

# Machine Learning

## Section 1: Introduction

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11. October 2021

# **Admin**

# Admin stuff (1)

Lecture Stefan Harmeling

Mon 10:30 - 12:00, 25.11 HS 5A

Wed 12:30 - 14:00, 22.01 HS 5D

Feedback after every lecture in ILIAS (for about a day)

Slides Sciebo (also for the exercise sheets and submission)

Video <http://mediathek.hhu.de>

Übungen Tobias Uelwer, Maike Behrendt, Stefan Wagner, Liz

Leutner, Christopher Orlowicz, Sebastian Konietzny

(Mon 18:30 - 20:00, HS 5G) enfällt

Thu 12:30 - 14:00, 22.01 HS 2C

Fri 12:30 - 14:00, 22.21 HS 2E

Q&A session Fri 10:30-12:00 via Zoom

Chat RocketChat for announcements / discussion

ILIAS for all links (and the feedback form)

## Admin stuff (2)

Communication please use RocketChat!

Exercises weekly on sciebo, two students hand in one submission,  
please follow the instructions!

First sheet available this Wed, deadline next Tue, solutions next Thu  
and Fri in the “Übungen”

Exam Wed 9.2.2022 11:30-14:00 in 3D and 5F (maybe)

Admission **NEW** to be admitted to the written exam you must reach  
at least 50% on **each** exercise sheet (with 2 jokers)

Grade final grade is the grade of the exam

# Some Literature

Free online:

- ▶ **David J C MacKay** Cambridge, 2003  
*Information Theory, Inference, and Learning Algorithms*  
<http://www.inference.phy.cam.ac.uk/itprnn/book.pdf>
- ▶ **David Barber** Cambridge, 2012  
*Bayesian Reasoning and Machine Learning*  
<http://web4.cs.ucl.ac.uk/staff/D.Barber/textbook/270212.pdf>
- ▶ **Carl E Rasmussen & Christopher K I Williams** MIT, 2006  
*Gaussian Processes for Machine Learning*  
<http://www.gaussianprocess.org/gpml/chapters/RW.pdf>
- ▶ **Kevin Murphy** MIT, 2012  
*Machine Learning: a Probabilistic Perspective* three nice books on:  
<https://probml.github.io/pml-book/>

Non-free:

- ▶ **Judea Pearl** Morgan Kaufmann, 1988  
*Probabilistic Reasoning in Intelligent Systems*
- ▶ **Christopher Bishop** Springer, 2007  
*Pattern Recognition and Machine Learning*
- ▶ **Bernhard Schölkopf & Alexander J Smola** MIT, 2001  
*Learning with Kernels*
- ▶ **Edwin T Jaynes & G Larry Bretthorst** Cambridge, 2003  
*Probability Theory – the Logic of Science*

Please ask your questions on rocketchat!

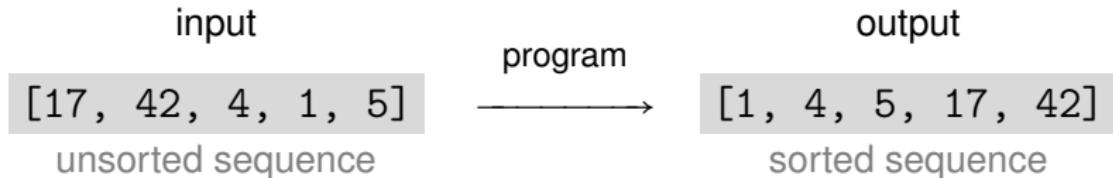


and you can leave feedback on ILIAS.

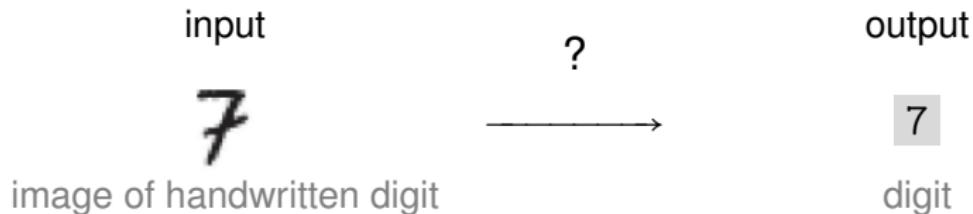
# What is Machine Learning?

