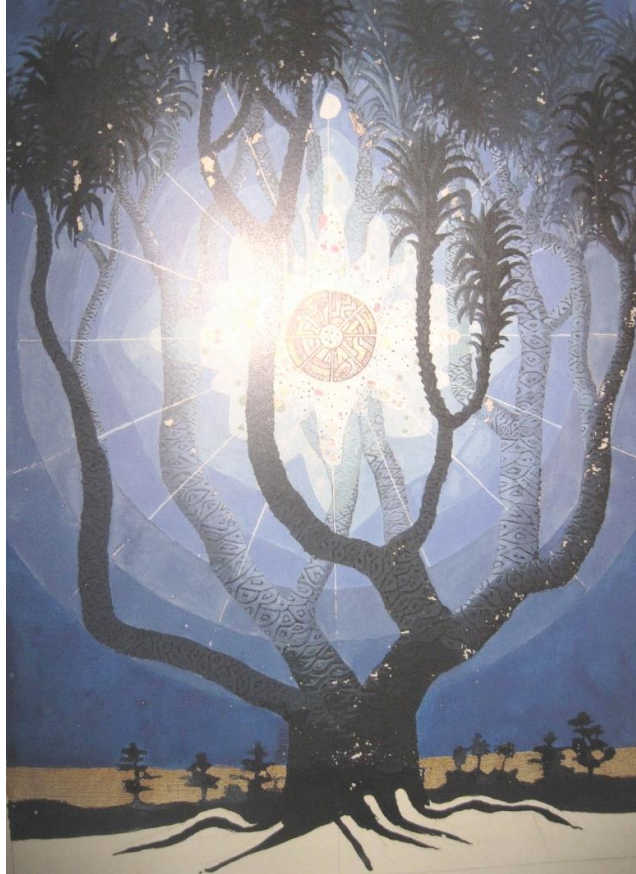


# Course organization

course objectives, assessment and plan



# Outline



- Faculty hosts
- Objectives
- Planning
- Bibliography
- Evaluation
- Homeworks
- Schedule
- Lab attendance
- Office hours

# Faculty hosts



– **Rui Henriques** (*responsible @ Alameda*)

[rmch@tecnico.ulisboa.pt](mailto:rmch@tecnico.ulisboa.pt)

Lectures [PT/EN] @ Alameda

Office: Room 433, INESC-ID, Alameda



– **Andreas (Andrzej) Wichert** (*responsible @ Tagus*)

[andreas.wichert@tecnico.ulisboa.pt](mailto:andreas.wichert@tecnico.ulisboa.pt)

Lectures [EN] @ Tagus

Office: Room N2 5-7, Taguspark



# Faculty hosts: *practical classes*



– **David Calhas**

[david.calhas@tecnico.ulisboa.pt](mailto:david.calhas@tecnico.ulisboa.pt)



– **Inês Magessi**

[ines.magessi@tecnico.ulisboa.pt](mailto:ines.magessi@tecnico.ulisboa.pt)



– **Gonçalo Correia**

[goncalo.m.correia@tecnico.ulisboa.pt](mailto:goncalo.m.correia@tecnico.ulisboa.pt)

# Faculty hosts: *practical classes*



- **Sérgio Gonçalves Pinto**  
[sergio.g.pinto@tecnico.ulisboa.pt](mailto:sergio.g.pinto@tecnico.ulisboa.pt)



- **Miguel Moreira**  
[miguelmatamoreira@tecnico.ulisboa.pt](mailto:miguelmatamoreira@tecnico.ulisboa.pt)



- **Guilherme Varela**  
[guilherme.varela@tecnico.ulisboa.pt](mailto:guilherme.varela@tecnico.ulisboa.pt)



- **Javier de Muller Santa-Maria**  
[javier.de.muller@tecnico.ulisboa.pt](mailto:javier.de.muller@tecnico.ulisboa.pt)

