

Machine Learning – Introduction

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Known under many names:

- Machine Learning
- Artificial Intelligence
- Pattern Recognition
- Data Mining
- Data Assimilation
- Big Data

Examples:

- Spam Filtering
- Optical Character Recognition (OCR)
- Search Engines
- Computer Vision
- Natural Language Processing (NLP)
- Advertising
- Fraud Detection
- Robotics
- Data Prediction
- Material Discovery
- Astronomy

How do we learn?

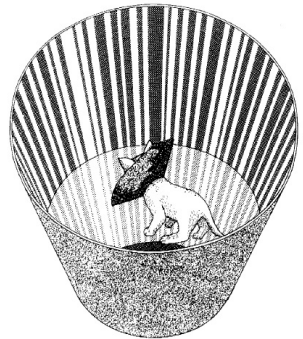
Sensory stimulation

- Holy Roman Emperor Frederic II (13th century),
- spoke six languages himself,
- desired to determine the "god given" language.
- Allegedly raised newborns without human interactions,
- physical needs satisfied.
- All children died.

⇒ Without experience, no learning.

Vision

- Sir Colin Blakemore, Grahame Cooper (1970),
- kittens brought up in an environment with either only horizontal or only vertical lines.
- Horizontal kittens show no reaction to vertical lines.
- Vertical kittens show no reaction to horizontal lines.
- No brain activity,
- blind to certain lines.



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Vision

- Only what the environment presents is learnt.
- Efficient preparation for the future.
- Also true for human vision.
- Australian aborigines have the sharpest vision ever measured,
- about four times better than those of white ethnicity.
- Ophthalmologist Professor Fred Hollows corrected the vision of an elderly Aboriginal man back to the average white person's vision with glasses. The reaction was "*Thank you for trying, but this is hopeless. I used to be able to see much better.*"
- Wide open landscapes are their environment,
- good vision in the distance is vital for survival.

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- What should be measured?
- And how should this then be interpreted?

Learning versus teaching

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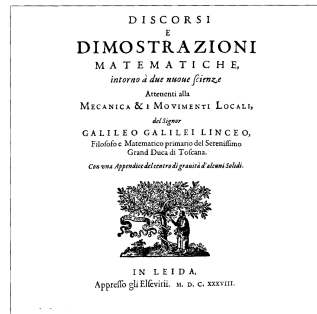
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- Instructions are given.
- Procedural programming.
- Object oriented programming makes instructions dependent on the nature of the data.

- Learning is much more fun when we make a discovery ourselves.
- Teacher led learning.
- Beautifully illustrated by Galileo.
- The inclined plane experiment is repeated by school children all over the world again and again.



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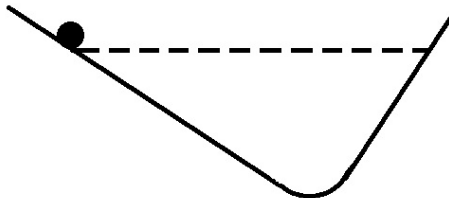
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- Linear relationship between a primary parameter and a secondary parameter \Rightarrow kernel trick.

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- *Feature detection*: Which parameters influence the outcome?
- Weight does not influence the experiment, however the inclination does.
- And last, but most importantly, Galileo was an *unsupervised learner*.



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- Teaching: completely governed by external input.

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