

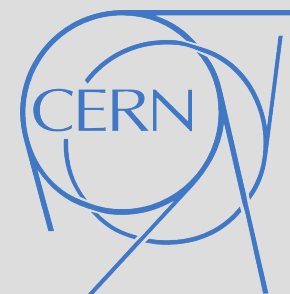
Digitalization & AI – What's it all about?

Bijeenkomst Vereniging van Arbeidsrecht

May 2019

About the speaker

- John Philip Ottersbach
 - German
 - Diploma thesis and Ph.D. in high energy physics working for the ATLAS experiment at CERN
 - Joined Anderson MacGyver in the early start-up phase to help organization with their transition into a data-driven, digital world
 - Digital Strategy
 - Digital Organization
 - Digital Solutions



Introduction

An international research facility

- CERN (Conseil Européen pour la Recherche Nucleaire):
 - Facility for fundamental research on matter in the Geneva area
 - Some 15.000 researches from 608 institutes/universities in 113 nations contribute
 - Objective: "Seek and find answer to the origin and nature of our universe."



A world map where countries are shaded in different colors to represent their status with CERN. Dark blue indicates full member states, medium blue indicates associate member states, light blue indicates associate members in the pre-stage to membership, grey indicates observers, and white indicates other states. The map shows a high concentration of member states in Europe, with some in North America, Africa, and Asia.