# Objectives




- Answer real-world data-rich problems: **formalize** and **translate them into ML tasks**
- Understand **major topics on ML**
  - master the **statistical** and mathematical **foundations** behind ML
  - critically compare the behavior of ML approaches
    - **Bayesian learning**
    - **lazy** and **associative learning**
    - **neural network learning**
- Master **supervised ML** approaches, including classification and regression
- Master **unsupervised ML** approaches, including clustering and data transformations
- Robustly **evaluate ML solutions** (loss, statistical significance, generalization ability)

# Planning

1. <b>Introduction to ML</b> 2. <b>Univariate ML</b> 3. <b>Associative learning</b> 4. <b>Evaluation</b>	<b>I</b>
5. <b>Bayesian learning</b> 6. <b>Lazy learning</b> 7. <b>Kernel learning</b> 8. <b>Regression learning</b>	<b>II</b>
9. <b>Gradient descent</b> 10. <b>Neural network learning</b> 11. <b>Deep Learning</b>	<b>III</b>
12. <b>Clustering</b> 13. <b>Dimensionality reduction</b>	<b>IV</b>

## Theoretical background

- Information theory (2,3)
- Probability theory (5,12)
- Algebra theory (6,7,13)
- Optimization theory (8-11)

-  *Foundations*
-  *Supervised learning*
-  *Unsupervised learning*

# Practical classes

- **L1** Univariate statistics, evaluation
- **L2** Associative learning (decision trees)

⇒ **HW1**

- **L3** Bayesian learning
- **L4** Lazy learning
- **L5** Linear regression

⇒ **HW2**

- **L6/7** Gradient descent learning
- **L8/9** Neural network learning

⇒ **HW3**

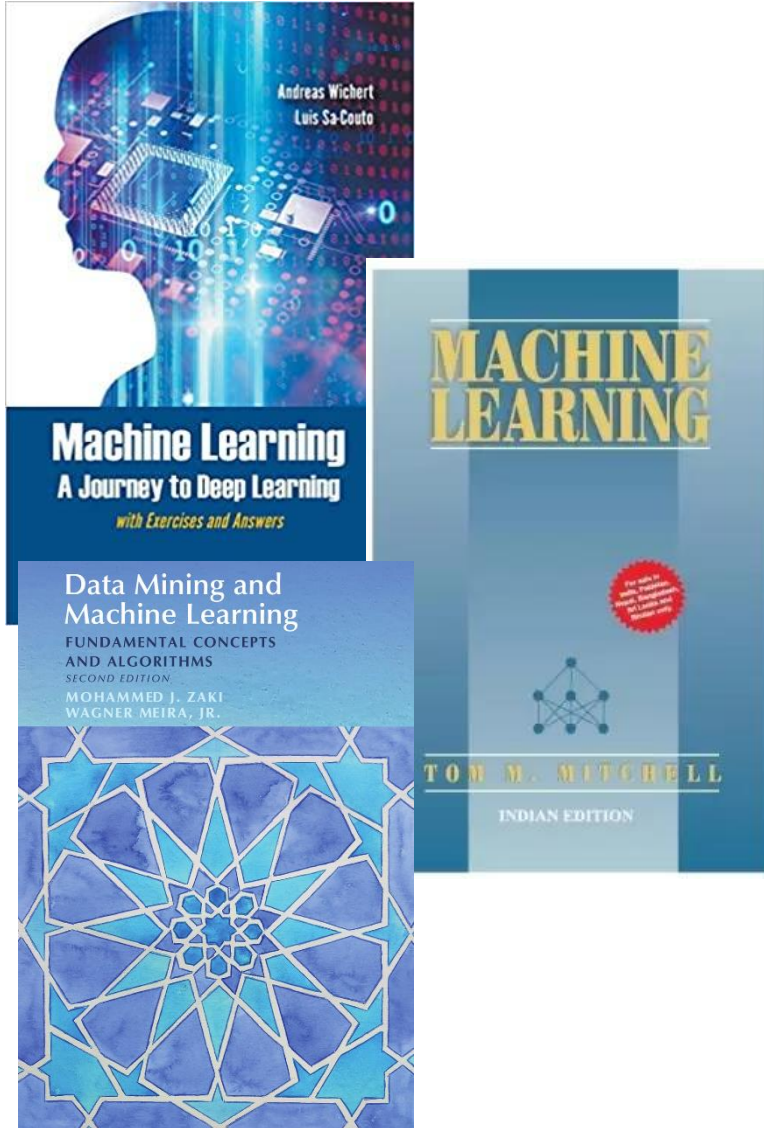
- **L10** Clustering
- **L11** Dimensionality reduction

