

Visions to Products

# What is Artificial intelligence, Machine Learning and Deep Learning?

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### This Session:

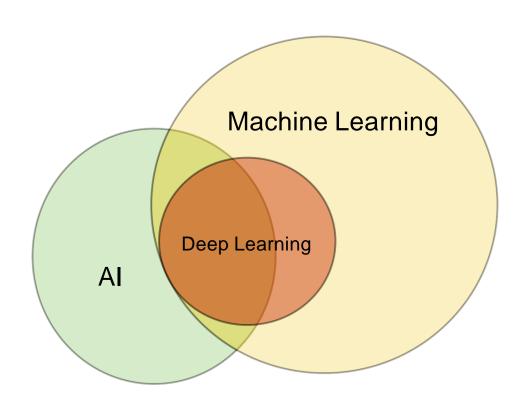


- Introduction
- What is Machine Learning?
- Examples
- How Machine Learning works?
- Machine Learning concepts



### **The Connection Between Fields**





### **The Connection Between Fields**



### **Artificial Intelligence (AI):**

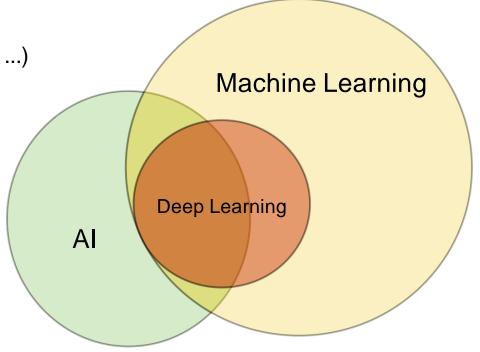
orig. subfield of computer science, solving tasks humans are good at (natural language, speech, image recognition, ...)

### **Artificial General Intelligence (AGI):**

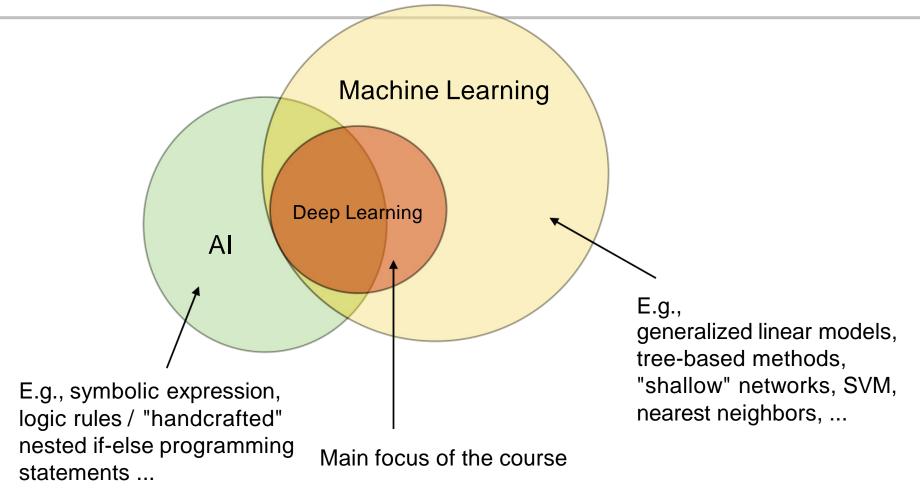
multi-purpose Al mimicking human intelligence across tasks