# Computer programming

## Simple example



## Hard example



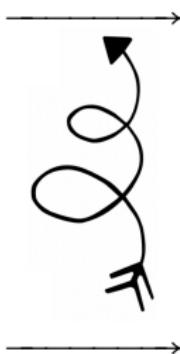
# What is machine learning

Difficult example

input

7

program



output

7

Input/output examples

|   |   |   |   |   |   |   |   |   |   |
|---|---|---|---|---|---|---|---|---|---|
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |

image of handwritten digit

|   |   |   |   |   |   |   |   |   |   |
|---|---|---|---|---|---|---|---|---|---|
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |

digit

# Lots of data — lots of questions

*We are drowning in information and starving in knowing.*

*John Naisbitt*

- ▶ 40 billion web pages
- ▶ 100 hours of video uploaded every minute on youtube
- ▶ 1000s of genomes sequences (each having about  $3.8 \times 10^9$  base pairs)
- ▶ bigger telescopes generate each night terabytes of data
- ▶ ...

A great situation for Machine Learning!

## Small demo: Iris flower data set



Radomil Binek, [Kosaciec szczecinkowaty Iris setosa](#),  
cropping by SH, CC BY-SA 4.0



D. Gordon E. Robertson, [Blue Flag, Ottawa](#), cropping  
by SH, CC BY-SA 4.0



Eric Hunt, [Iris virginica 2](#), cropping by SH, CC BY-SA  
4.0

| <i>Iris setosa</i> |             |              |             | <i>Iris versicolor</i> |             |              |             | <i>Iris virginica</i> |             |              |             |
|--------------------|-------------|--------------|-------------|------------------------|-------------|--------------|-------------|-----------------------|-------------|--------------|-------------|
| Sepal length       | Sepal width | Petal length | Petal width | Sepal length           | Sepal width | Petal length | Petal width | Sepal length          | Sepal width | Petal length | Petal width |
| 5.1                | 3.5         | 1.4          | 0.2         | 7.0                    | 3.2         | 4.7          | 1.4         | 6.3                   | 3.3         | 6.0          | 2.5         |
| 4.9                | 3.0         | 1.4          | 0.2         | 6.4                    | 3.2         | 4.5          | 1.5         | 5.8                   | 2.7         | 5.1          | 1.9         |
| 4.7                | 3.2         | 1.3          | 0.2         | 6.9                    | 3.1         | 4.9          | 1.5         | 7.1                   | 3.0         | 5.9          | 2.1         |
| 4.6                | 3.1         | 1.5          | 0.2         | 5.5                    | 2.3         | 4.0          | 1.3         | 6.3                   | 2.9         | 5.6          | 1.8         |
| 5.0                | 3.6         | 1.4          | 0.2         | 6.5                    | 2.8         | 4.6          | 1.5         | 6.5                   | 3.0         | 5.8          | 2.2         |
| 5.4                | 3.9         | 1.7          | 0.4         | 5.7                    | 2.8         | 4.5          | 1.3         | 7.6                   | 3.0         | 6.6          | 2.1         |
| 4.6                | 3.4         | 1.4          | 0.3         | 6.3                    | 3.3         | 4.7          | 1.6         | 4.9                   | 2.5         | 4.5          | 1.7         |
| 5.0                | 3.4         | 1.5          | 0.2         | 4.9                    | 2.4         | 3.3          | 1.0         | 7.3                   | 2.9         | 6.3          | 1.8         |
| 4.4                | 2.9         | 1.4          | 0.2         | 6.6                    | 2.9         | 4.6          | 1.3         | 6.7                   | 2.5         | 5.8          | 1.8         |
| 4.9                | 3.1         | 1.5          | 0.1         | 5.2                    | 2.7         | 3.9          | 1.4         | 7.2                   | 3.6         | 6.1          | 2.5         |
| 5.4                | 3.7         | 1.5          | 0.2         | 5.0                    | 2.0         | 3.5          | 1.0         | 6.5                   | 3.2         | 5.1          | 2.0         |
| 4.8                | 3.4         | 1.6          | 0.2         | 5.9                    | 3.0         | 4.2          | 1.5         | 6.4                   | 2.7         | 5.3          | 1.9         |
| 4.8                | 3.0         | 1.4          | 0.1         | 6.0                    | 2.2         | 4.0          | 1.0         | 6.8                   | 3.0         | 5.5          | 2.1         |
| 4.3                | 3.0         | 1.1          | 0.1         | 6.1                    | 2.9         | 4.7          | 1.4         | 5.7                   | 2.5         | 5.0          | 2.0         |
| 5.8                | 4.0         | 1.9          | 0.2         | 5.8                    | 2.0         | 2.6          | 1.2         | 5.8                   | 2.8         | 5.1          | 2.4         |