⇒ **HW4**





# Main literature



- reference slides @ course's webpage
- **Machine Learning: A Journey to Deep Learning**  
Andreas Wichert and Luis Sá Couto, World Scientific, 2021  
<https://www.worldscientific.com/worldscibooks/10.1142/12201>
- **Machine Learning**  
Tom Mitchell, McGraw Hill, 1997  
<http://www.cs.cmu.edu/~tom/mlbook.html>
- **Data Mining and Machine Learning: Fundamental Concepts and Algorithms**  
Mohammed J. Zaki and Wagner Meira Jr, Cambridge Univ. Press, 2<sup>nd</sup> Ed., 2020  
<https://dataminingbook.info/>

# Secondary literature

- **Pattern Recognition and Machine Learning**

Christopher M. Bishop, Springer 2006

<https://www.microsoft.com/en-us/research/people/cmbishop/#!prml-book>

- **Neural Networks and Learning Machines**

Simon O. Haykin, (3rd Edition), Pearson 2008

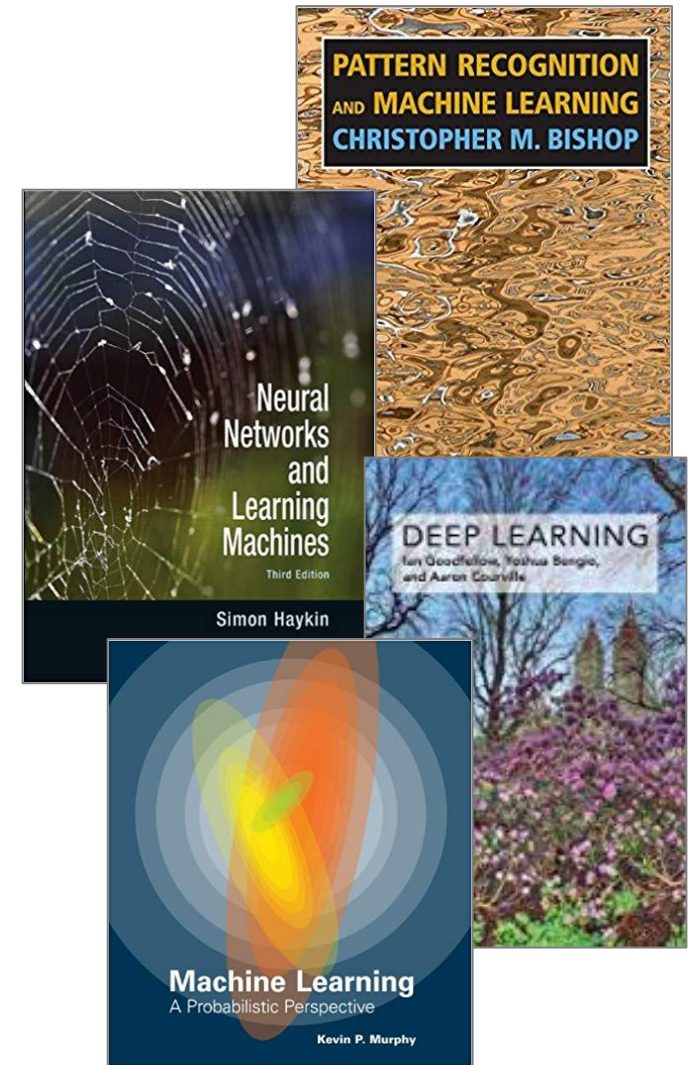
- **Deep Learning**

I. Goodfellow, Y. Bengio, A. Courville, MIT Press 2016

<https://www.deeplearningbook.org>

- **Machine Learning: A Probabilistic Perspective**

K. Murphy, MIT Press 2012



# Tertiary literature

- *Software literature*

- **Hands-On Machine Learning with Scikit-Learn and TensorFlow: Concepts, Tools, and Techniques to Build Intelligent Systems**