#### Narrow Al:

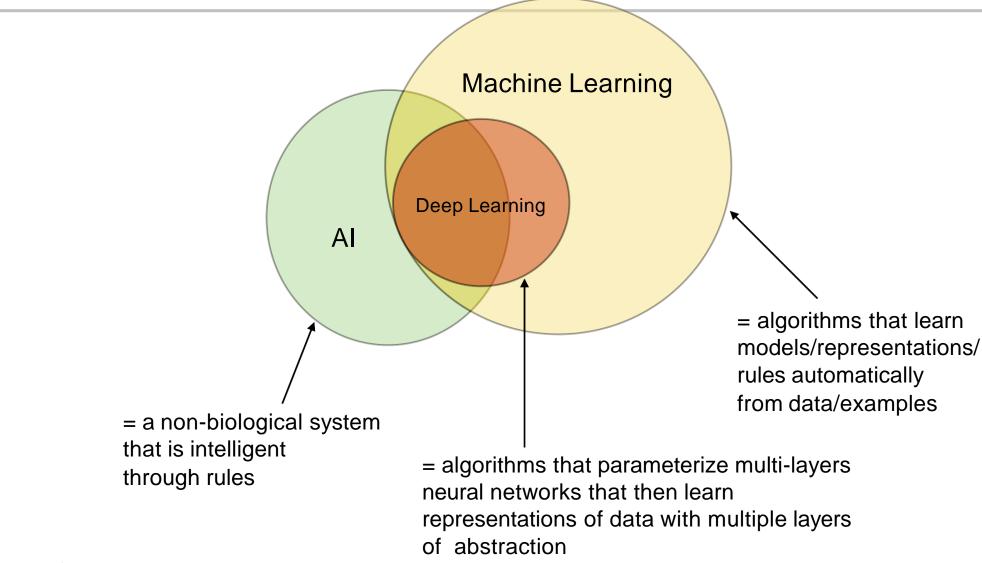
solving "a" task (playing a game, driving a car, ...)











# What's Machine Learning?

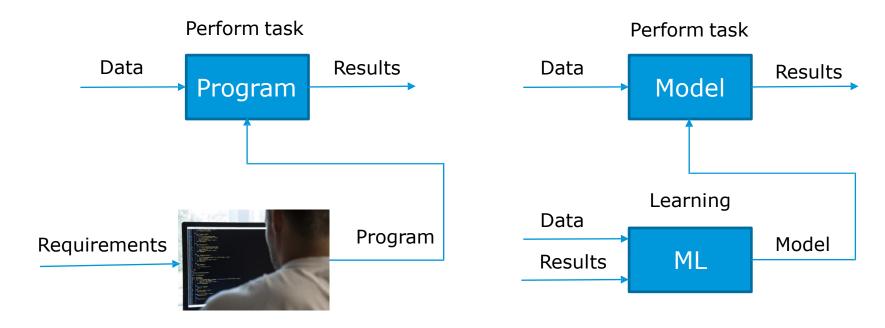


• The science (and art) of programming computers so that they can learn from data [Aurélien Géron, 2017]

- "The field of study that gives computers the ability to learn without being explicitly programmed" [Artur Samuel, 1959]
- "A computer program is said to learn from experience E with respect to some task T and some performance measure P, if its performance on T, as measured by P, improves with experience E" [Tom Mitchell, 1997]

