

Demystifying AI in the math classroom:

Exploring machine learning concepts with high-school students

Stephan Kindler, Sarah Schönbrodt, Martin Frank



The era of AI

Students should be empowered to critically, reflectively, and creatively engage with AI technologies.

This requires early AI education on opportunities and design possibilities, limitations, and the functioning of AI.

UNESCO, 2024; European Commission, 2022; Long & Magerko, 2020, Touretzky et al., 2022

AI education in mathematics education

Reasonable? Possible? Even necessary?

- Data-driven ML methods are **based on mathematics and data!**
mathematics education plays a crucial role in fostering a deeper understanding of AI technologies
- **Authentic** insight into ML methods is possible!
connection to school mathematical contents (vectors, dot product, Euclidean distance, functions, lines, planes, ...)
- Variety of suitable **real and relevant problems** exists!
accessibility, availability of data

Math education is an essential component for developing AI skills.

Teaching machine learning in mathematics education

- Decision trees (Biehler & Fleischer, 2021)
- K-nearest neighbour (Hazzan & Mike, 2022; Bata, 2024; Schönbrodt et al., 2024)
- Support Vector Machine (Schönbrodt et al., 2025)
- Artificial Neural Networks (Kindler et al., 2024)
- Linear regression (Biehler et al., 2024)
- N-Gram model (Hofmann & Frank, 2022)
- ...

Starting point: data

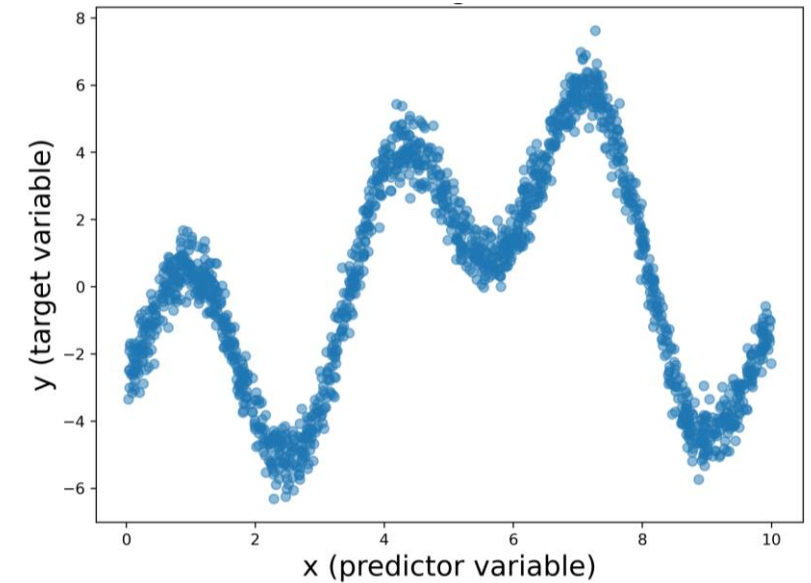
Supervised learning

Input data: (predictor variables):

$$x_1, \dots, x_N \in \mathbb{R}^n$$

Output data: (values of the target variables):