Aurélien Géron , O'Reilly Media, 2017

<https://github.com/amitanalyste/aurelienGeron>

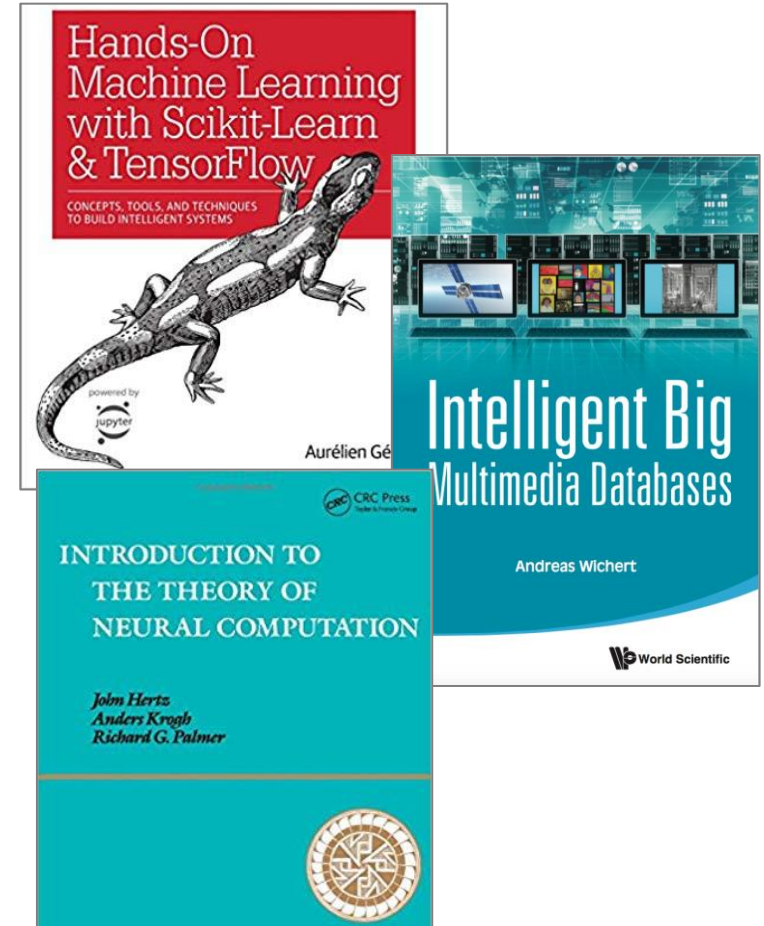
- *Complementary*

- **Intelligent Big Multimedia Databases**

A. Wichert, World Scientific, 2015

- **Introduction to the theory Of Neural Computation**

J.A. Hertz, A.S. Krogh, R.G. Palmer, Addison-Wesley, 1991



# Evaluation



**Final grade:**  $\text{Homeworks} \times 50\% + \text{Exam} \times 50\%$

**Homeworks:**

- $\text{average}(\text{HW1}, \text{HW2}, \text{HW3}, \text{HW4})$
- groups of **2** (no exceptions)

**Exam:**  $\text{maximum}(\text{Exam1}, \text{Exam2})$

**Minimum exam grade:** 8v

**Working students**

- $(\text{Homeworks} \times 50\% + \text{Exam} \times 50\%)$  **or** 100% Exam
- decision communicated until September 22<sup>nd</sup>