# How is Machine Learning different from programming





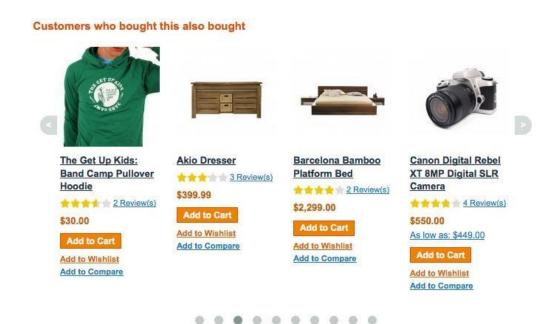
Only for some tasks



Let's see some examples

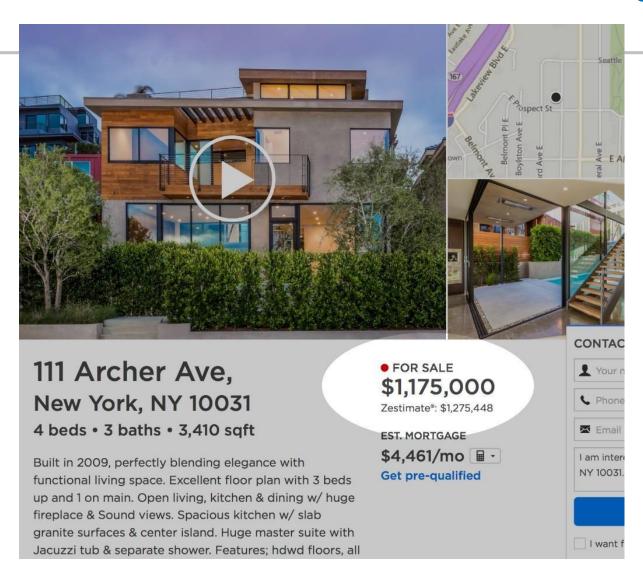


Spam Filter



Recommendation Engine

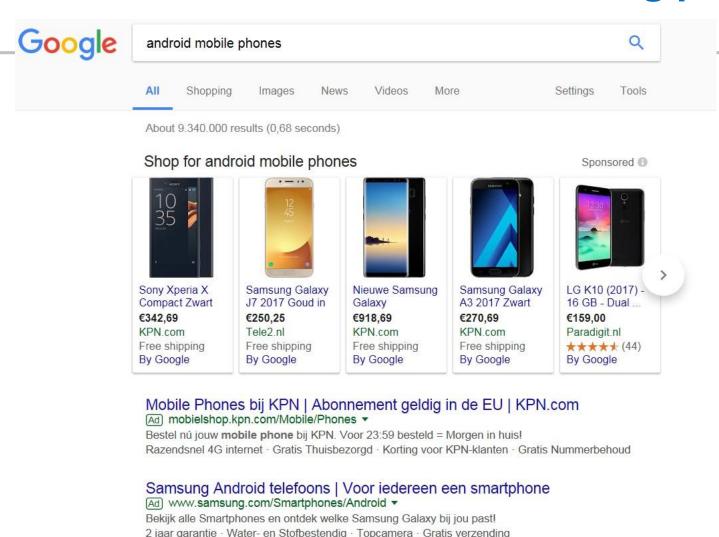




Real Estate

Predict at which price a property will be sold

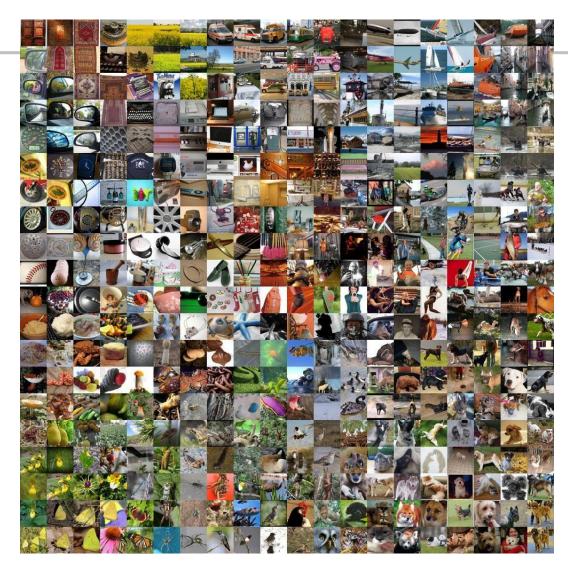


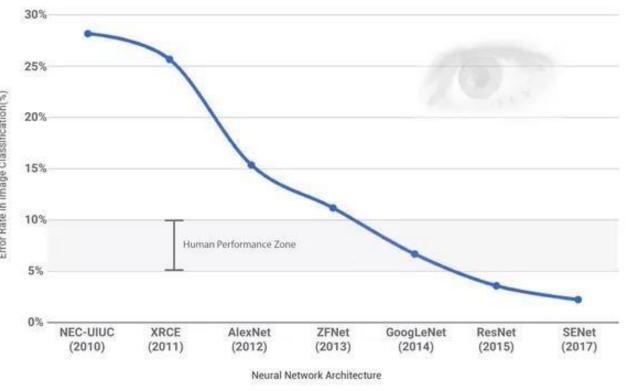


### Advertising

Predict which ads you are more likely to click on







### **Image Classification**





Self-driving cars





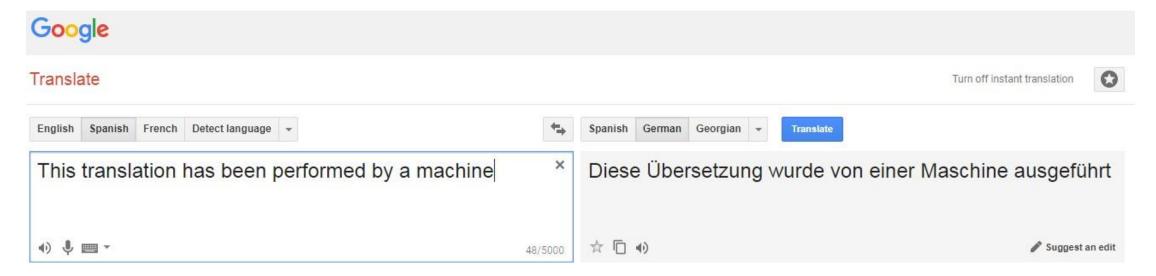




Speech Recognition & Synthesis