MEMBER STATES
ASSOCIATE MEMBER STATES
ASSOCIATE MEMBERS IN
THE PRE-STAGE TO MEMBERSHIP
OBSERVERS
OTHER STATES

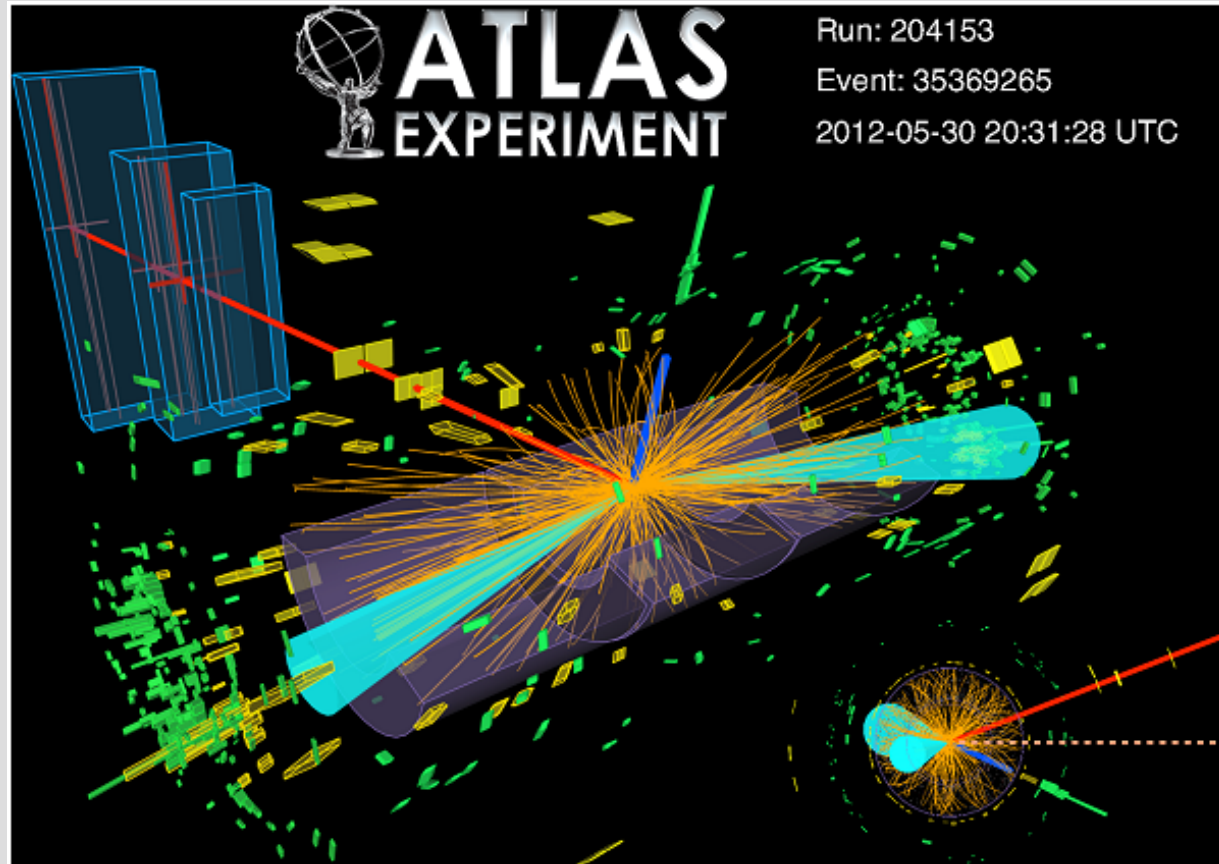
Accelerating science

- Large Hadron Collider:
 - 27 km circumference proton accelerator
 - Center of mass energy: 14 TeV → important to produce heavy, i.e. rare particles ($E=\gamma mc^2$)
 - 4 dedicated collision points, experiments



Challenges and reward

- Finding the Higgs particle after an intensive search
- Analyzing years of collision data - 40 million events per second \rightarrow 60 Terabyte of data per second
- Nobel prize 2013



The hype

From science to society and economy

- Developed at/in cooperation with CERN:
 - world wide web (Tim Berners-Lee)
 - grid computing
 - close to infinite hardware and software to deal with enormous amounts of data
- From science to economy = from fact to fiction (marketing)