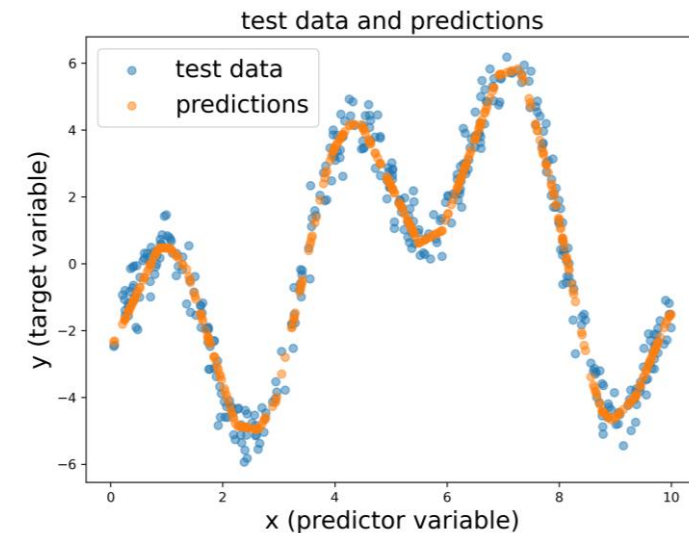
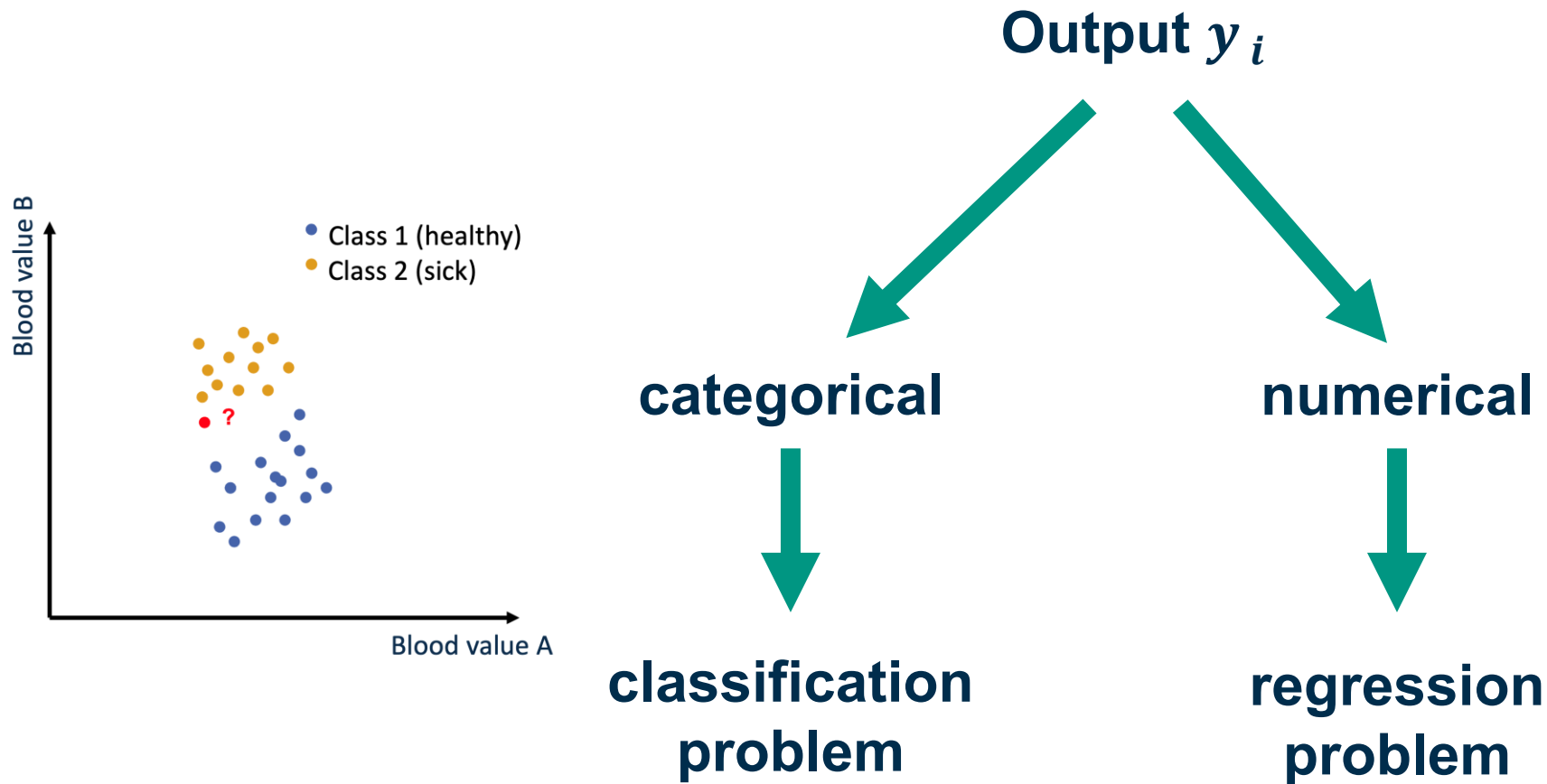
$$y_1, \dots, y_N \in \mathbb{R}$$



Goal: mathematical model that best describes the unknown relationship between input and output data

→ a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$, such that predictions $\hat{y}_i := f(x_i)$ are close to the known outputs y_i .

