



Algorithmus Schmiede

Data Science | Numerik | Physik

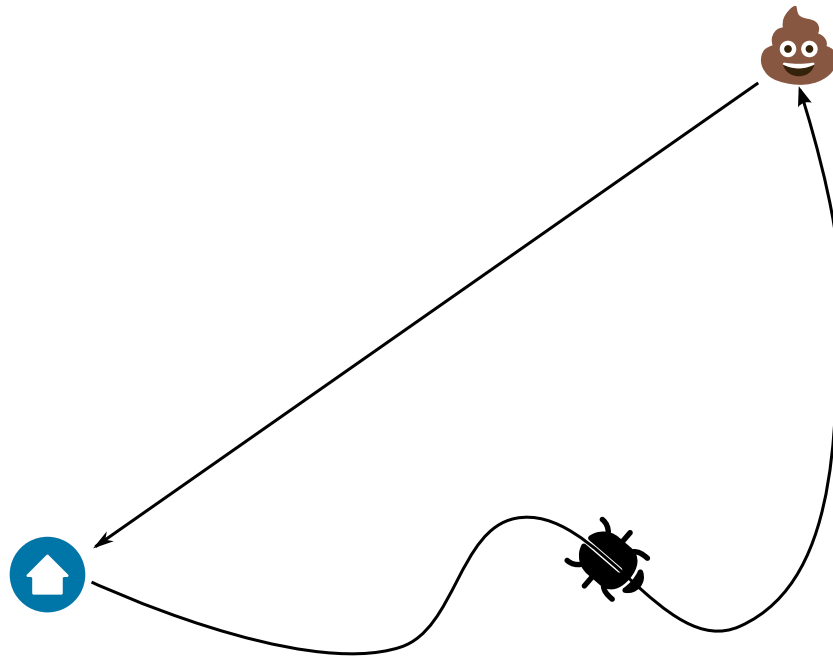
Understanding AI in Manufacturing

Dr. Markus Dutschke

Definition AI

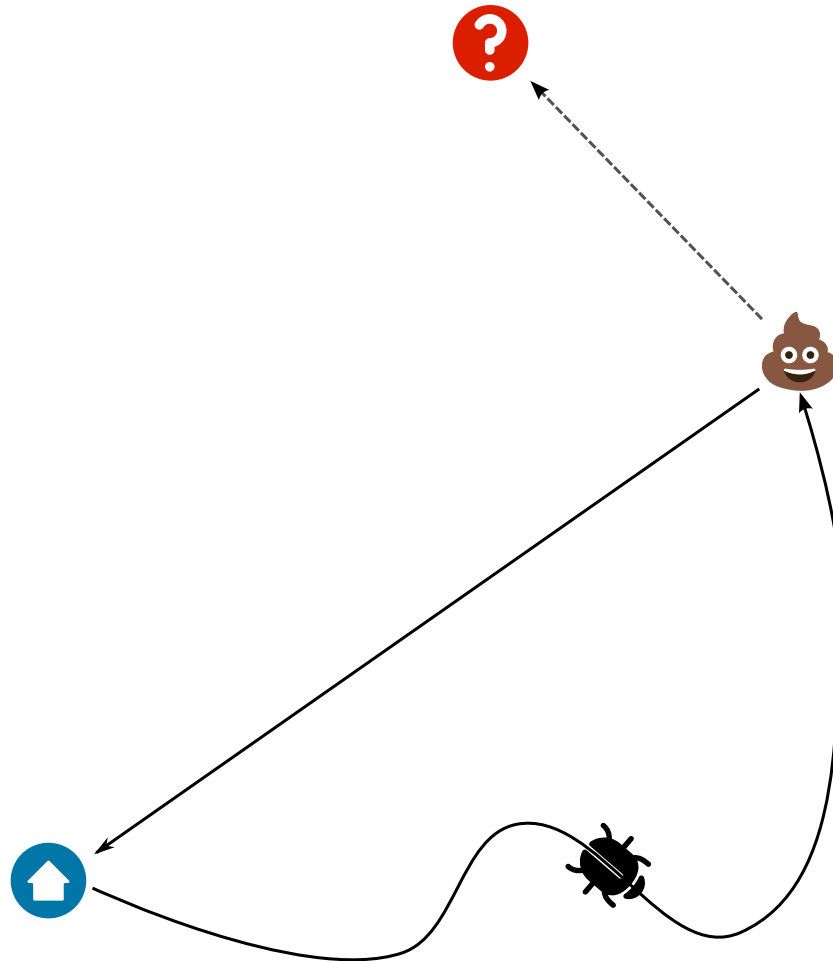
- Artificial Intelligence (AI):
Solving complex cognitive challenges
e.g. also chess computers (1951, 1967, 1996)
- Machine Learning (ML):
Algorithms configure themselves on the basis of data
e.g. neural networks, transformers

Intelligence of the dung beetle



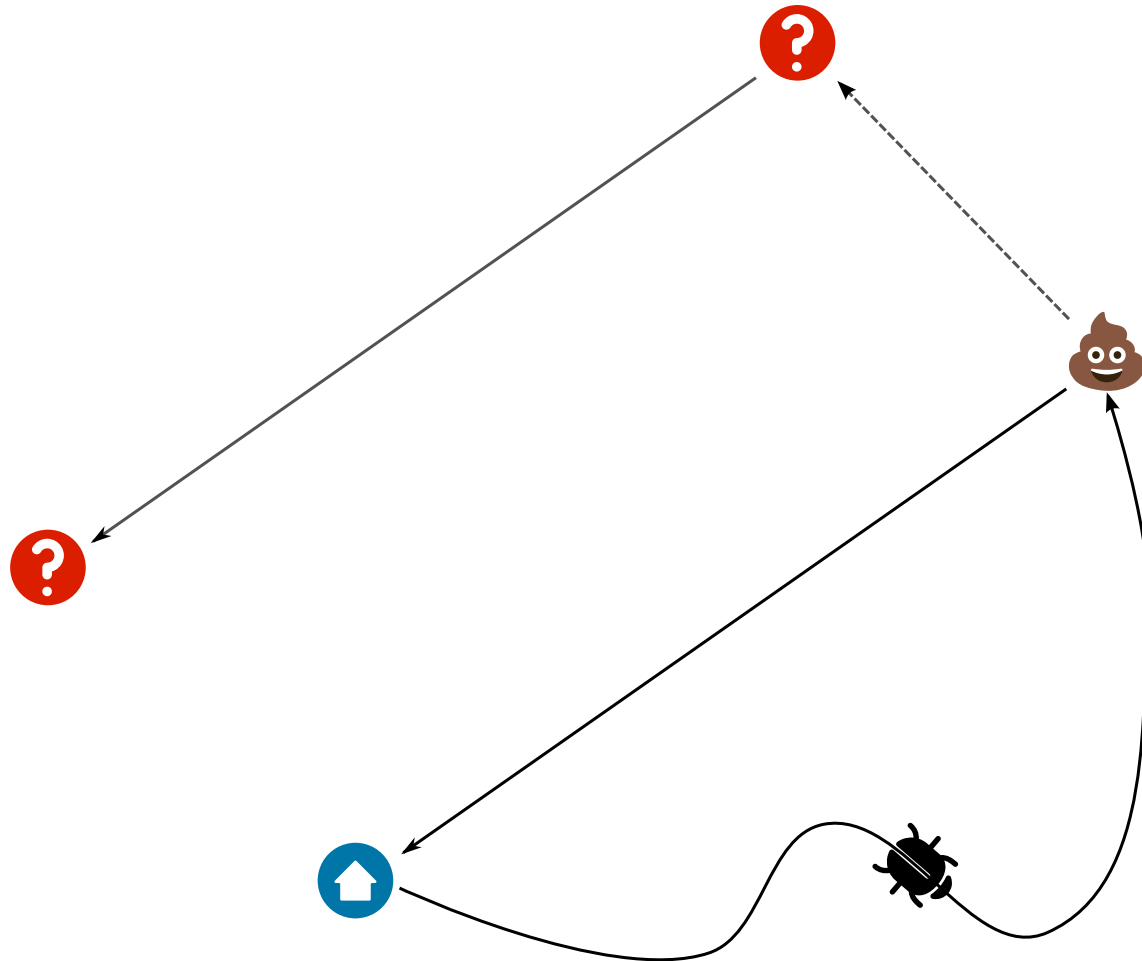
Künstliche Intelligenz
Joachim Reinhart, Oliver Mayer, Christian Greiner
ISBN: 978-3-8343-3511-1

Intelligence of the dung beetle



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Function ► Regressor

$$x=8$$



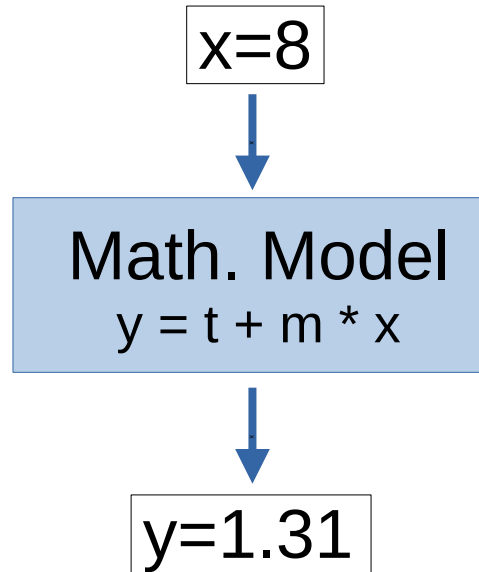
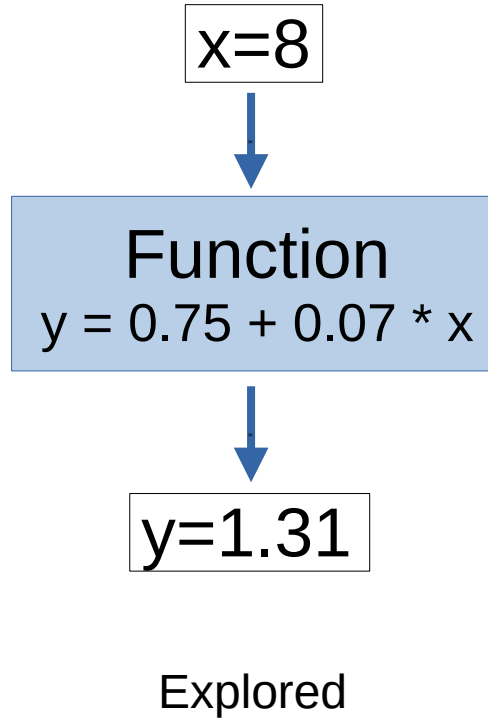
$$\text{Function}$$
$$y = 0.75 + 0.07 * x$$