The hype

Marketing vs. reality



The world of data

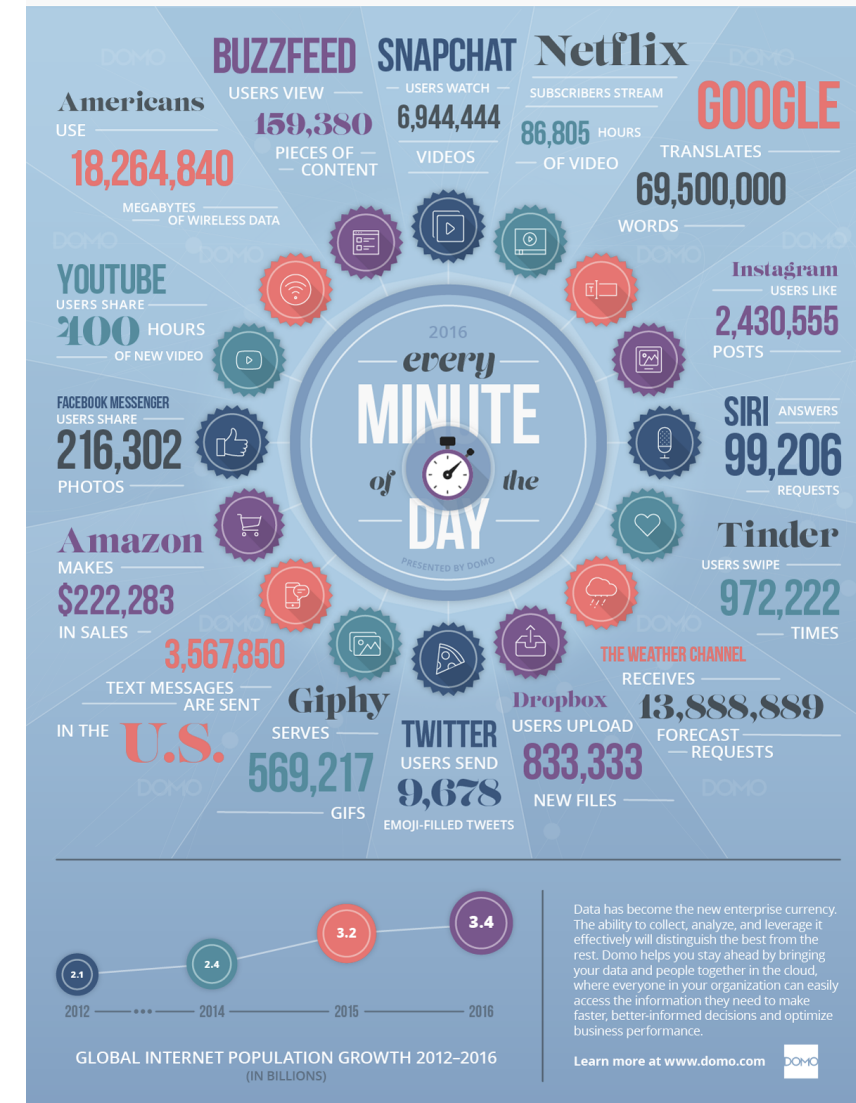
Data as the new oil

- The amount of data is growing exponentially at a rate of about 40% p.a. => the world's data volume is doubling every 2 years
- > 55% of the global population has internet access
- Every minute: >31 million Facebook messages, 400h YouTube movies, etc.
- Technology finds one's way into more and more domains of our life's (example: wearable's like the Apple Watch) => "Internet of Things"
- We create (un)consciously more and more data
- However, it needs to be "refined" to add value



DATA NEVER SLEEPS 4.0

How much data is generated every minute? In the fourth annual edition of Data Never Sleeps, newcomers like Giphy and Facebook Messenger illustrate the rise of our multimedia messaging obsession, while veterans like Youtube and Snapchat highlight our insatiable appetite for video. Just how many GIFs, videos, and emoji-filled Tweets flood the internet every minute? See for yourself below.



SOURCES: SNAPCHAT, NETFLIX, GOOGLE, INSTAGRAM, TINDER, THE WEATHER COMPANY, DROPBOX, GITHUB, GIPHY, YOUTUBE, BUZZFEED, AMAZON, CTA, MARY MECKERS 2016 INTERNET TRENDS REPORT, USA TODAY, GLOBAL WEB INDEX

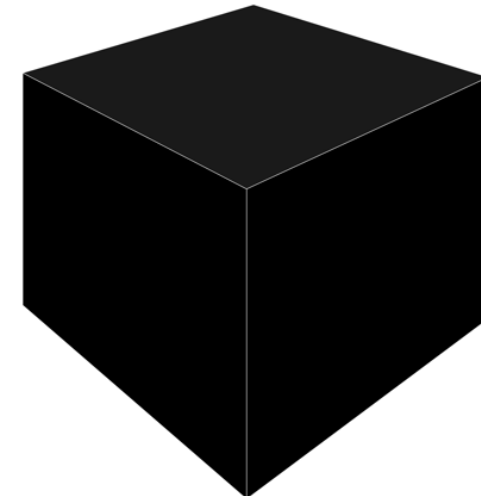
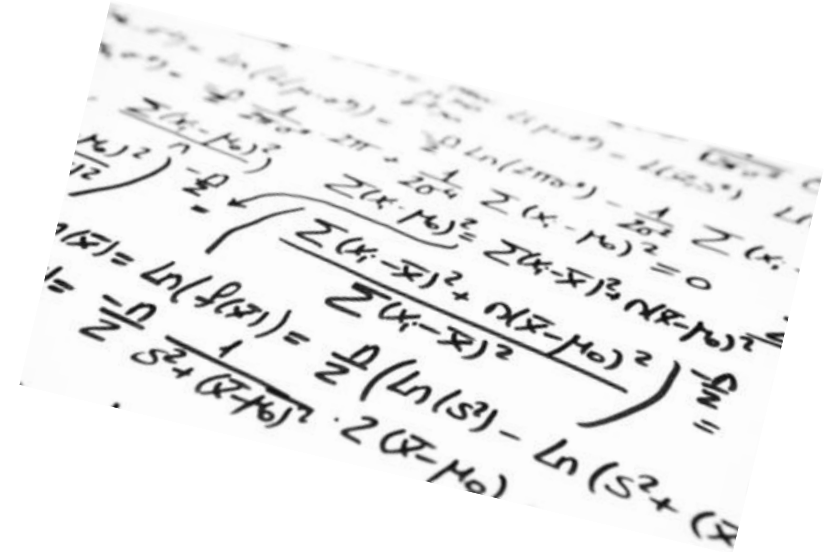
What data is good for ...