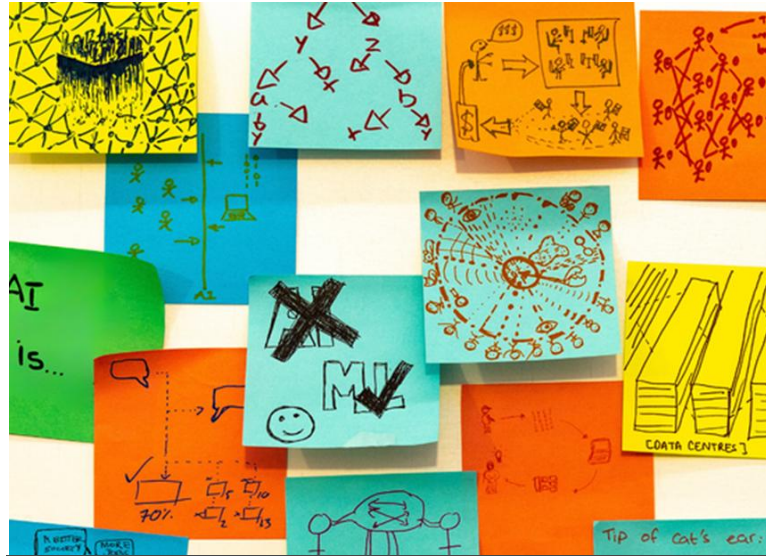
Two main problem classes



Our focus today



A workshop on image classification and the Support Vector Machine



A learning path to better understand Artificial Neural Networks



Conclusions about AI education and ways to implement

CAMMP

Computational And Mathematical Modeling Program



Goal:

Allow students to solve
authentic, real and **relevant** problems
using mathematical modeling and
computers

- Teaching and learning material on AI (and beyond!)
- Workshops for students and teachers
- Locations: Aachen, Karlsruhe, Stuttgart (Germany), Salzburg (Austria)



mathematical modeling = use of math to
solve a real-world problem

AI workshops (blue = ML-method)

Privacy in social networks

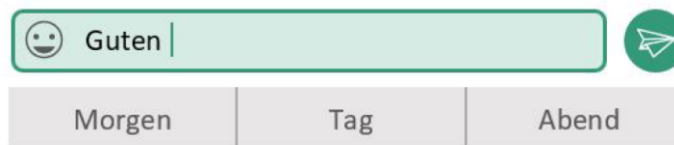
Decision Trees



Sube (2019)

Word predictions

N-Gram model



Hofmann (2022)

Topics

Recommender systems

k-nearest-neighbour

Netflix
Challenge



Personalized
recommendations



Schönbrodt et al. (2021)

Human activity recognition

k-nearest-neighbour

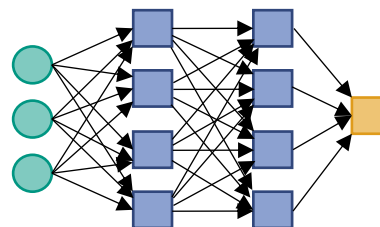


Running? Walking? Climbing stairs?