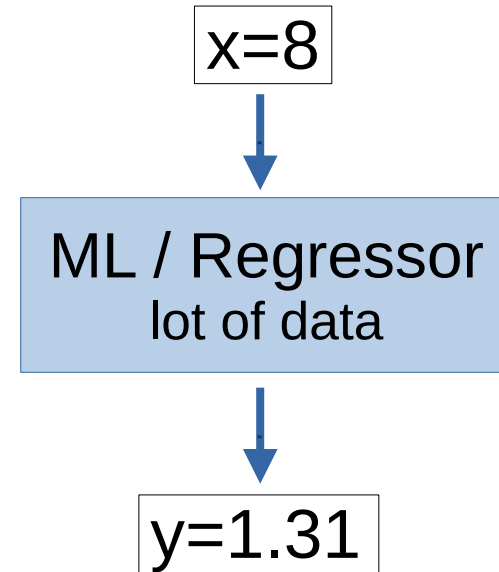
$$y=1.31$$

Explored

Function ► Regressor

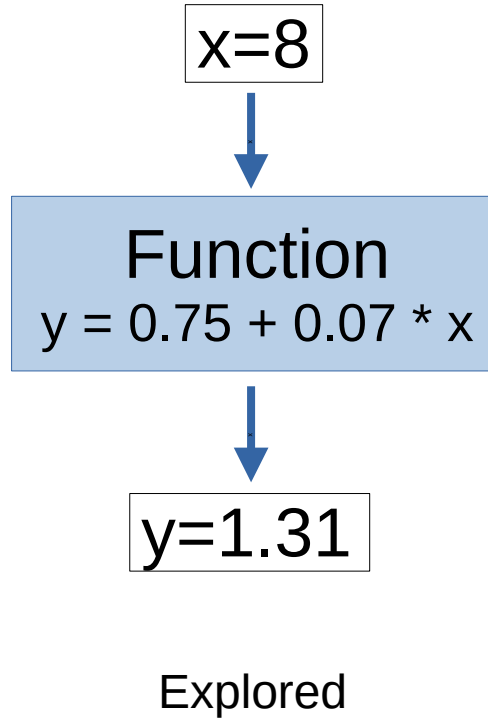


Domain Knowledge

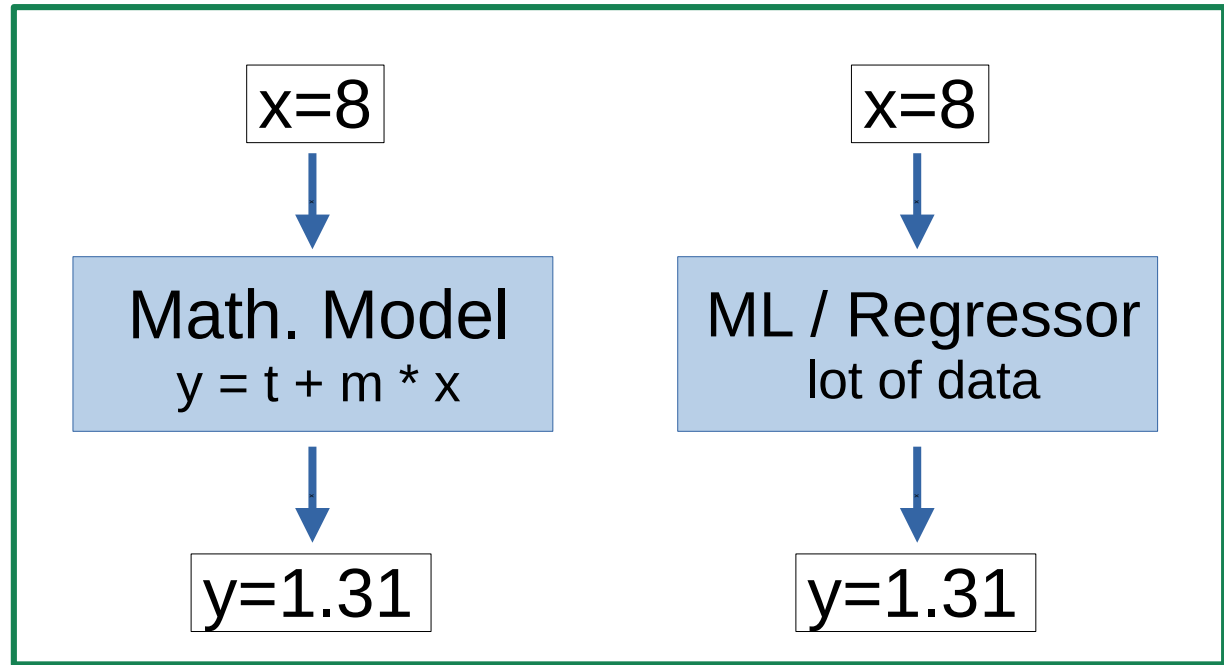


Data Driven

Function ► Regressor



This Talk

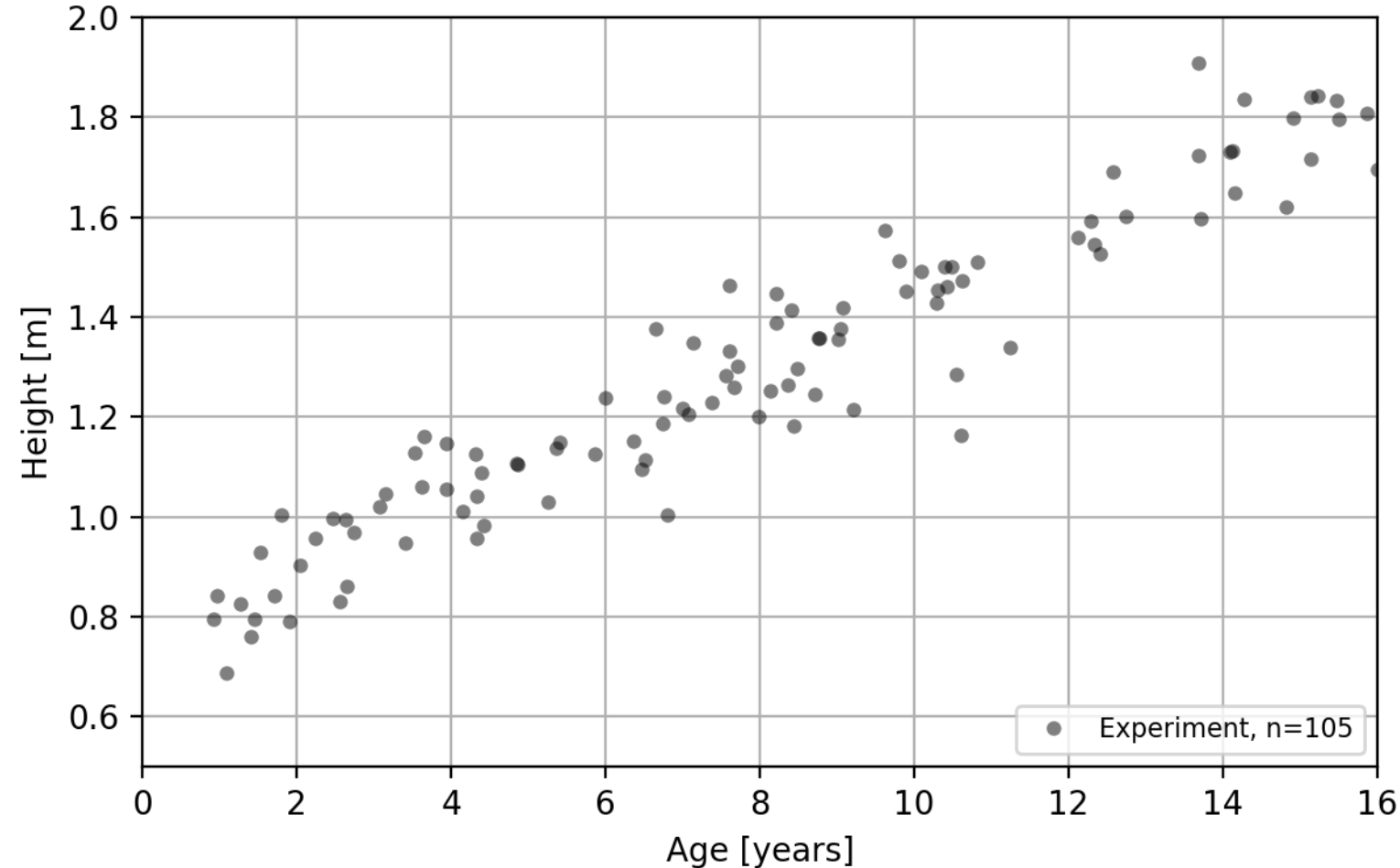


Domain Knowledge

Data Driven

Linear Regression

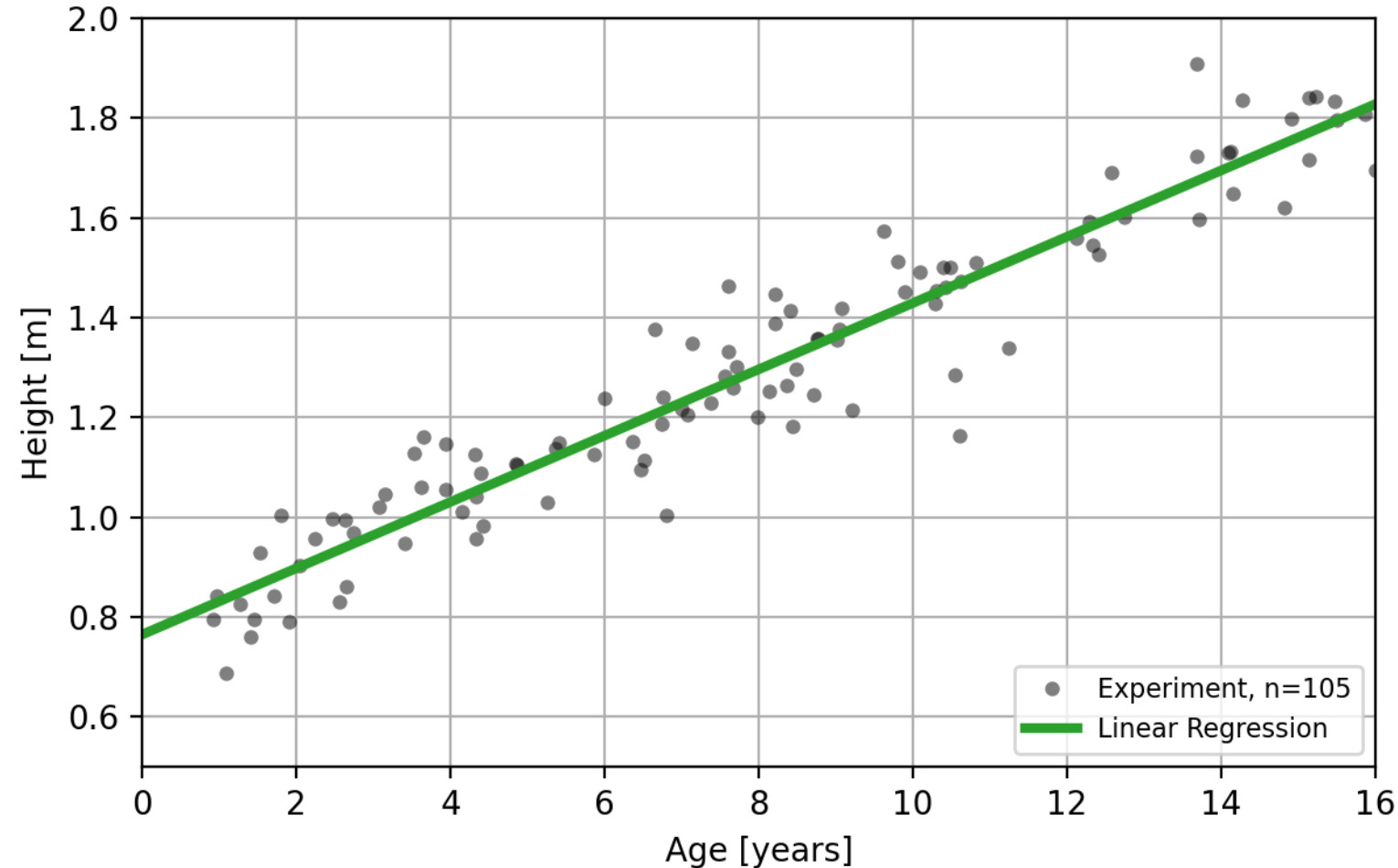
Growth of Children