Refining the new oil

The science of collecting, organizing and drawing conclusions from data:

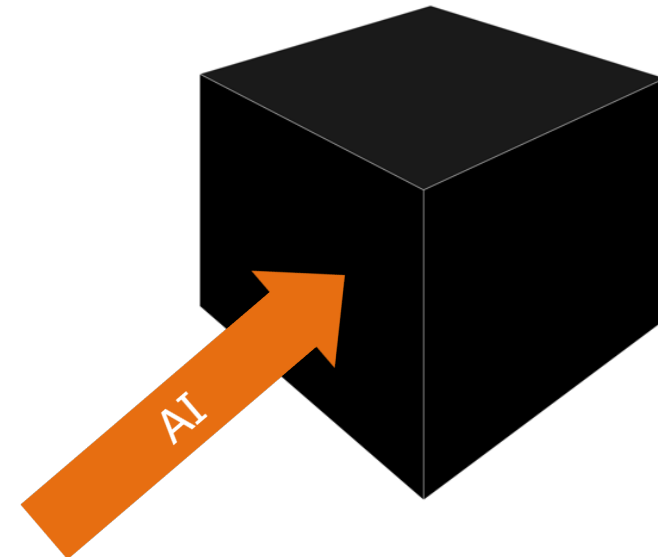
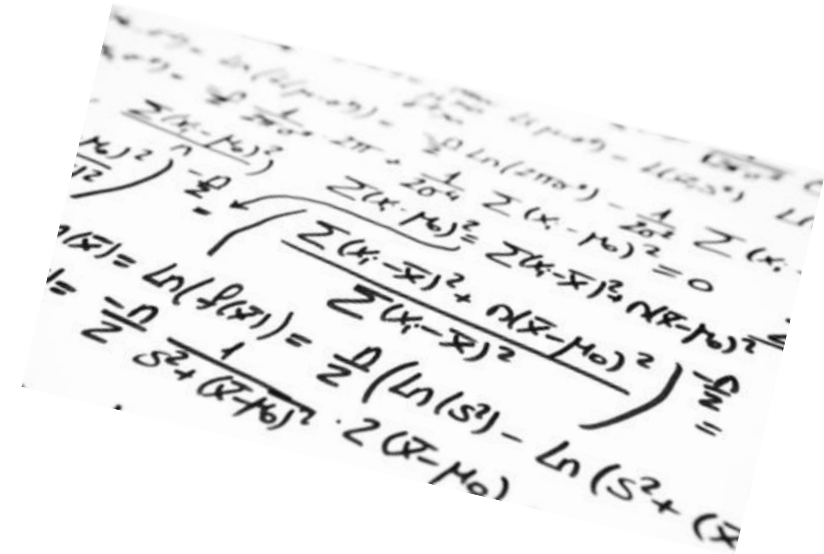
- Classical statistics:
 - A domain of mathematics, partially centuries old, well proven and tested
 - Charts/graphs, parameter estimation, hypothesis testing, regression, maximum likelihood, ...
 - Results are easily verifiable
- Data mining:
 - A relatively recent development (made possible through the advance of computing power -> Moore's law)
 - Algorithms for predictions, classifications, segmentation (e.g. neural networks, boosted decision tree's, support vector machines, ...)
 - Results are often difficult to verify => black boxes



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The beginnings

- Artificial Intelligence is not a well and uniquely defined term
- Dynamic definitions; AI effect → as machines become more capable, tasks that are considered to require “intelligence” are removed from the definition of AI
- In the past decade, term captured by marketers
- Research started in the mid 1950’s
- Boost through nowadays available computing capacities

“The ability of a digital computer or computer-controlled robot to perform tasks commonly associated with intelligent beings.”