### Language Translation







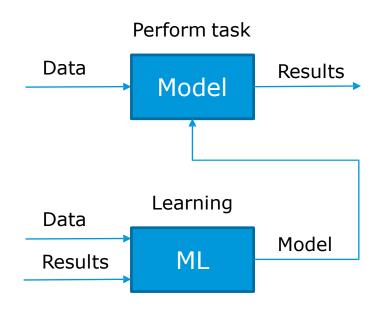


### Playing Games



# **Considerations on using Machine Learning**

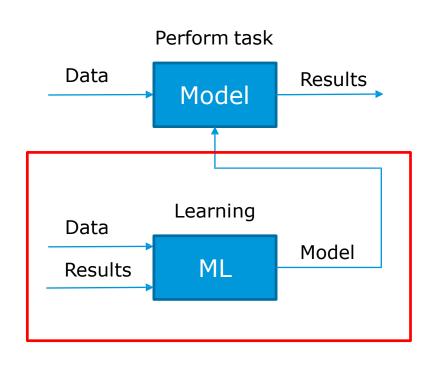




- There must be a pattern in the input output relationship (lottery winning numbers cannot be predicted with ML)
- There must be enough data to discover this pattern
- It's difficult to formulate a mathematical expression (otherwise we will just use this formula instead)

### **How Machine Learning Learns?**

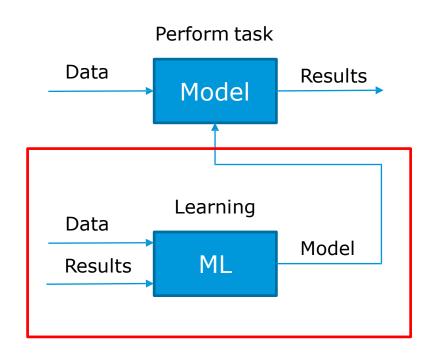




- Different Machine Learning techniques for different kinds of tasks
- Learning is finding which model's parameters represent best the input – output mapping