Schönbrodt et al. (2024)

Predicting life expectancy

lin. Regression & neural networks



Kindler et al. (2024)

(Image) Classification

Support Vector Machine



oder



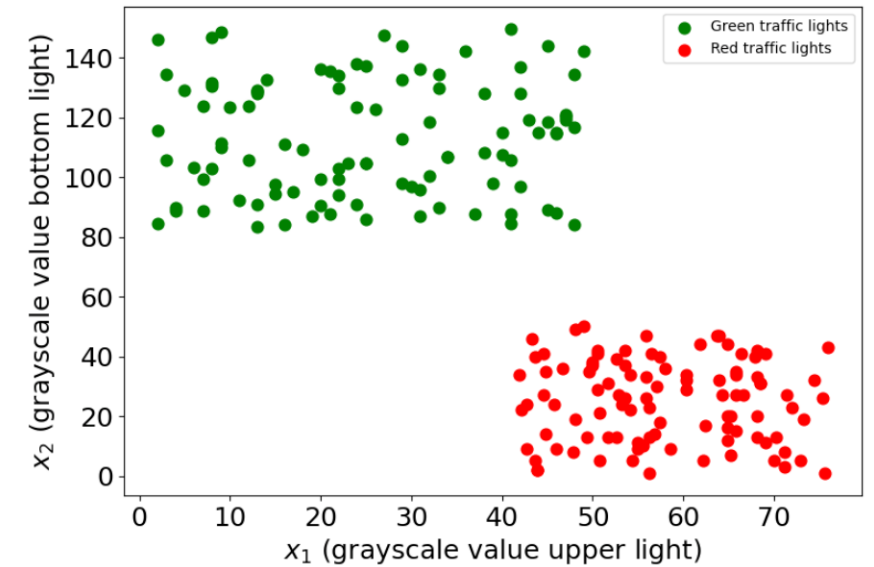
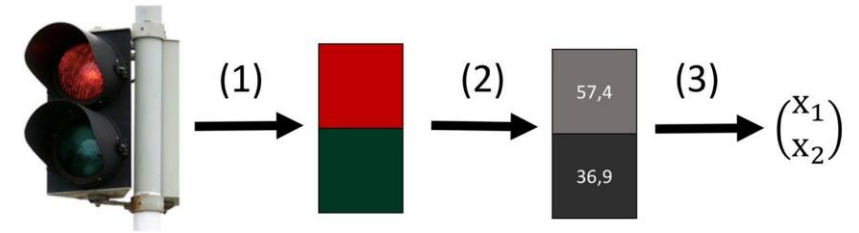
Ampel steht auf rot Ampel steht auf grün

Schönbrodt et al. (2021)