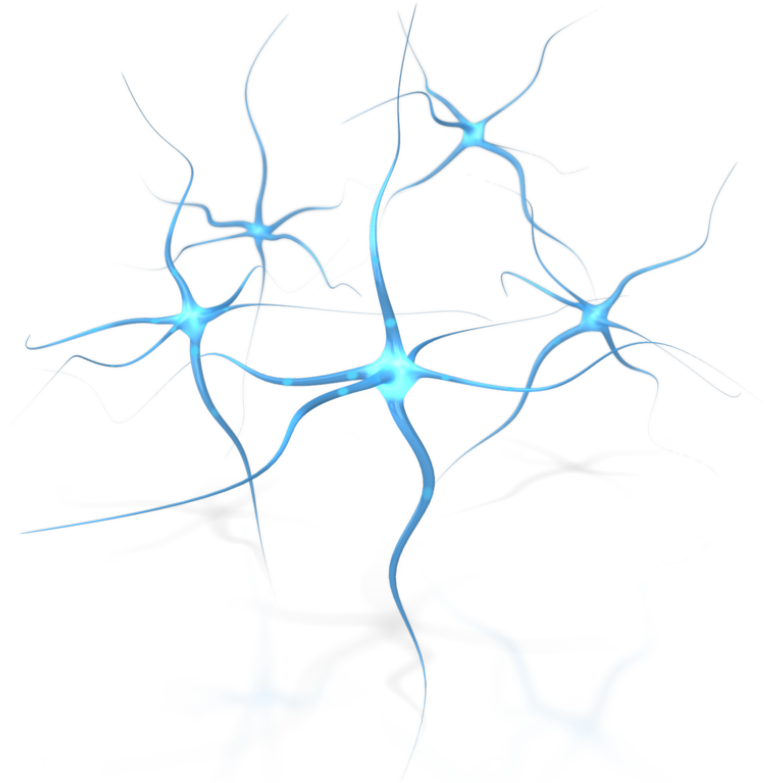
Encyclopedia Britannica

“The capability of a machine to imitate intelligent human behavior.”

Merriam-Webster

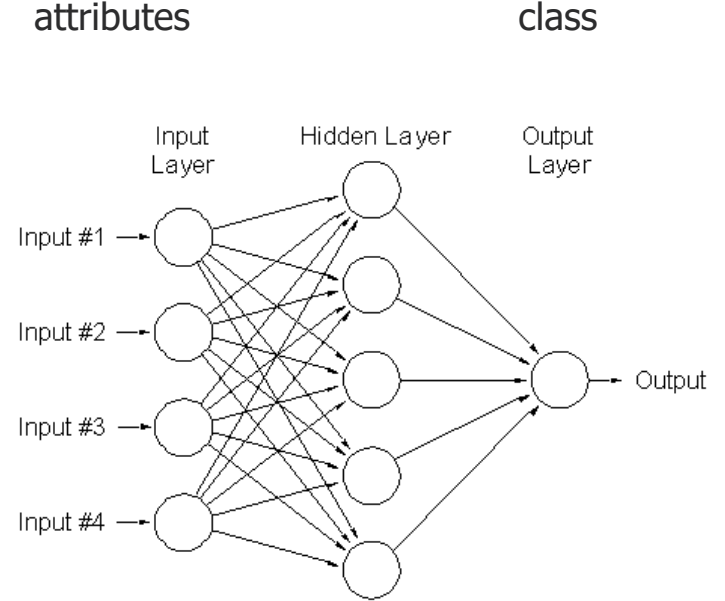
Neural networks

- First publication on the basic perceptron algorithm (Rosenblatt, 1957) => breakthrough in the 1990's with development of *multilayer perceptrons* and *backpropagation*
- Multilayer perceptrons have input, often hidden, and output layer
- Each connection has a weight (a real number)
- Each node performs a weighted sum of its inputs and thresholds the result (often using a sigmoid function)
- The output is determined using a linear combination of weighted input from the hidden layer
- Perceptron convergence theorem: Converges when cycled repeatedly through the training data, provided the problem is linearly separable!

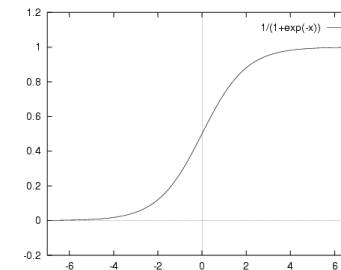


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$$x = w_0 + w_1 a_1 + w_2 a_2 + \dots + w_n a_n = \sum_{i=0}^n w_i a_i$$



The case

A challenge for a company in the logistics sector:

- Around 5000 independent parties can deliver goods daily and unannounced
- Rented cars and deployed staff for internal processing/distribution of the goods are expensive => good, reliable planning is crucial