Source:
German federal health
reporting
www.gbe-bund.de

Linear Regression

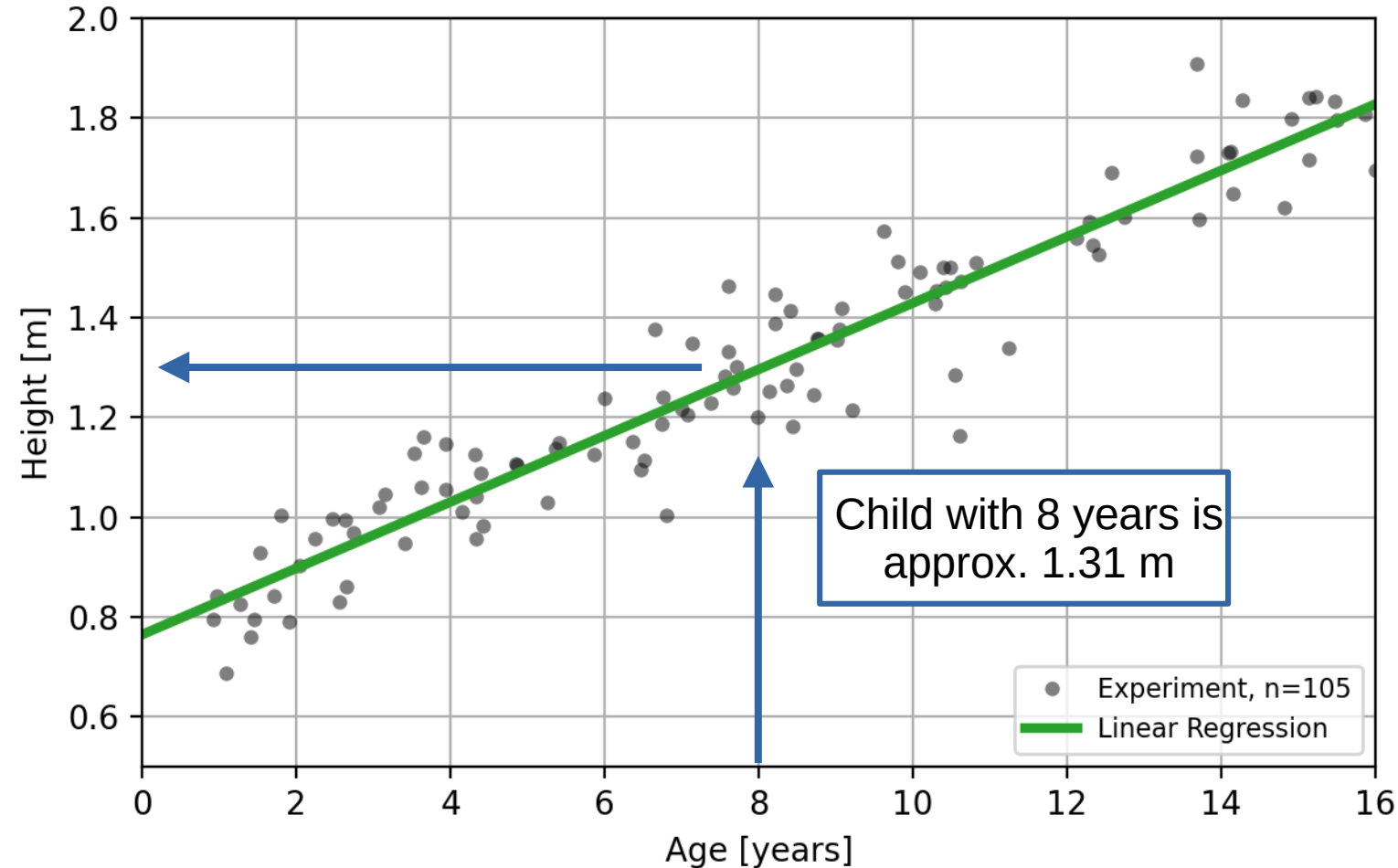
Growth of Children



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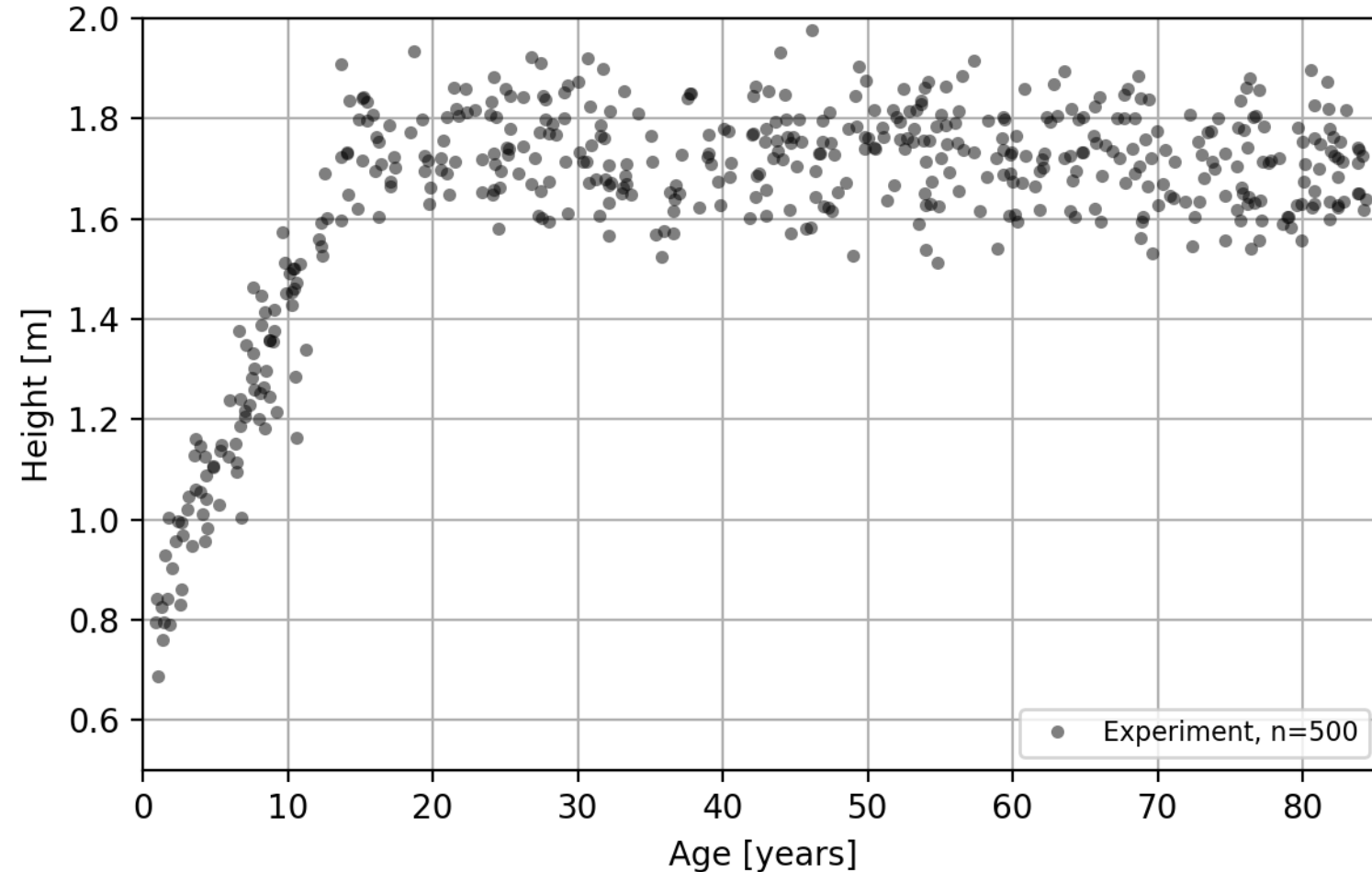
Linear Regression

Growth of Children



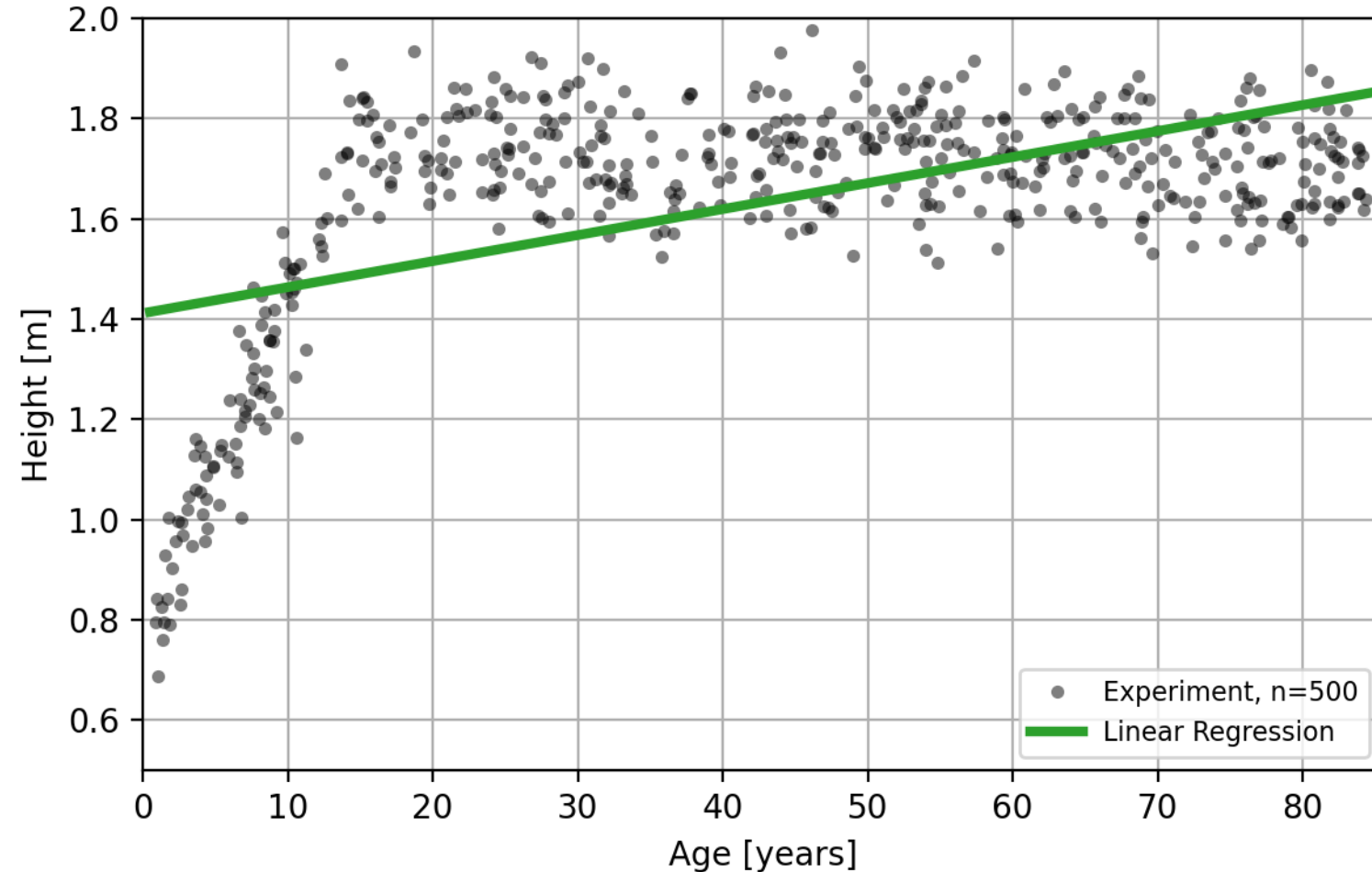
Source:
German federal health reporting
www.gbe-bund.de