Classification problems

The Support Vector Machine – upper secondary education

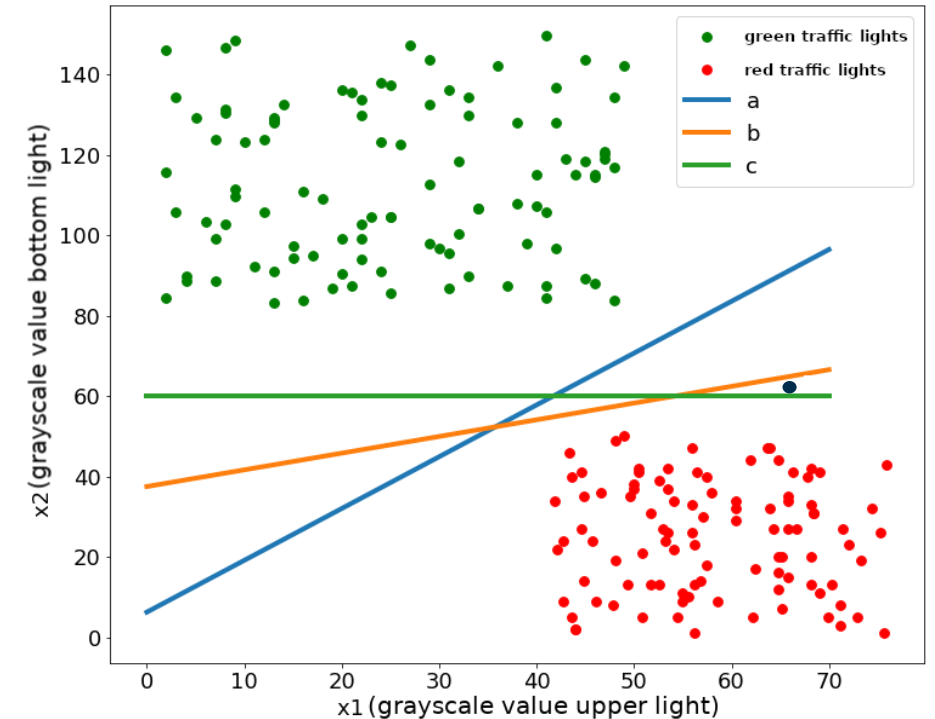
- Data points of two classes



Classification problems

The Support Vector Machine – upper secondary education

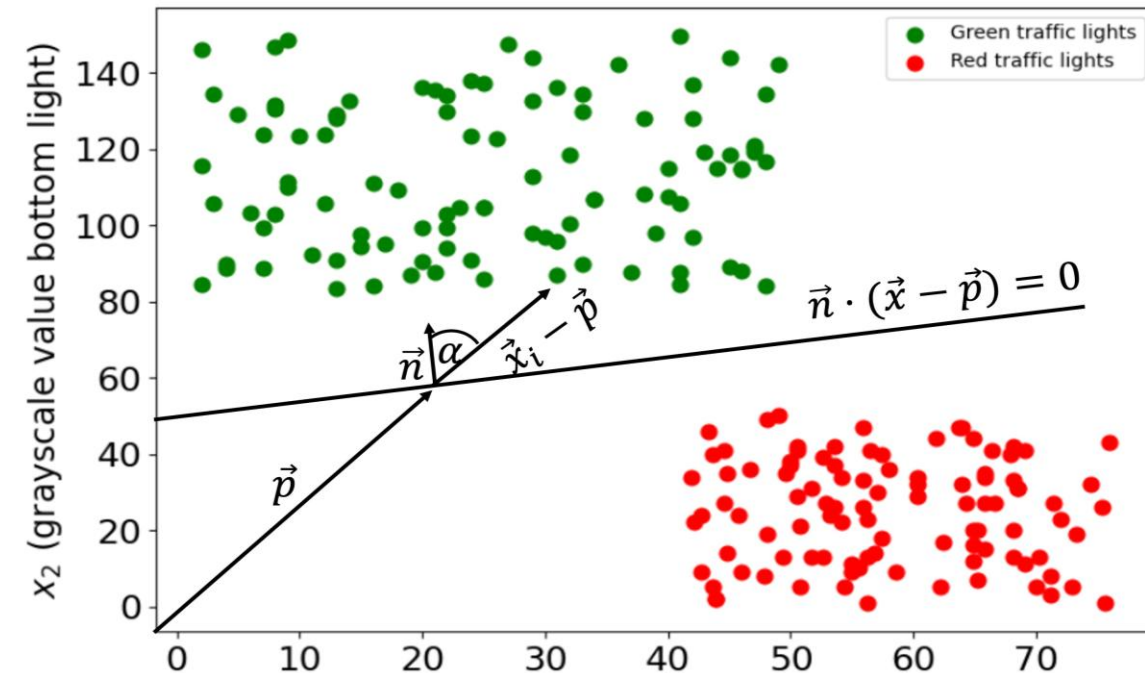
- Data points of two classes
- Find a **line**, that best separates the data points of both classes.