### **How Machine Learning Learns?**

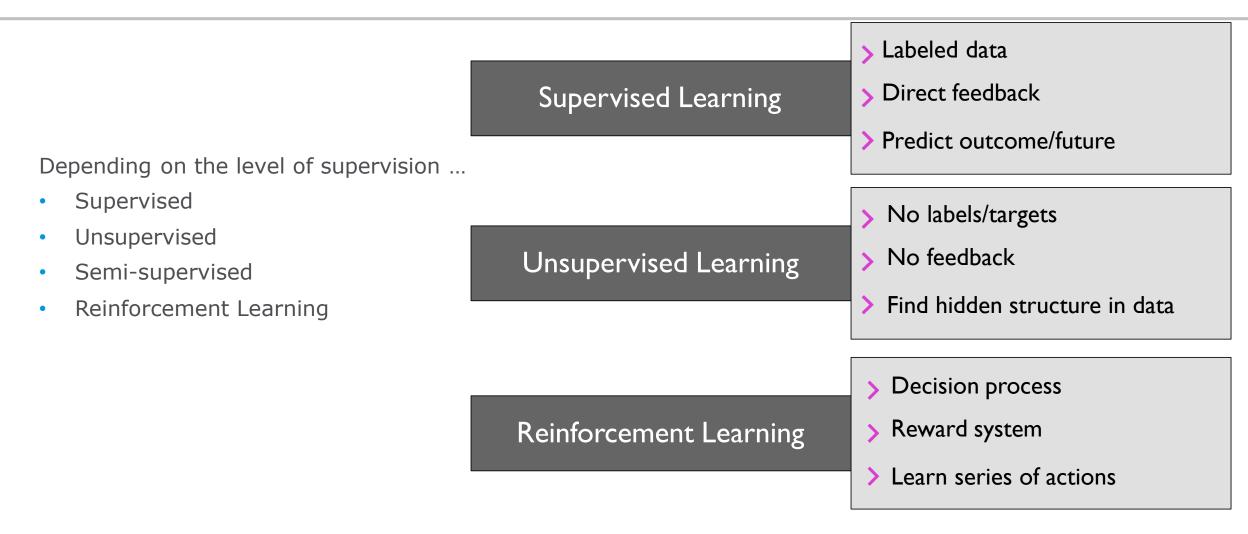




### Linear Regression example:

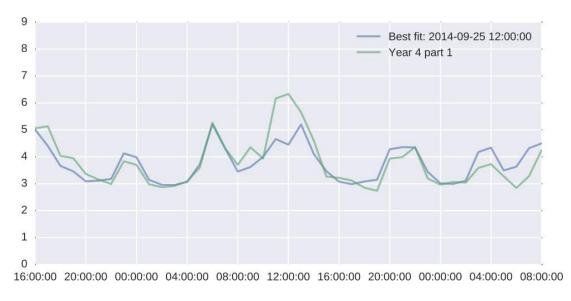
- Model: f(x) = mx + b
- Model's parameters: m, b
- Parameter values: m=1, b=0
- Learning is finding which values of 'm' and 'b' fit the data best (e.g. minimizes the prediction error)







- Supervised
  - Supervision: we can tell for every case what the correct answer was
  - Example: predict the thermal power consumption



# **Focus**Predict the future



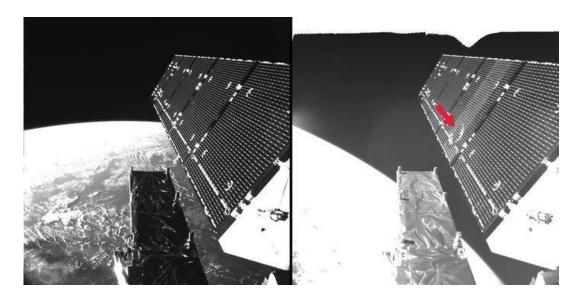
- Unsupervised
  - Supervision: there is no right answer, we are looking for insights
  - Example: market basket analysis for supermarkets



# **Focus**Understand the past



- Semi-supervised
  - Supervision: we can tell the correct output for a limited number of cases
  - Example: characterize what a particle impact looks like in TM



### **Focus**

Understand the past

Sentinel-1A: particle impact on August 23th 2016



- Reinforcement Learning
  - Supervision: we only know the final outcome, but not intermediate steps
  - Example: playing Go



### **Focus**

Find which is the next action most likely to lead to the desired outcome



The type of learning with most industrial applications is

### **Supervised Learning**

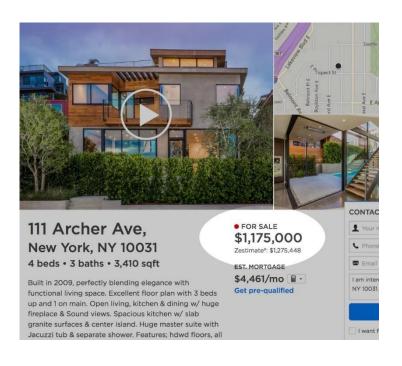
(Predictive Analytics)

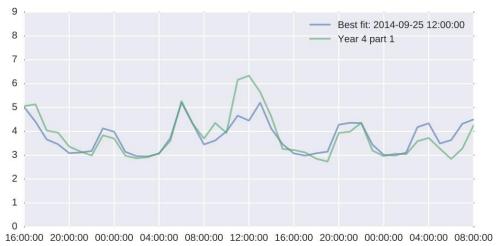
- Depending of what kind of data is predicted we can talk about:
  - Regression
  - Classification