**Special season:** 100% Exam

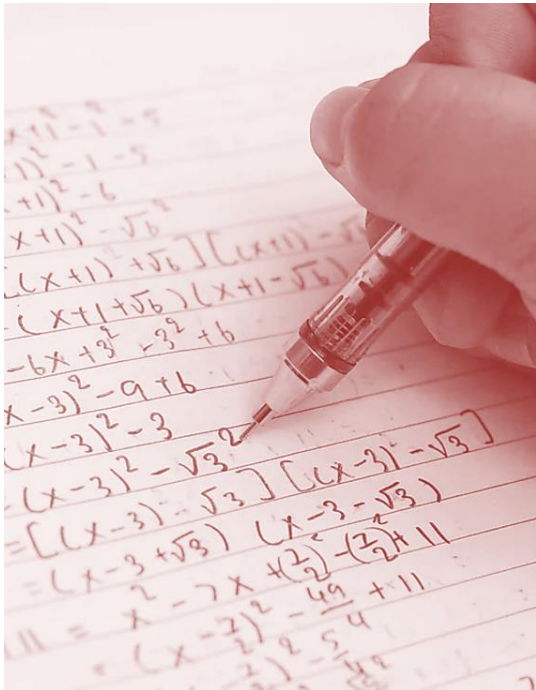
# Deadlines

- **Homework deliveries**
  - HW1: Friday, September 29<sup>th</sup> 23h59
  - HW2: Monday, October 9<sup>th</sup> 23h59
  - HW3: Friday, October 20<sup>th</sup> 23h59
  - HW4: Monday, October 30<sup>th</sup> 23h59
  - *statements* released 10 days before delivery
- **Exams**
  - Exam1: November 6, 13h
  - Exam2: February 1, 10h30





# Homeworks



- groups of **2** (*no exceptions*)
- statements at the course webpage two weeks prior delivery
- always consult webpage **FAQ** before posting questions
- three major **components**
  - **practical** pen-and-paper **quest**
  - small **computational quest**
  - **critical thinking**
    - excelling answers are clear, precise and succinct, and establish scientific hypothesis grounded on empirical evidence



# Final exam

- **extensive** number of **contents** to prepare
  - requiring a good mastery: *critical analysis* and *practical calculus*
- comprehensive exam with a **2h duration**
- you are allowed to bring a **single A4 page with notes** (e.g. formulas)
- carefully follow theoretical classes to minimize risks
- objective assessment: quantitative result answers, true-or-false



# Schedule @ Alameda

Horas	Segunda		Terça					Quarta	Quinta				Sexta		
8:00	T1 VA4														
8:30			T1 VA4												
9:00			VA4	L7	L6	L3	L5			L8	L5		L10	L11	L12
9:30															
10:00	T2 VA4									T2 FA2					
10:30															
11:00				L10	L11		L9			T1 FA1					
11:30			T2 VA3			L8									
12:00															
12:30		L4		L12							L3	L9	L4	L7	L6
13:00															
13:30															
14:00															

*provisory schedule*

# Schedule @ Tagus

Horas	Segunda	Terça	Quarta	Quinta	Sexta
8:00					
8:30					
9:00		L9	L5	T A1	
9:30					
10:00	L5	L8	L4	L9	
10:30					
11:00	T A4	L4		L8	
11:30					
12:00					
12:30					
13:00					
13:30					
14:00					

*provisory schedule*

# Office hours

- Explore your questions during **practical classes** and **lectures**
- **Check webpage for latest information on office hours**
  - *Rui Henriques* office hours
    - Monday 12h, Tuesday 9h30, Thursday 12h (end of lectures)
    - additional office hours available via appointment
  - office hours of other faculty hosts on the webpage
- **Golden rule**
  - general e-mail contacts with subject preceded by **[Aprendizagem]**
  - homework-related questions preceded by **[HW Aprendizagem]**



# Thank You



[rmch@tecnico.ulisboa.pt](mailto:rmch@tecnico.ulisboa.pt)