A real-world example of applied AI

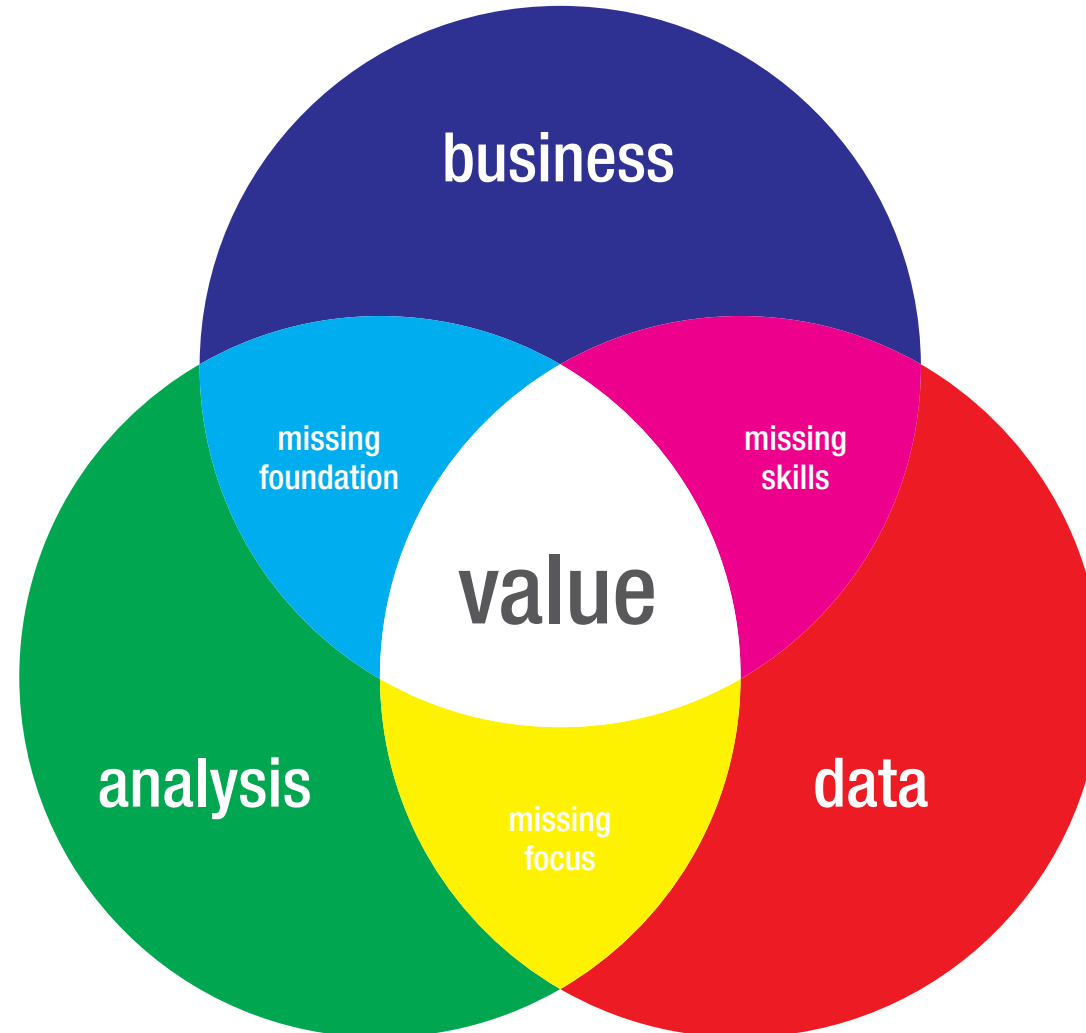
Approach

- Can we predict the quantity of delivered goods based on historic data?
- The *class* is here the quantity of delivered goods (in number of cars) => regression problem
- Challenges and approach:
 - Find *attributes* that determine/have a large impact on the *class*
 - If we can find *attributes* that have an impact on future instances of the *class*, we can make predictions!
 - Find an algorithm that handles this specific regression problem well



A real-world example of applied AI

Finding the right attributes



A real-world example of applied AI

Finding the right attributes

Heuristic, iterative process:

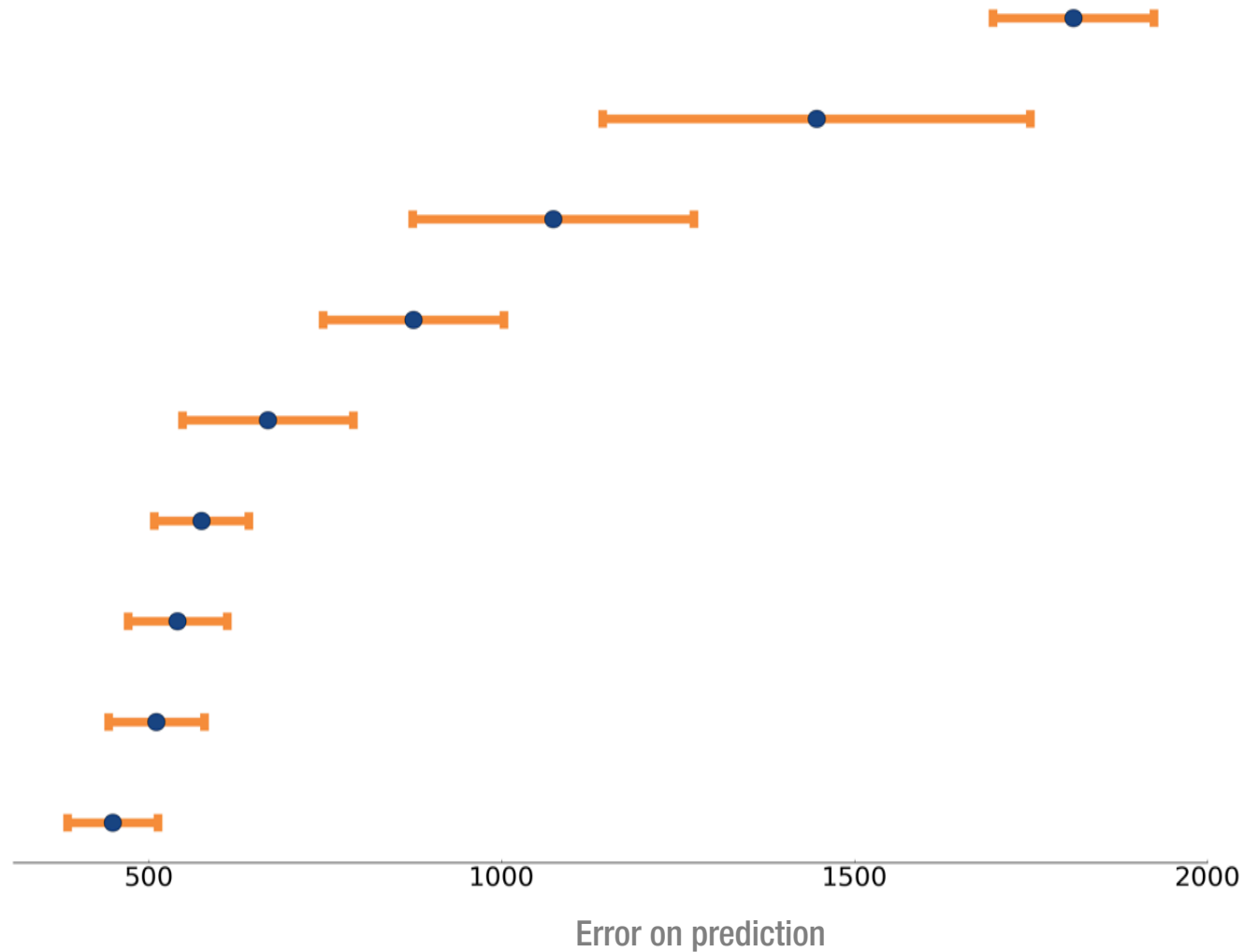
1. Think of effects that might have an influence on the *class*. What might a model of the *class* depend on (e.g. the weather) ?
2. Translate them into variables that can be computed using data. What data will be needed? How can we get access to still missing data (e.g.: use open weather data to get access to temperatures, precipitation, etc.) ?
3. Try different implementations of each variable and see what works best (e.g.: average air temperature in different altitudes, averages over different time periods, etc.)
4. Continue adding attributes until the optimal performance of the model is reached