Overview Regressors



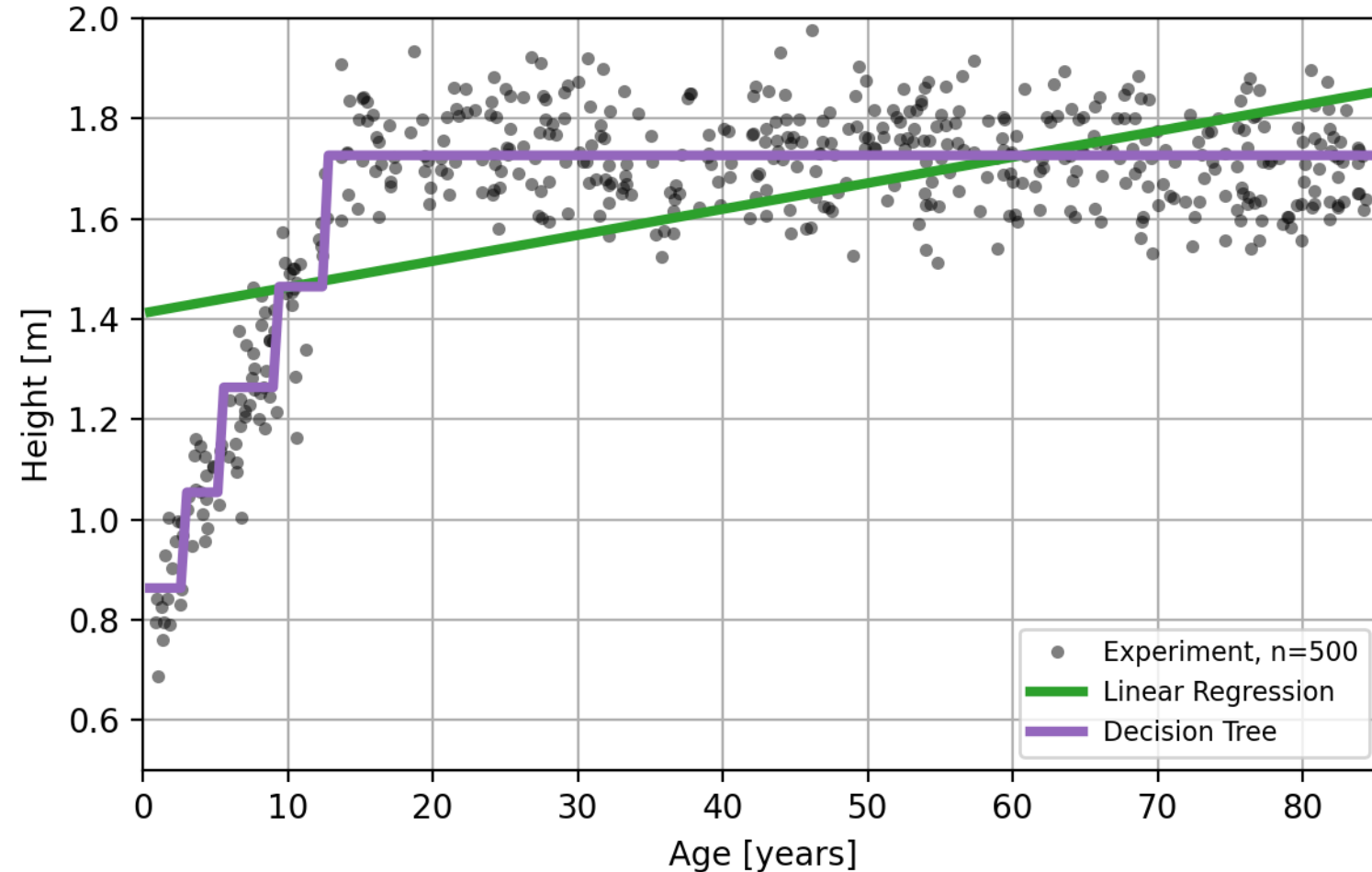
Source:
German federal health
reporting
www.gbe-bund.de

Overview Regressors



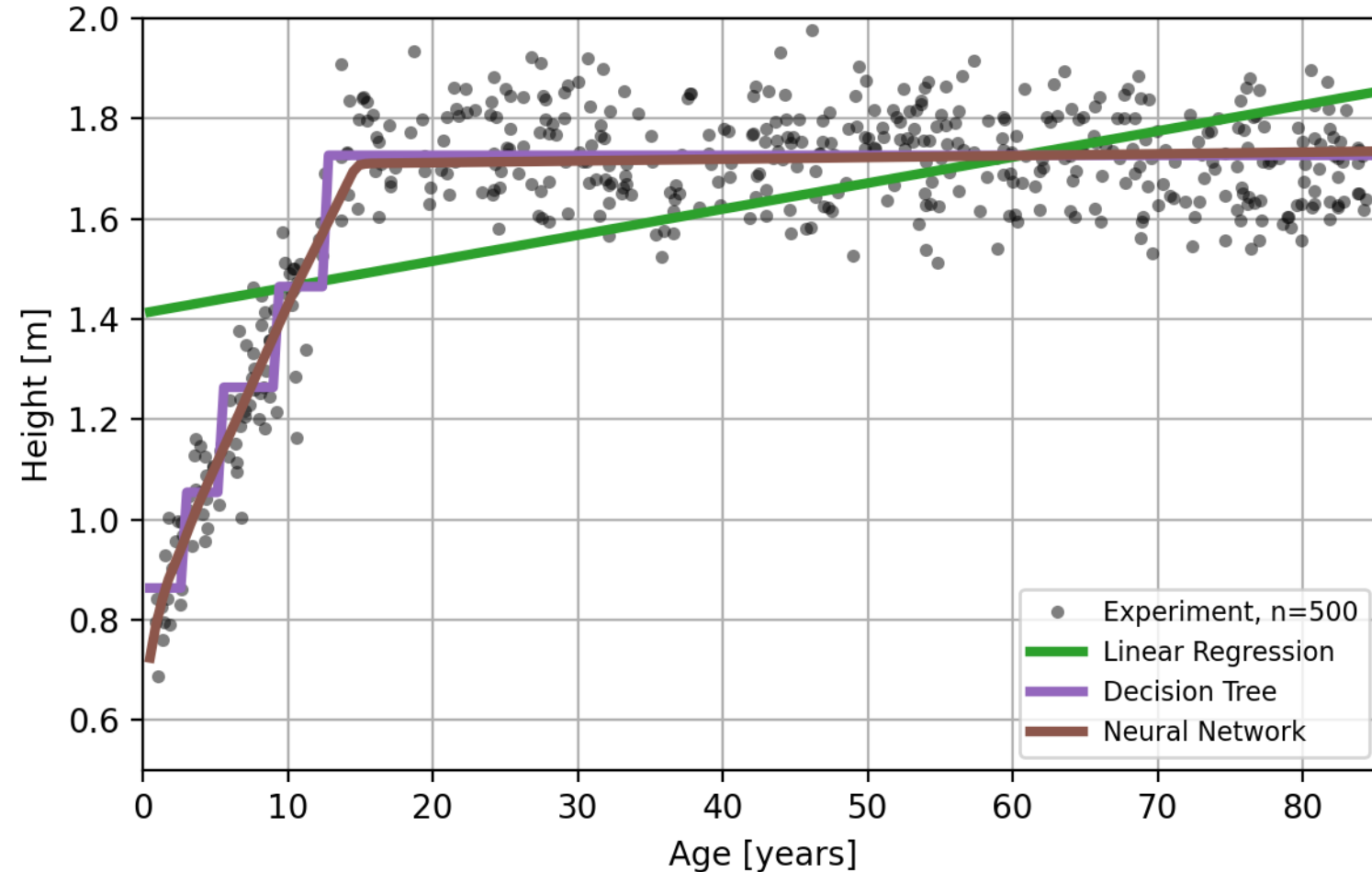
Source:
German federal health reporting
www.gbe-bund.de

Overview Regressors



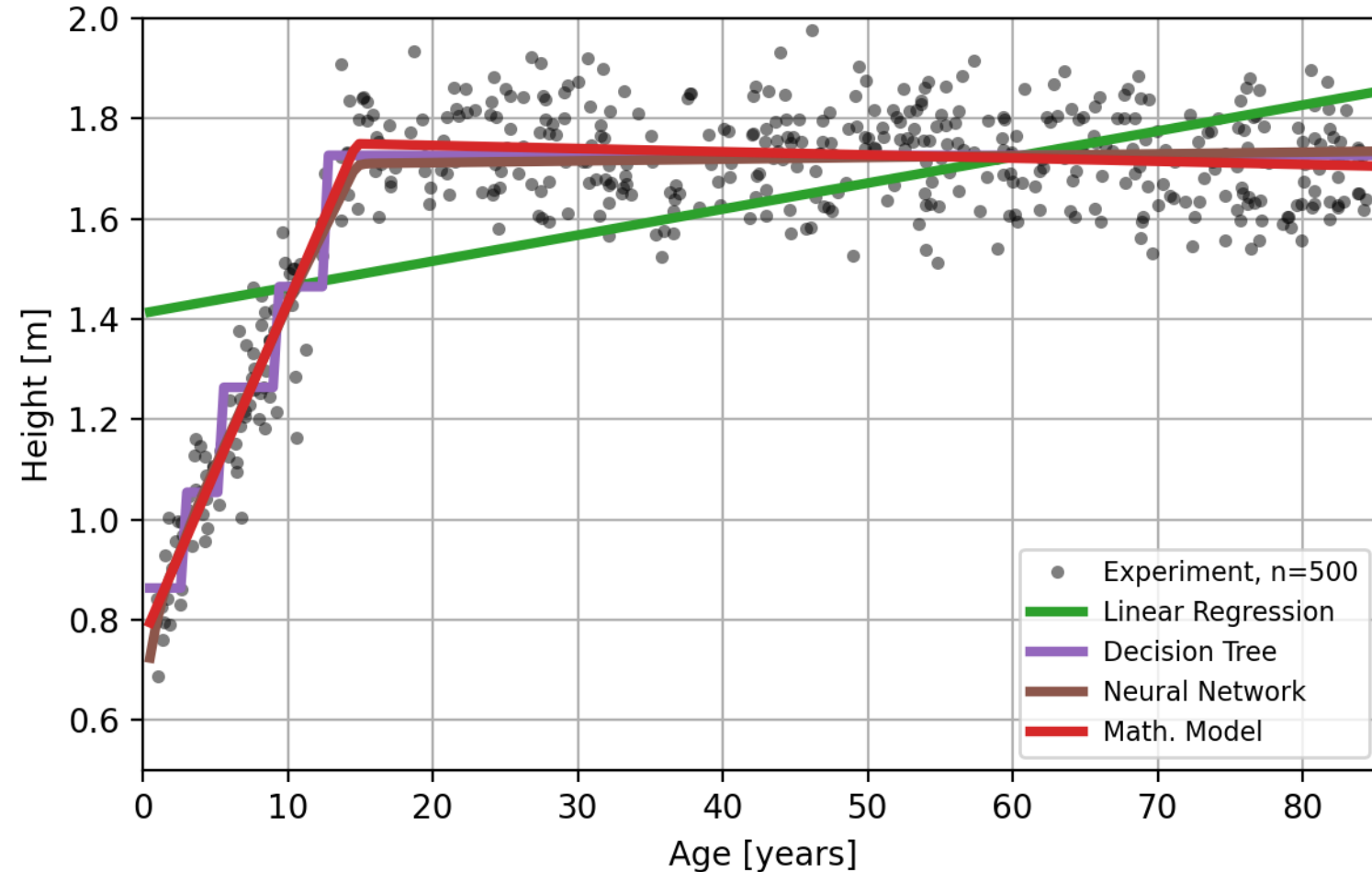
Source:
German federal health reporting
www.gbe-bund.de