Classification problems

The Support Vector Machine – upper secondary education

- Data points of two classes
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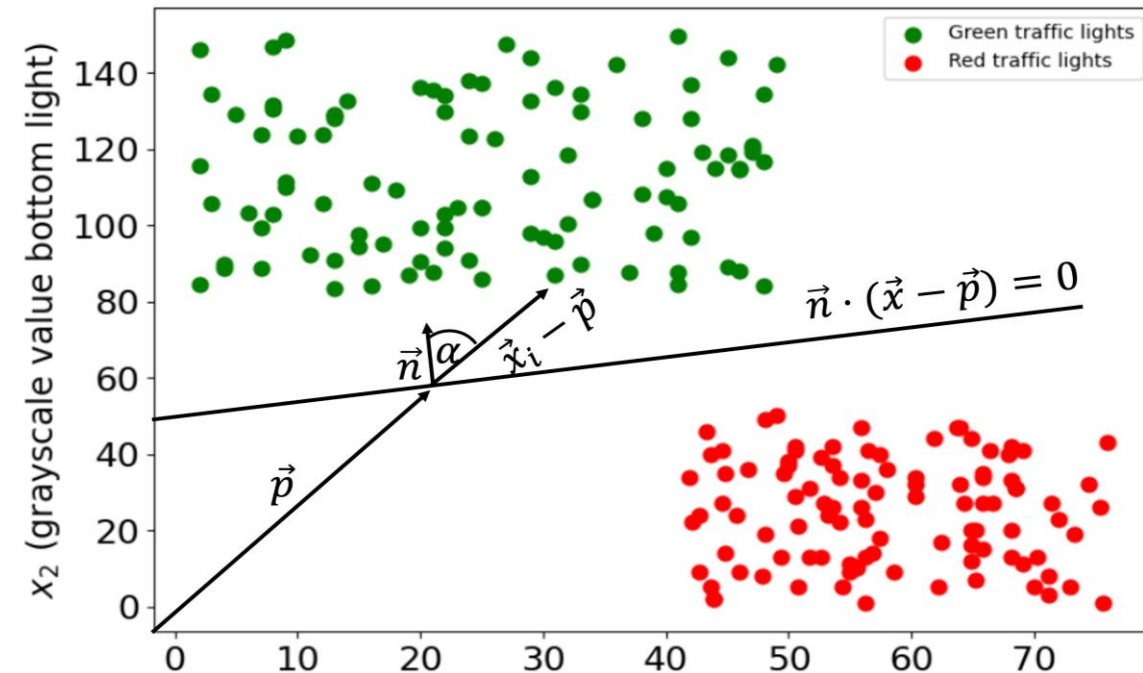


Classification problems

The Support Vector Machine – upper secondary education

- Data points of two classes
- Find a **line**, that best separates the data points of both classes.
- Define „best“.

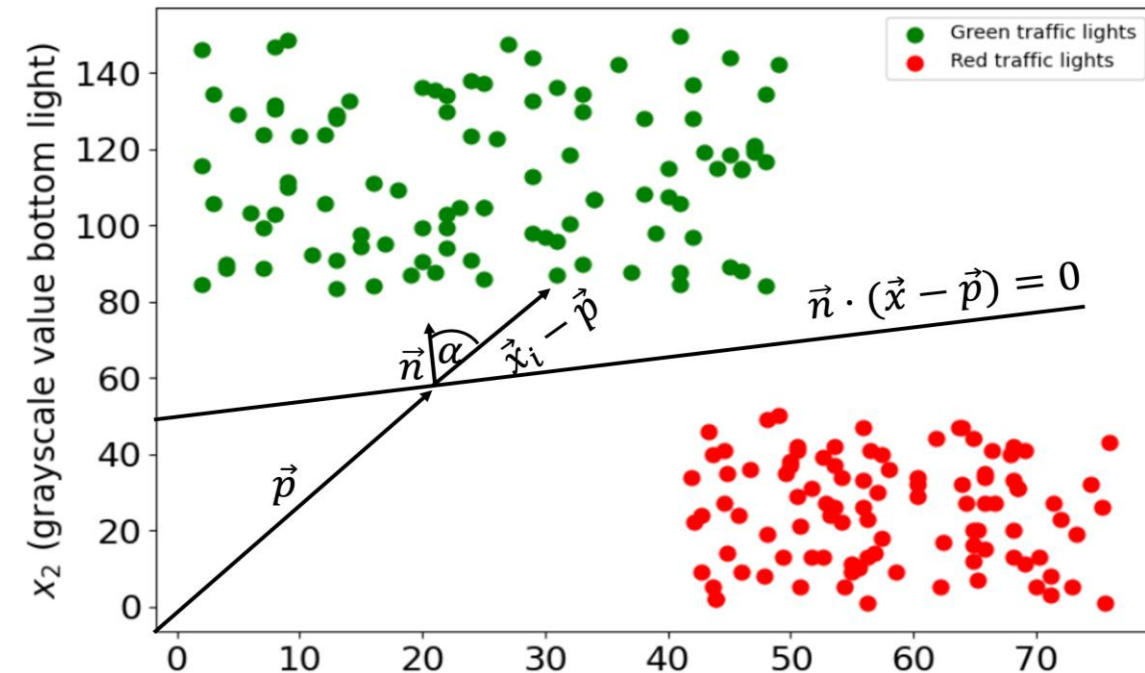
Select line with maximum distance to the nearest data points!