## Switch to Jupyter notebook

# Types of machine learning

## Supervised learning:

- ▶ discrete output: classification
- ▶ continuous output: regression
- ▶ ..., e.g. graphs or other structures

## Unsupervised learning:

- ▶ discrete output: clustering
- ▶ continuous output: dimensionality reduction

## Semi-supervised learning:

- ▶ ...

## Reinforcement learning:

- ▶ (different setup)...

# Tools of machine learning

- ▶ probability theory (e.g. Bayes' rule)

$$p(B|A) = \frac{p(A|B)p(B)}{p(A)}$$

- ▶ linear algebra (e.g. vectors and matrices)

$$y = Ax + n$$

- ▶ optimization (e.g. gradient descent)

$$\begin{aligned} & \text{minimize } f(x) \\ & \text{subject to } g(x) \leq 0 \\ & \quad h(x) = 0 \end{aligned}$$

- ▶ computer science

all the things you learned so far and more!

# Tentative syllabus

|            |    |  |    |
|------------|----|--|----|
| 2021-10-11 | 01 | Introduction and probability as logic                | 01 |
| 2021-10-13 | 02 | More on probabilities, Bayes nets                    |    |
| 2021-10-18 | 03 | Continuous probabilities                             | 02 |
| 2021-10-20 | 04 | The Gaussian distribution                            |    |
| 2021-10-25 | 05 | More on distributions, models, MAP, MLE              | 03 |
| 2021-10-27 | 06 | Linear regression                                    |    |
| 2021-11-03 | 07 | Matrix differential calculus                         | 04 |
| 2021-11-08 | 08 | SVM and constrained optimization                     |    |
| 2021-11-10 | 09 | Kernel trick   | 05 |
| 2021-11-15 | 10 | More on kernels                                      |    |
| 2021-11-17 | 11 | Dimensionality reduction and PCA                     | 06 |
| 2021-11-22 | 12 | ISOMAP, MDS, LLE and MVU                             |    |
| 2021-11-24 | 13 | kPCA, Intro to Gaussian mixture models               | 07 |
| 2021-11-29 | 14 | Gaussian mixture models, k-means, EM                 |    |
| 2021-12-01 | 15 | General EM, missing data, latent variables           | 08 |
| 2021-12-06 | 16 | Neural networks: intro                               |    |
| 2021-12-08 | 17 | Neural networks: backprop                            | 09 |
| 2021-12-13 | 18 | Neural networks: building blocks                     |    |
| 2021-12-15 | 19 | Neural networks: tutorials about Pytorch             | 10 |
| 2021-12-20 | 20 | Neural networks: tricks, unsupervised                |    |
| 2021-12-22 | 21 | Gaussian processes theory                            | 11 |
| 2021-01-10 | 22 | Gaussian processes / kernels                         |    |
| 2022-01-12 | 23 | Gaussian processes continued again                   | 12 |
| 2022-01-17 | 24 | Gaussian processes / GPC                             |    |
| 2022-01-19 | 25 | Sampling   | 13 |
| 2022-01-24 | 26 | Statistical learning theory                          |    |
| 2022-01-26 | 27 | MCMC   |    |
| 2022-01-31 | 28 | No-free-lunch theorem, biased overview and questions |    |
| 2022-02-02 | 29 | maybe AlphaGo and AlphaGoZero                        |    |
| 2022-02-09 | 30 | WRITTEN EXAM   |    |

# Summary

## What is machine learning?

- ▶ Supervised learning: classification, regression
- ▶ Unsupervised learning: regression, dimensionality reduction
- ▶ ...