Overview Regressors



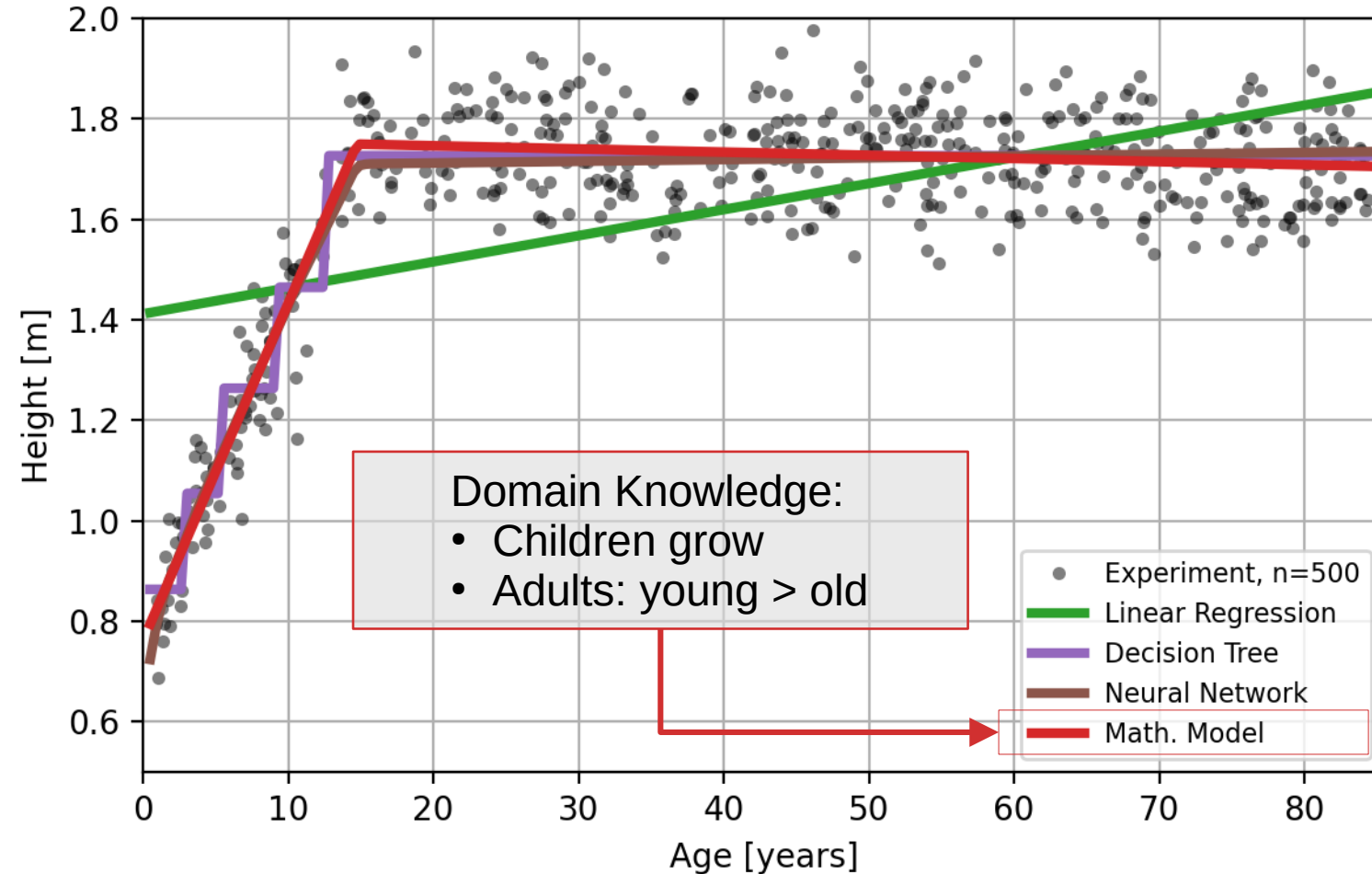
Source:
German federal health reporting
www.gbe-bund.de

Overview Regressors



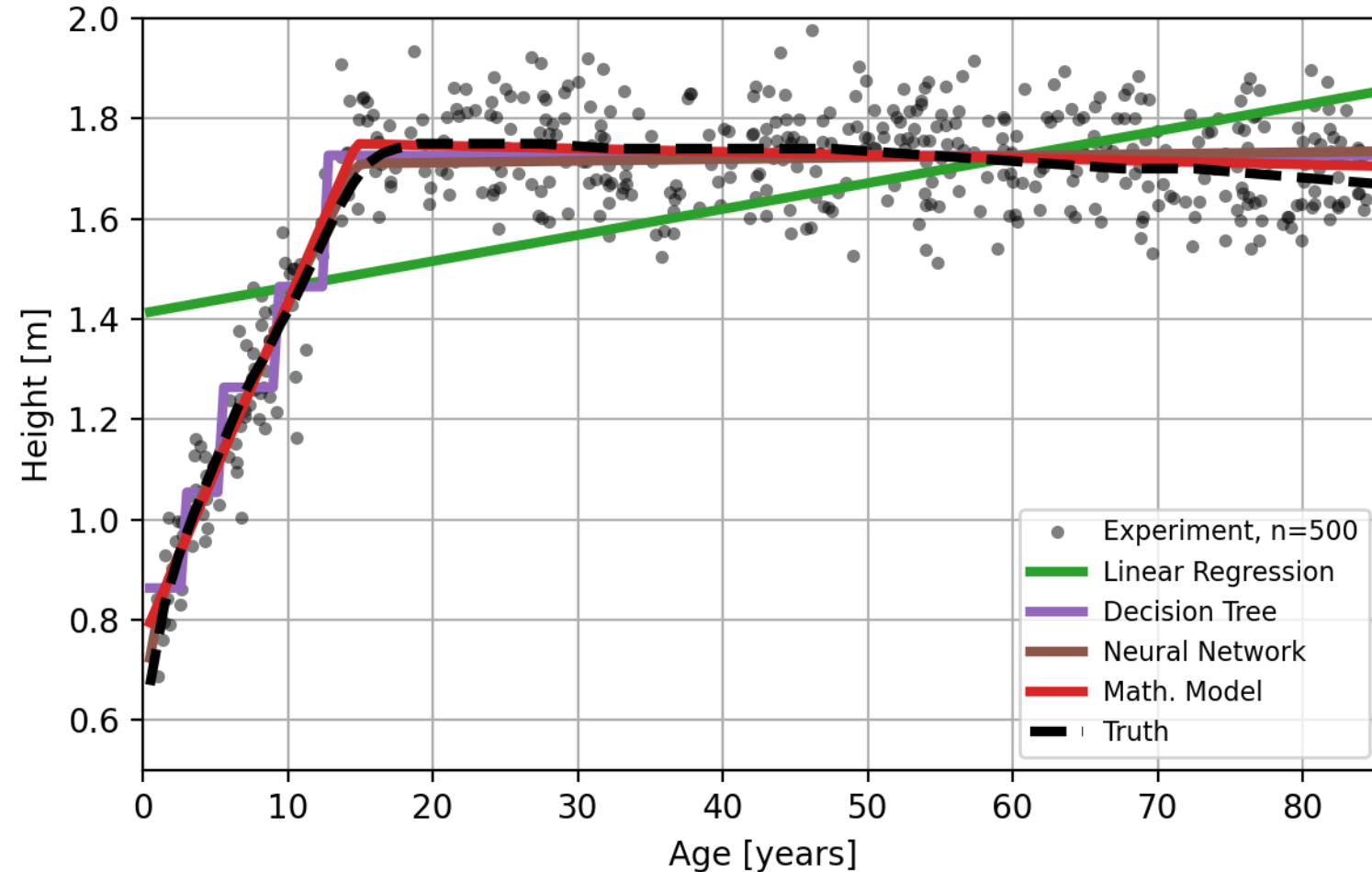
Source:
German federal health reporting
www.gbe-bund.de

Overview Regressors



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German federal health
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Overview Regressors



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AI Solution Strategies



„Developing a fancy theory“

x=8



Math. Model
 $y = t + m * x$

little data
more effort
more precise



y=1.31

„Compensating for ignorance with data“

x=8



ML / Regressor
lot of data

lots of data
little effort
automatable



y=1.31

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Data Driven

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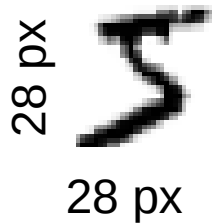
Domain Knowledge

Data Driven

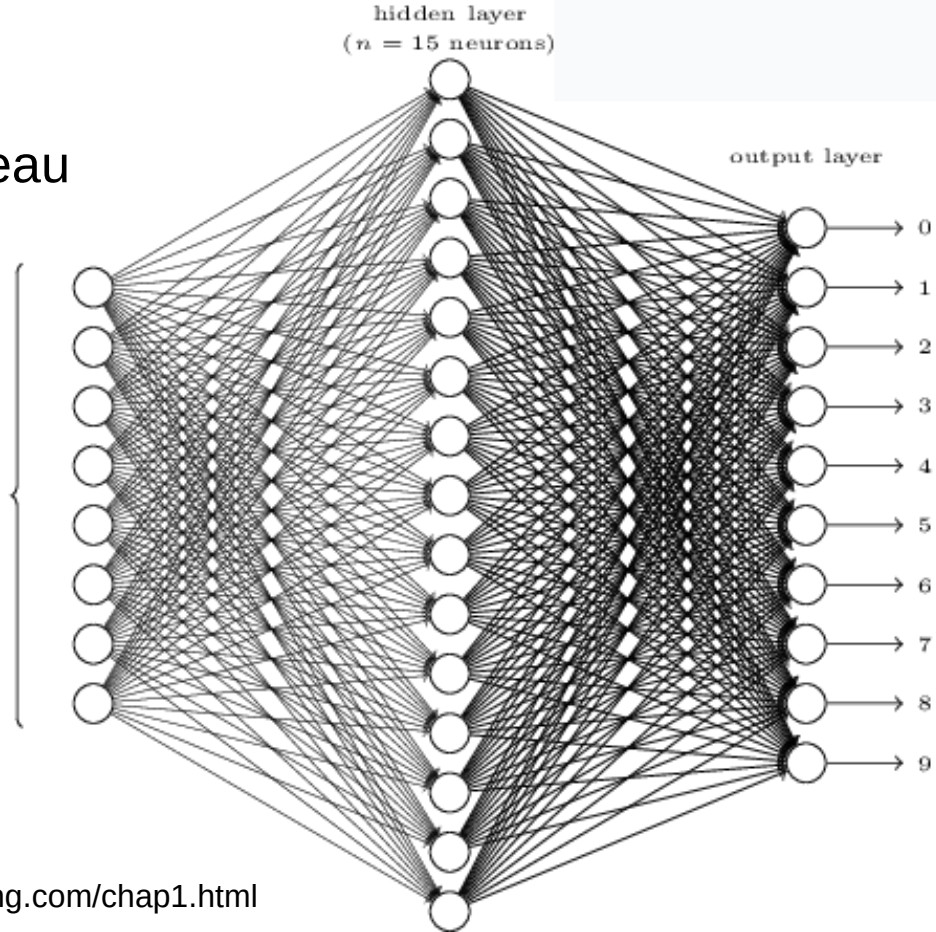
NN @ Handwritten Digits

5 0 4 1 9 2

- NIST data set „handwritten digits“
- 250 authors from US Census Bureau
- Training: **60 000 Digits**
- Test Data: 10 000 Digits



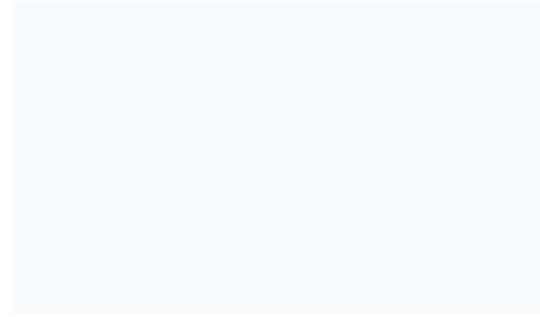
input layer
(784 neurons)



- Precision with ML: 99,79%
(<https://proceedings.mlr.press/v28/wan13.pdf>)
- Precision without ML: 20% - 50%