Classification problems

The Support Vector Machine – upper secondary education

- Data points of two classes
- Find a **line**, that best separates the data points of both classes.
- Define „best“.
- Classify data points according to their position in relation to the **separating line**.



Classification problems

The Support Vector Machine – upper secondary education

- Data points of two classes
- Find a **line**, that best separates the data points of both classes.
- Define „best“.
- Classify data points according to their position in relation to the **separating line**.
- Validate the model statistically.

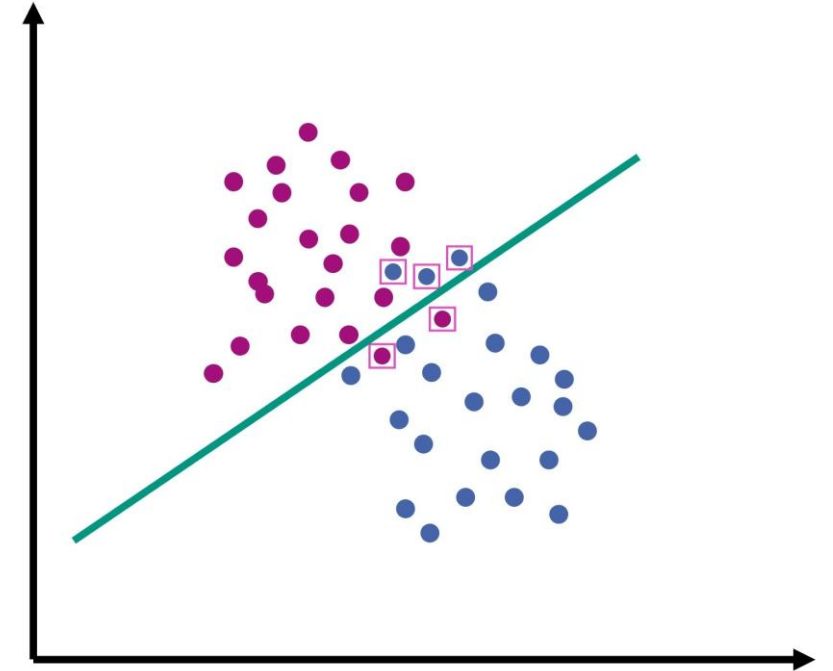
	Predicted class 1	Predicted class 2
Actual class 1	350	80
Actual class 2	20	550

Accuracy: 90%

Classification problems

The Support Vector Machine – upper secondary education

- Data points of two classes
- Find a **line**, that best separates the data points of both classes.
- Define „best“.
- Classify data points according to their position in relation to the **separating line**.
- Validate the model statistically.
- Discuss limitations.