# Regression



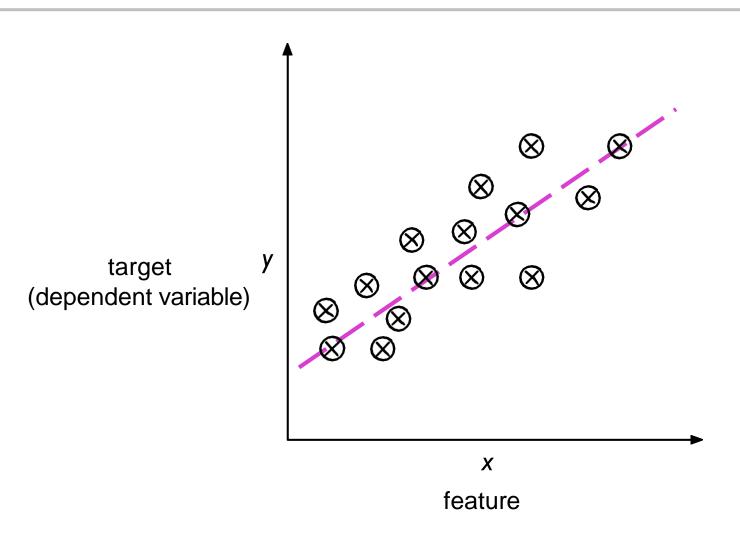
### Predict real numbers





# Regression







Artificial Intelligence

$$\hat{y} = \boldsymbol{w}^{\top} \boldsymbol{x} + b$$

### Classification



Predict which option out of a limited set of possibilities



Spam Filter

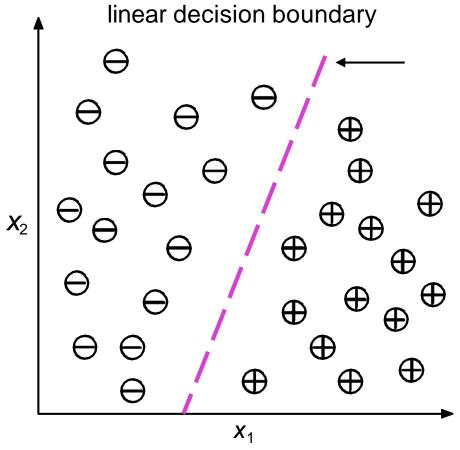


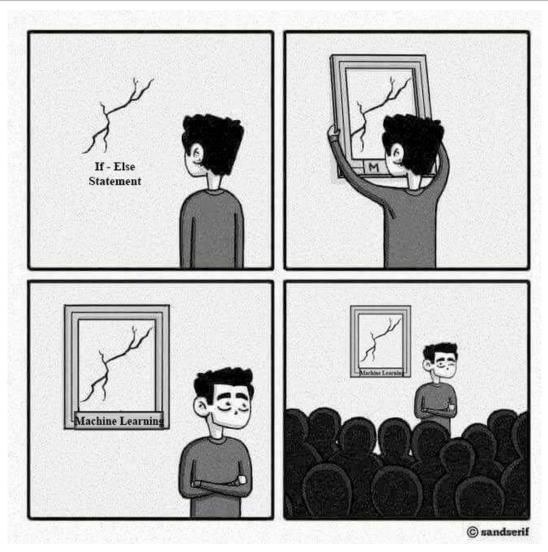
What's in the picture

### Classification



Binary classification example with two *features* ("independent" variables, predictors)

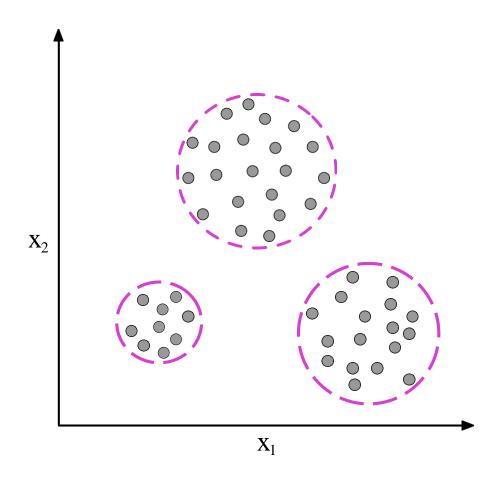




# **Unsupervised Learning: Clustering**



Assigning group memberships to unlabelled examples (instances, data points)

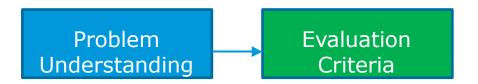




Problem Understanding If I had an hour
to solve a problem, I'd spend 55
minutes thinking about the problem
and five minutes thinking
about solutions!