<http://neuralnetworksanddeeplearning.com/chap1.html>

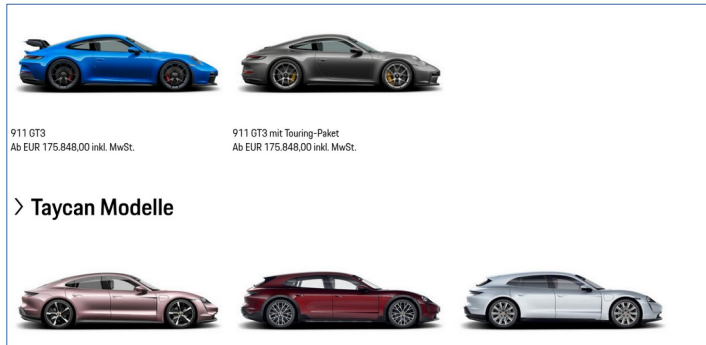
AI: Problems and Risks



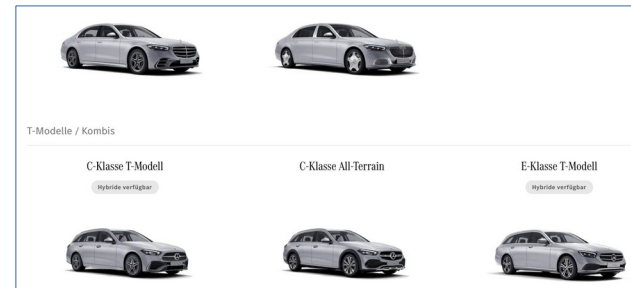
Bias in Training Data

Task: Picture -> Brand

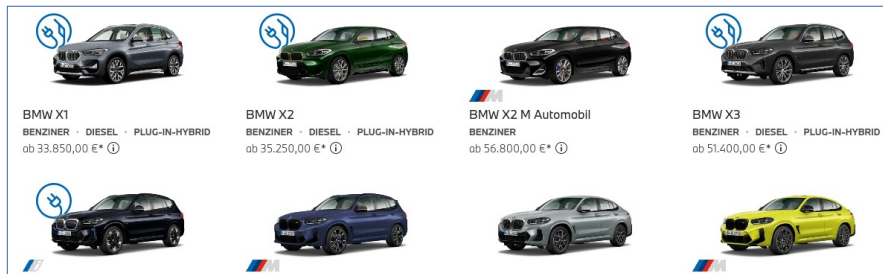
<https://www.porsche.com/germany/models/>



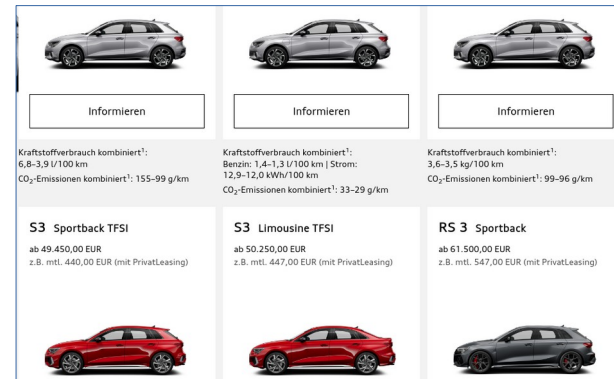
<https://www.mercedes-benz.de/passengercars/models.html>



<https://www.bmw.de/de/neufahrzeuge.html>



<https://www.audi.de/de/brand/de/neuwagen.html>



Hacking Neural Networks



ALEX LEE SECURITY 11.05.2020 06:00 AM

wired.co.uk

This ugly t-shirt makes you invisible to facial recognition tech

Researchers at Northeastern University have developed an adversarial example that works even when printed onto a moving fabric

Fooling a Real Car with Adversarial Traffic Signs

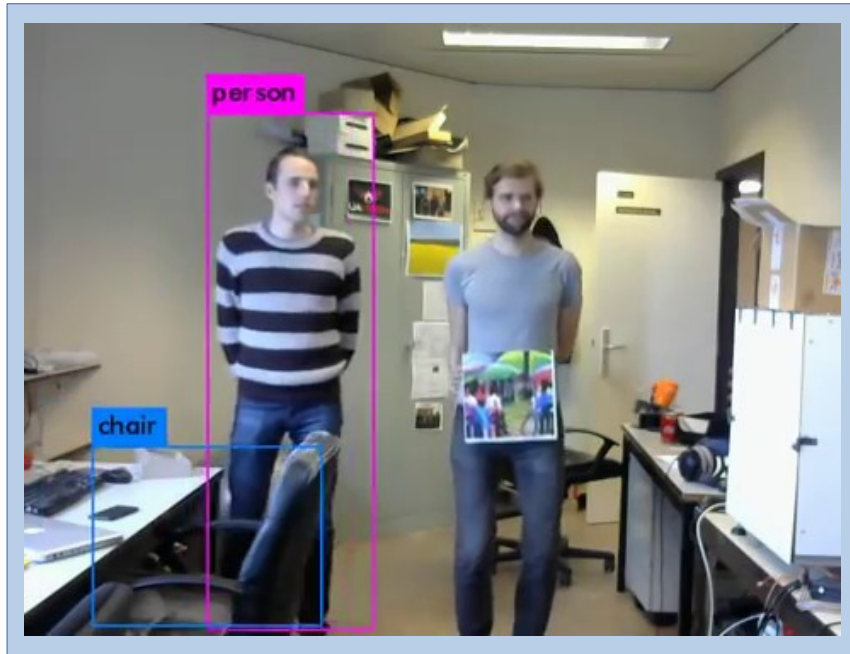
Nir Morgulis, Alexander Kreines, Shachar Mendelowitz, Yuval Weisglass

Harman International, Automotive Security Business Unit arxiv.org

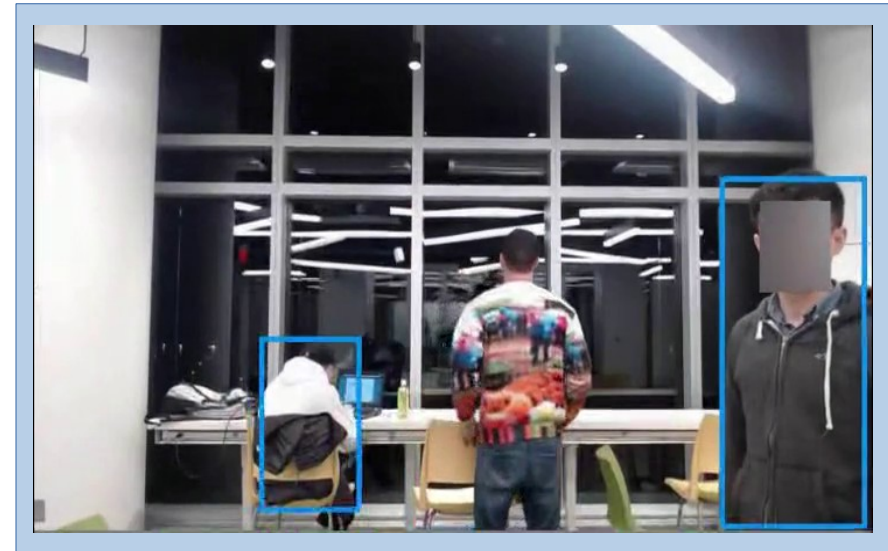


classification: 30 km/h

Hacking Neural Networks



Anonymous CVCOPS:
Generating adversarial patches against YOLOv2
<https://www.youtube.com/watch?v=MlbFvK2S9g8>

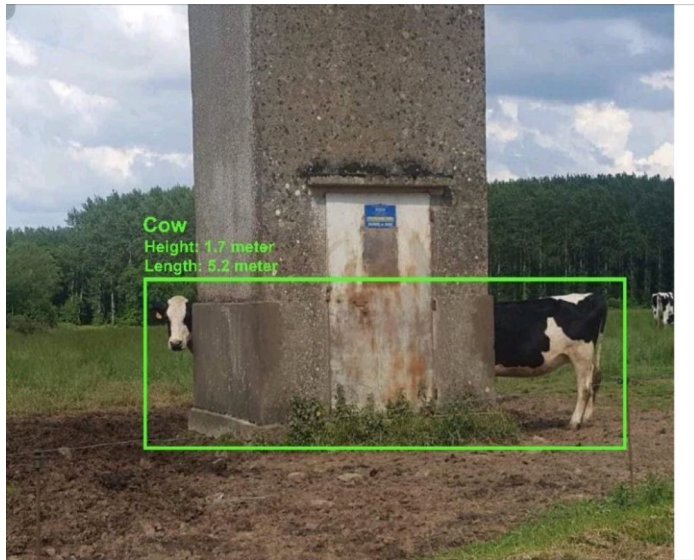


<https://www.cs.umd.edu/~tomg/projects/invisible/>

Plausibility & Context

 **Pascal BORNET** • 3rd+
LinkedIn Top Voice in Tech | ...
1mo • 🌐 [+ Follow](#)

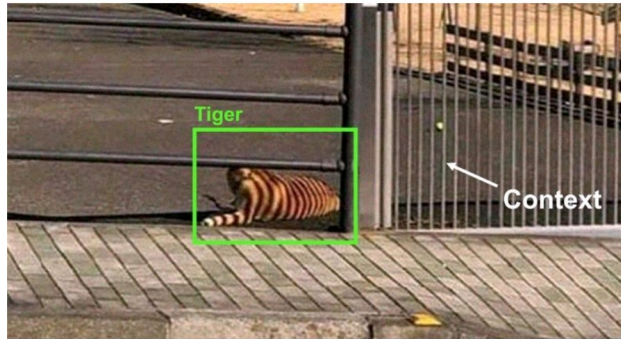
The longest cow in the world 😊...
Or how to deceive an AI program!... see more



Source: LinkedIn

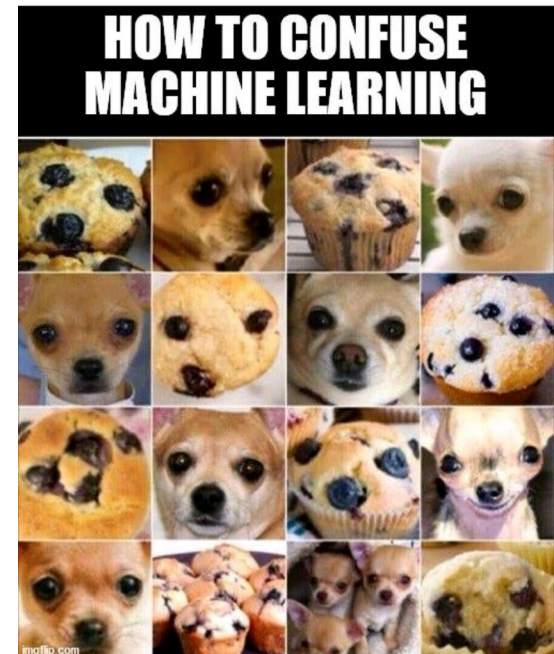
 **Pascal BORNET** • 3rd+
LinkedIn Top Voice in Tech | Keynote ...
1d • 🌐 [+ Follow](#)

Data without context is just useless and misleading!
😊... see more



 **avyana**
324 followers
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How to Confuse Machine Learning... see more



Spectacular AI-Fails



ACTIONS   **NEW YORK POST**

TECH     

Chinese users claim iPhone X face recognition can't tell them apart

By Guy Birchall, Tom Michael, The Sun
Published Dec. 21, 2017 | Updated Dec. 21, 2017, 4:02 p.m. ET



The Verge / Tech / Reviews / Science

MICROSOFT / WEB / TL:DR

Twitter taught Microsoft's AI chatbot to be a racist asshole in less than a day



Business U.S. Politics Economy Tech Finance Opinion Arts & Culture Lifestyle Real Estate Personal Finance

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Google Mistakenly Tags Black People as 'Gorillas,' Showing Limits of Algorithms

Alistair Barr [Follow](#)
Updated July 1, 2015 3:41 pm ET

Areas of application

Machine Learning (esp. Neuronal Networks):

- Images, sound, language, art
- Rules that cannot be clearly described (e.g. marketing, trends, psychology)
- Lots of (cheap) data
- Rapidly changing system (e.g. social media)
- Individual errors are no big deal

Model Based Algorithms:

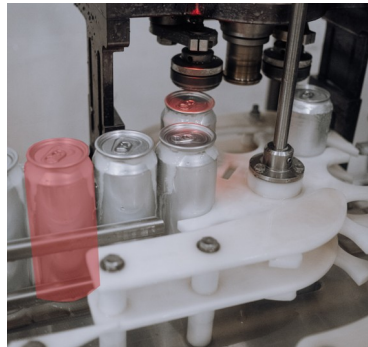
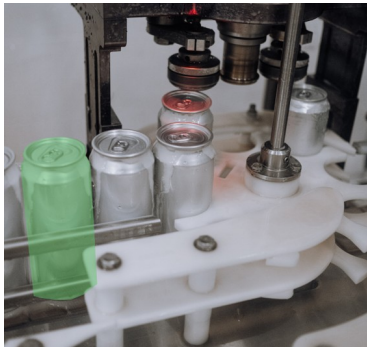
- Extremely reliable
- Data points very expensive
- Clear rules in the system (e.g. physics, mathematics)

Hybrid Solutions: e.g. visual inspection

Bad:

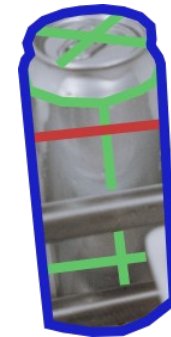
Directly classify as good/bad

- Lot of training
- Lot of bad examples needed
- New defects are not detected



Good:

- Object detection by NN / YOLO
- Testing according to physical criteria (color gradient, light reflection, ...)