A real-world example of applied AI

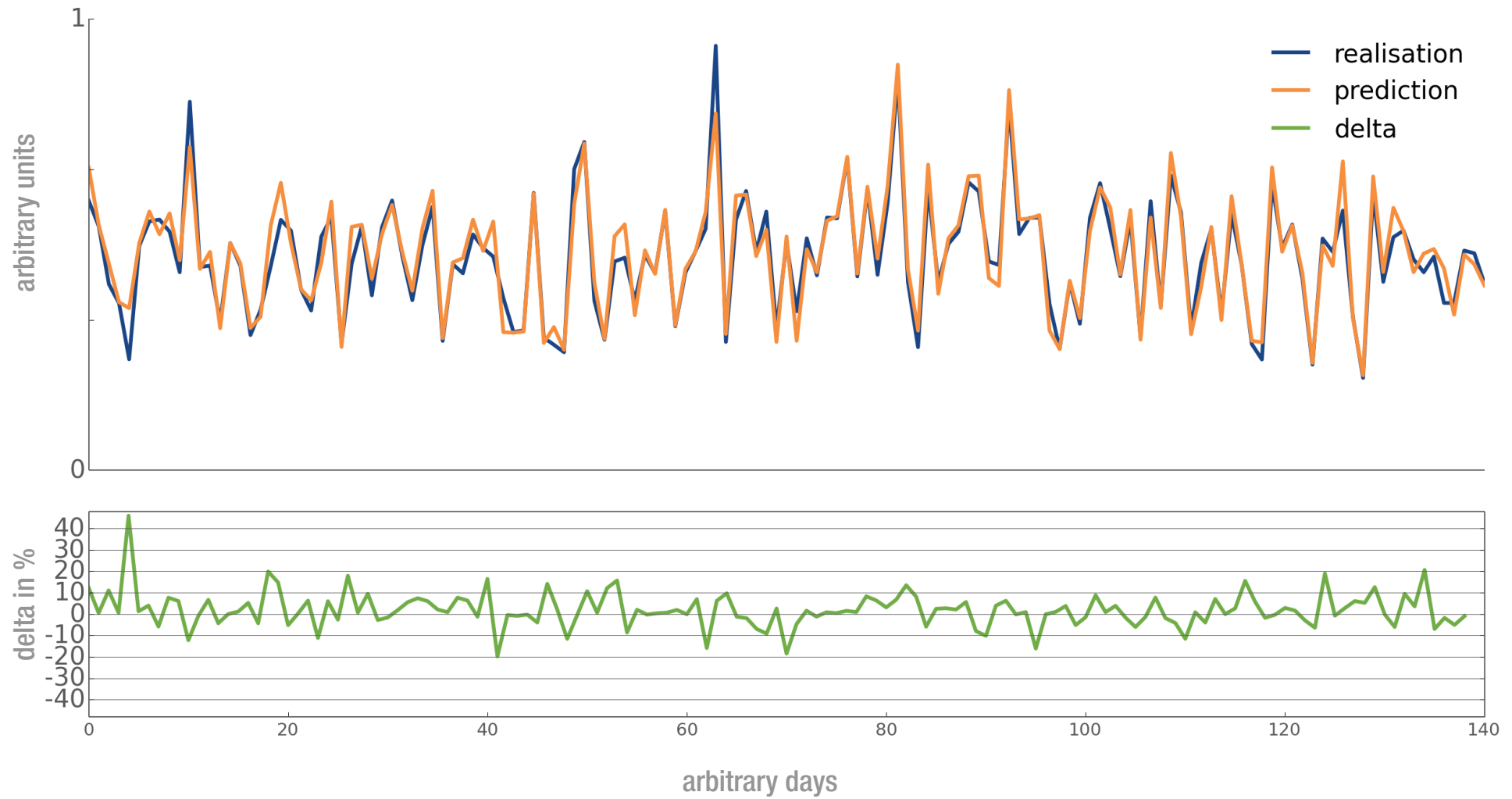
Attribute selection

- Benchmark (ZeroR)
- Neural network
- + Holidays
- + Weather + pricing
- + Trend
- + Parameter optimisation
- + Additional data
- + Holiday optimisation
- + Removing history



A real-world example of applied AI

Results



Take aways

- What's in the black box?
- Artificial intelligence – painstakingly stupid
- The smarter you want to get, the more data (of the same type) you need
- Predictions – looking back in time
- Nevertheless, with the right amount of data and proper analysis you can get a very long way with AI (e.g. Siri, autonomous cars, real-time translation, etc.)
- It's booming and it will continue to do so! Focus areas:
 - specific computer chips for more computing power
 - lots of R&D to shed some light on that black box





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