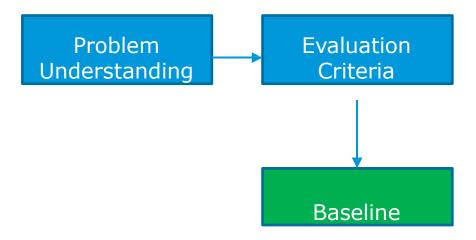


- How will we measure how good the model is performing?
- Do we know already at what point it would be enough?



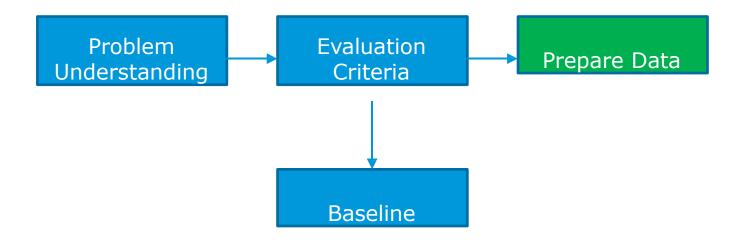


- How the current approach is performing against the evaluation criteria? Define a simple baseline if there is none (e.g. mean value)
- This will allow us to quantify how much Machine Learning helps and if it is worthwhile compared to simpler solutions



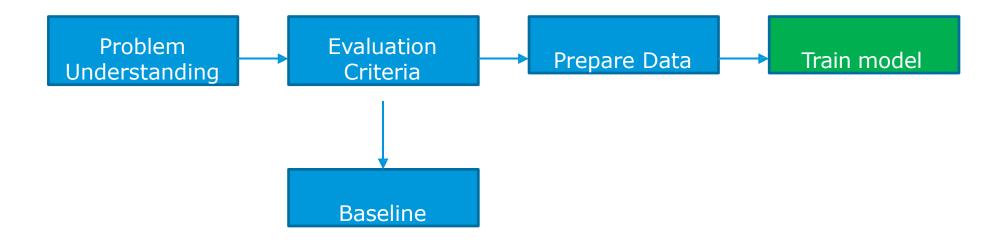


- "Enough" data in the sense that it's representative of the behaviour the model needs to learn
- Features: data transformations that encode your knowledge



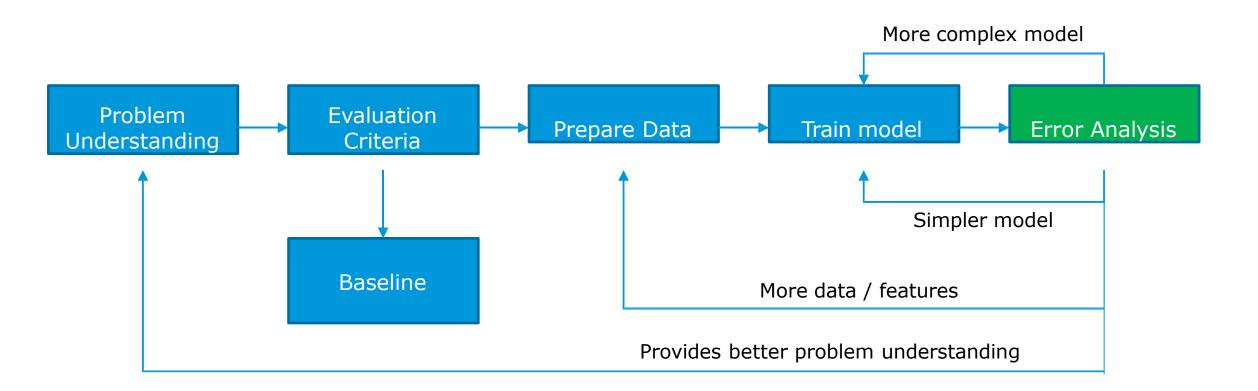


 Use data / features to tune the parameters that optimize the evaluation criteria (e.g. minimise error)