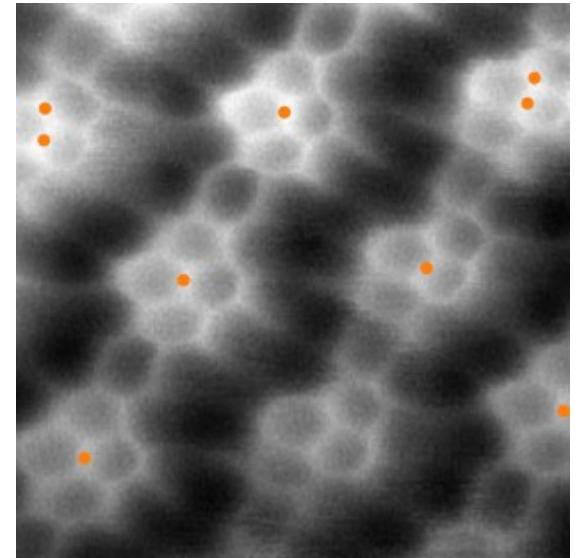


Sensor Tuning with AI

- Topography by laser distance measurement:
gray level = height
- **Task:** reconstruct color information from real object
- Experience:
strong measurement artifacts at certain spots

Solution:

- Neural Network to construct color at specific spots
- Precision: 80% (requirement fulfilled)
- 1-2 weeks work
- No system understanding
- No domain transfer investigated



<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC4050271/>

Natural Language Processing

Vor einigen Wochen ließ ich eine Darmspiegelung (zur Vorsorge) durchführen. Nachdem ich das Thema seit mehreren Monaten verschoben hatte, war ich überrascht, wie schnell, taktvoll und komplikationslos alles verlaufen ist. Die Angst war völlig unbegründet!

woch liess darmspiegel vorsorg durchfuhr nachd thema seit mehr monat verschob uberrascht schnell taktvoll komplikationslos verlauf angst vollig unbegrundet

vector with 100 000 entries (most of them 0)
[0,00; 0,20; 0,00; 0,03; 0,00; 0,00; 0,71; ...]

Rating: 2
German scale:
1 best, 6 worst

Guy doing colonoscopy:
reads like rating 1

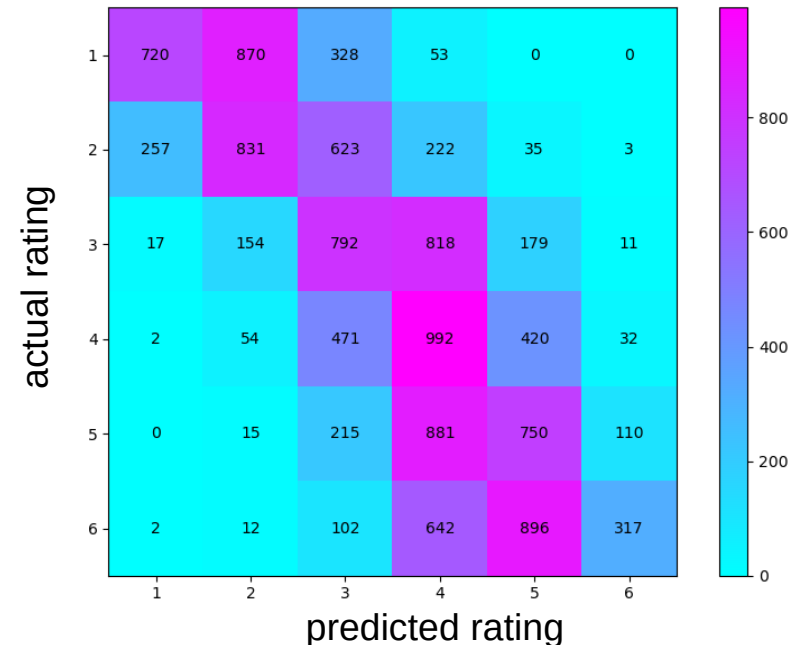
Natural Language Processing

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Natural Language Processing

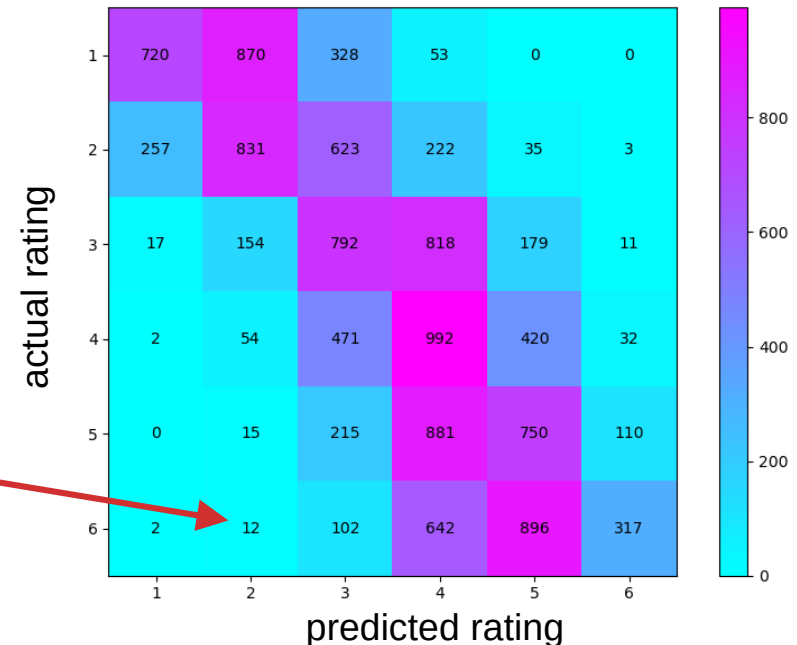
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vector with 100 000 entries (most of them 0)
[0,00; 0,20; 0,00; 0,03; 0,00; 0,00; 0,71; ...]

Great doctor! Thanks for everything
[translated from German]

Rating: 2
German scale:
1 best, 6 worst



Clustering

Mostly intermediate step in AI projects

- Gaining an overview
- Fewer data points for training
- Identification of outliers

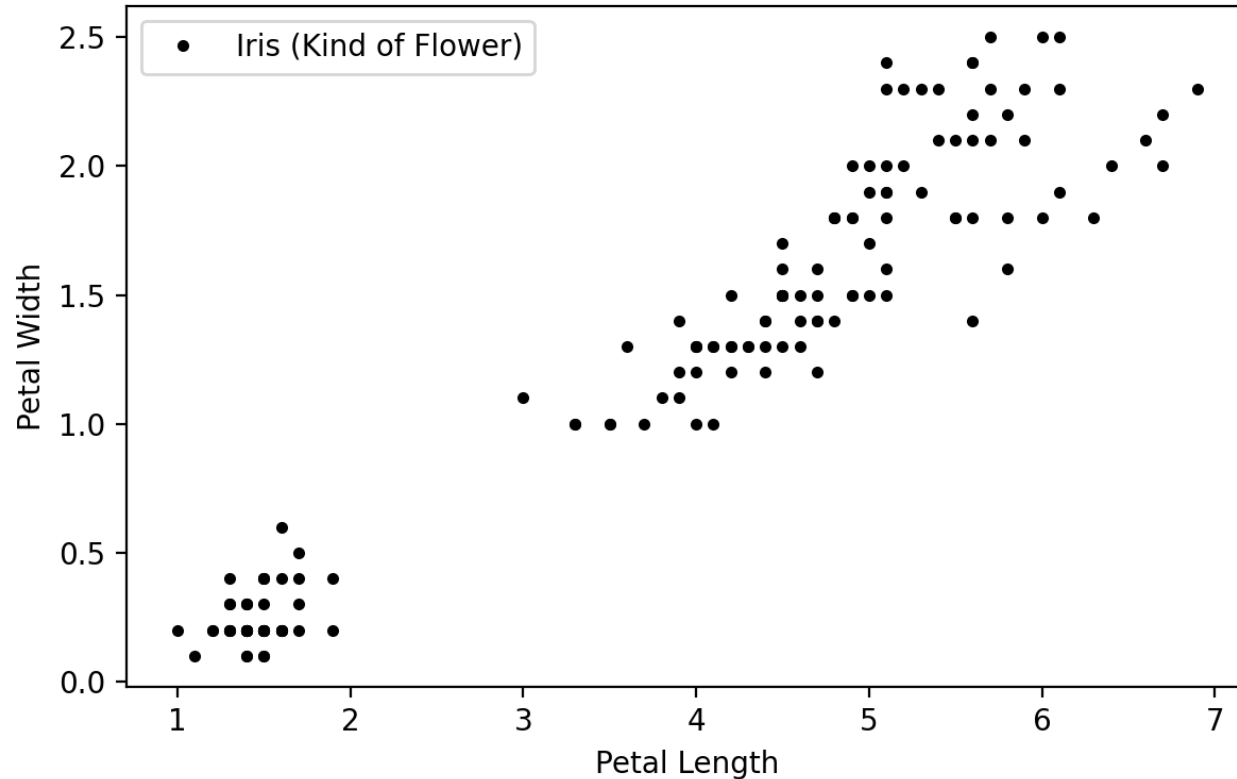
Example:

Quality fluctuations in the processing of irises (kind of flower) into cosmetics

Clustering

Idea: analysis of processed irises

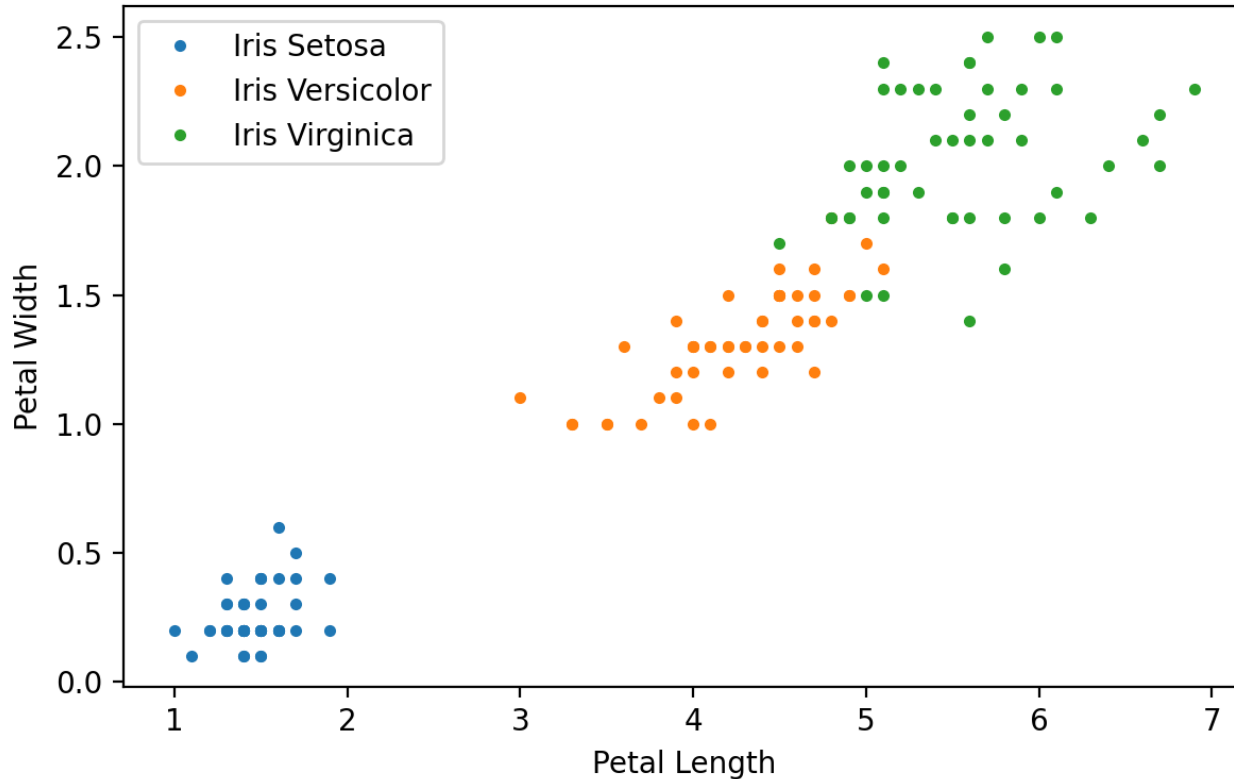
Data: sepal/petal + length/width



Source: Iris flower data set
https://en.wikipedia.org/wiki/Iris_flower_data_set

Clustering

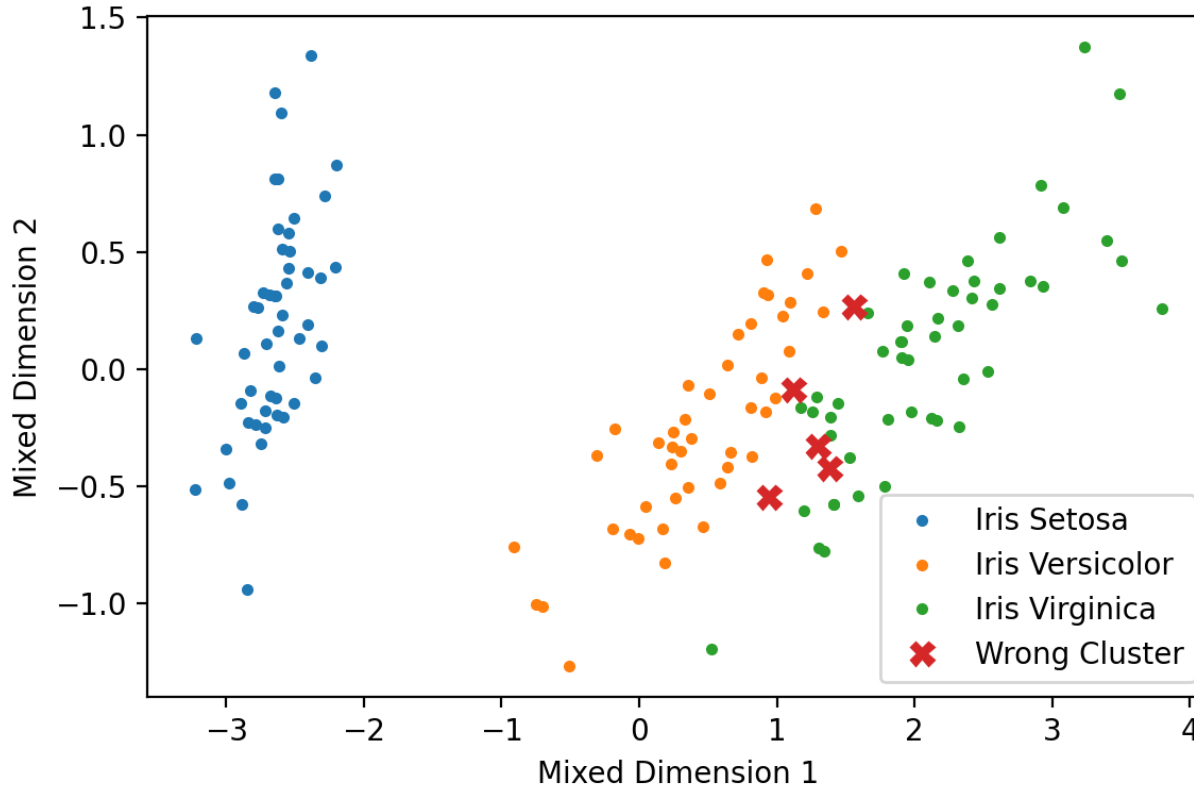
Approach: species dependent processing



Source: **Iris flower data set**
https://en.wikipedia.org/wiki/Iris_flower_data_set

Clustering

Next step: expensive determination of species at boundary



Source: **Iris flower data set**
https://en.wikipedia.org/wiki/Iris_flower_data_set

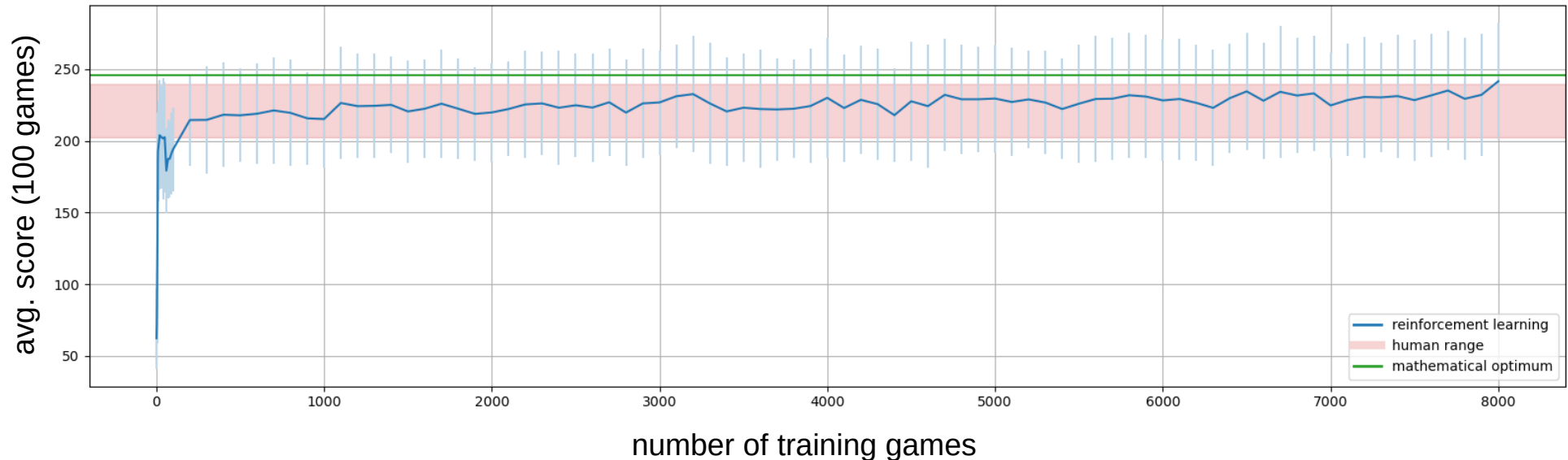
Reinforcement Learning

Current situation, control signal ► final score

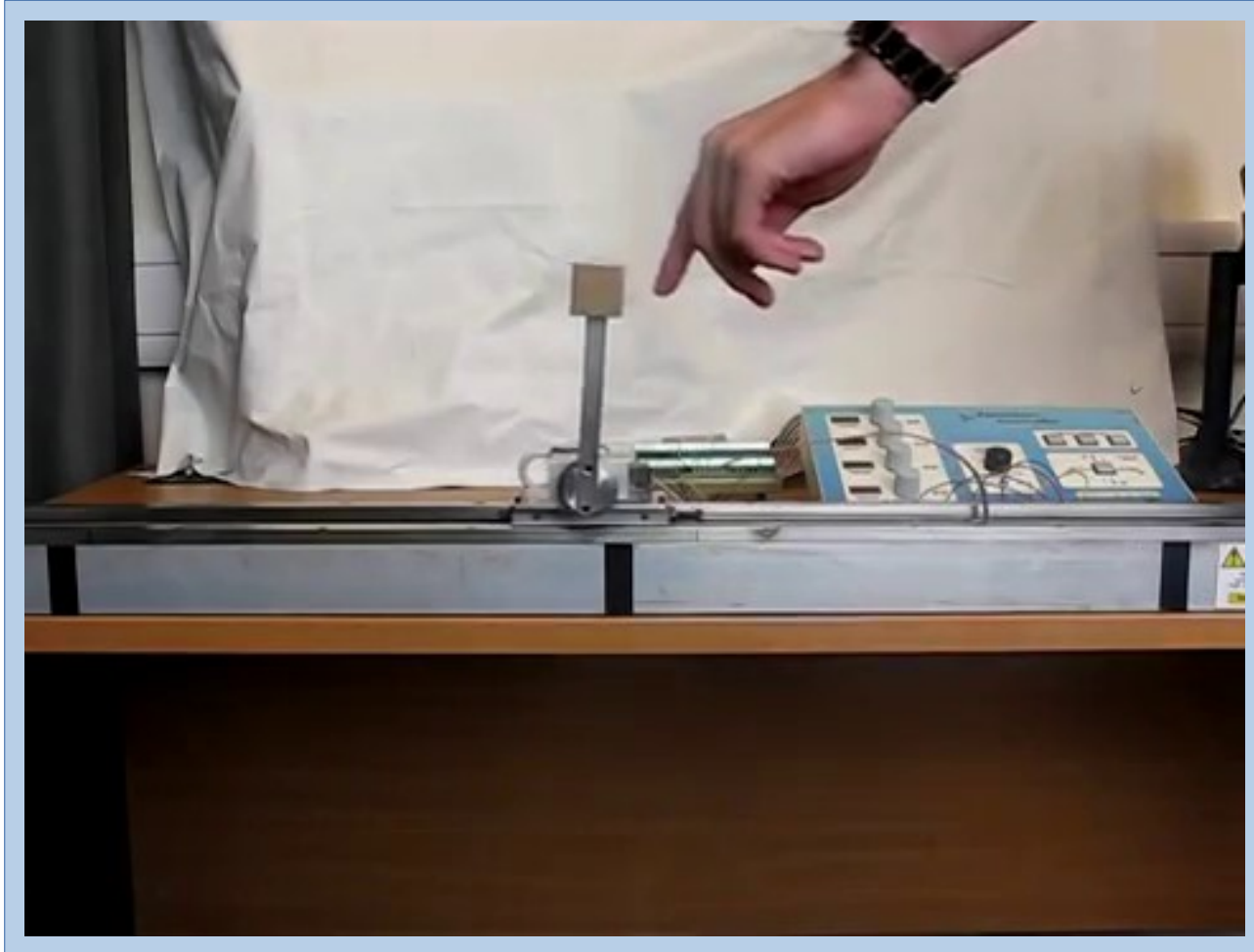
e.g. dynamic pricing,

e.g. compensation of raw material fluctuations in production

AI learns Yahtzee



Reinforcement Learning



PilcoLearner: Cart-Pole Swing-up
<https://youtu.be/XiigTGKZfks>

Kind regards



Algorithmus Schmiede

Data Science | Numerik | Physik



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Inhaber

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