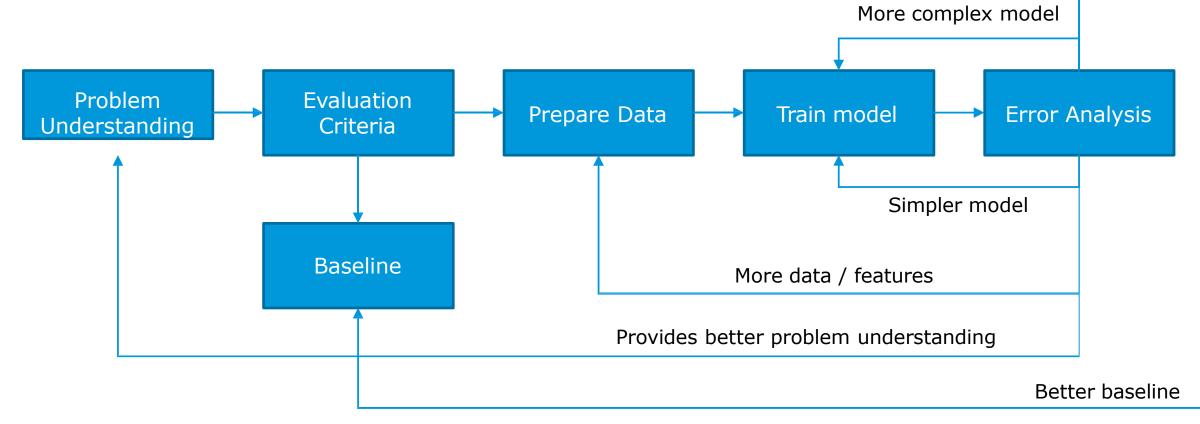




Understand what the model is doing: where is it right / wrong



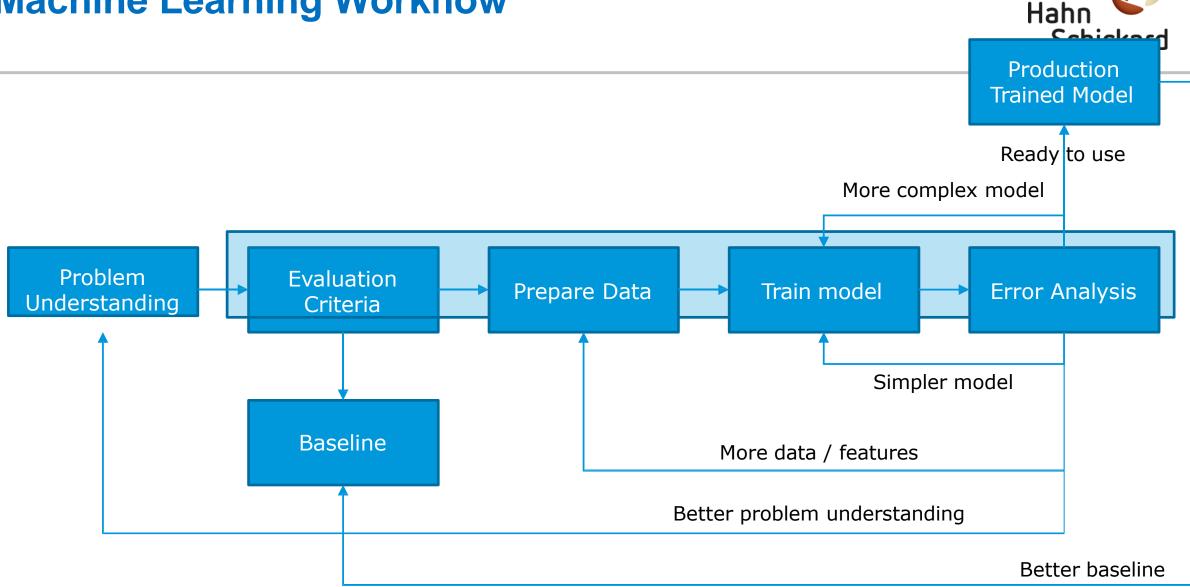
- When we are happy with the error, we can use the trained model: new data → features computation → model → results
- We now have a better baseline



**Production** 

Trained Model

Ready to use



### **Questions:**



What is Artificial intelligence, Machine learning and deep learning.

Where do we use Al already?

Explain: Supervised, Unsupervised, Semi-supervised, Reinforcement Learning

**Explain: Regression, Classification** 