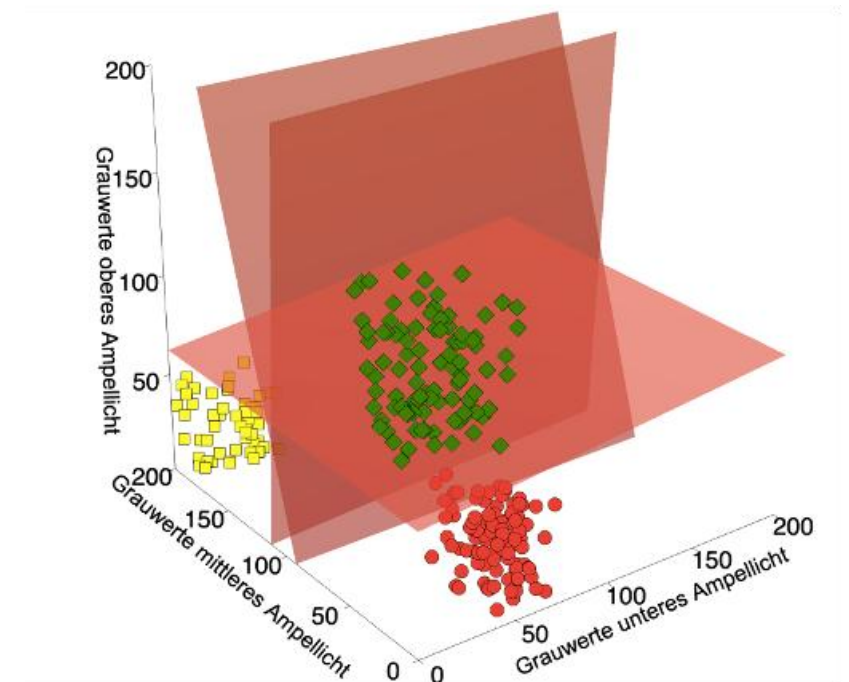
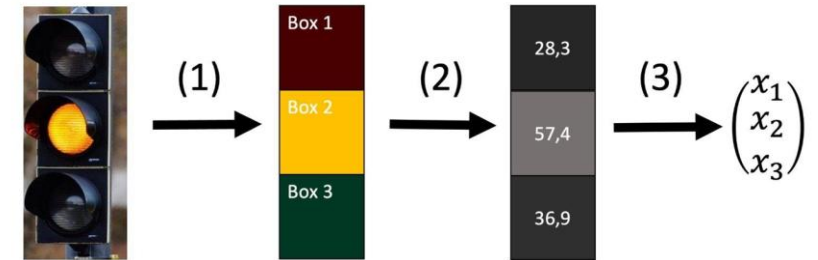


Classification problems

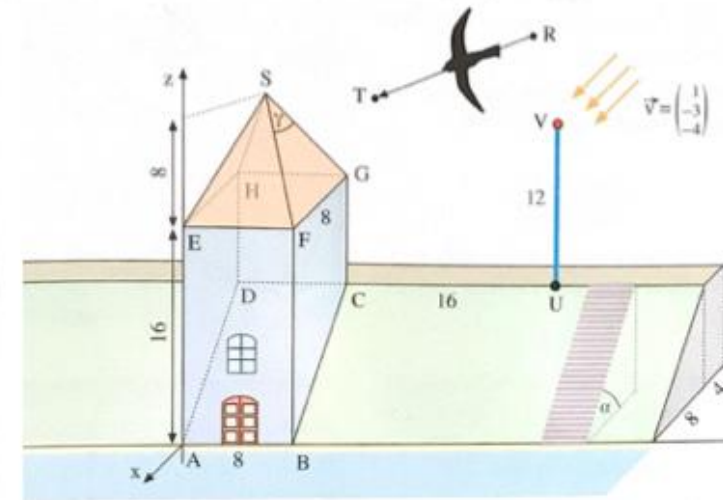
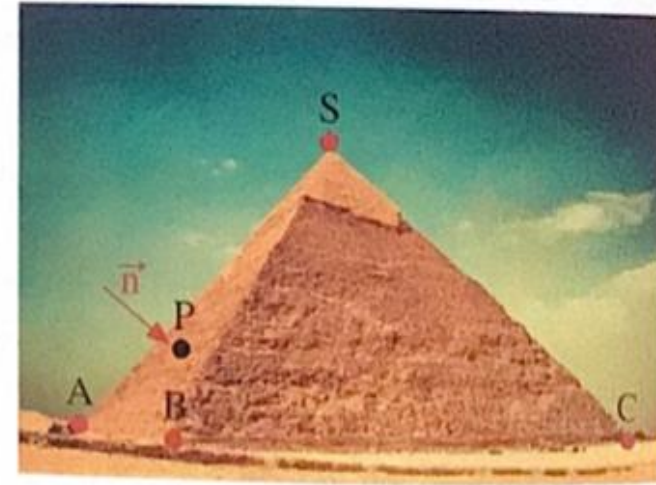
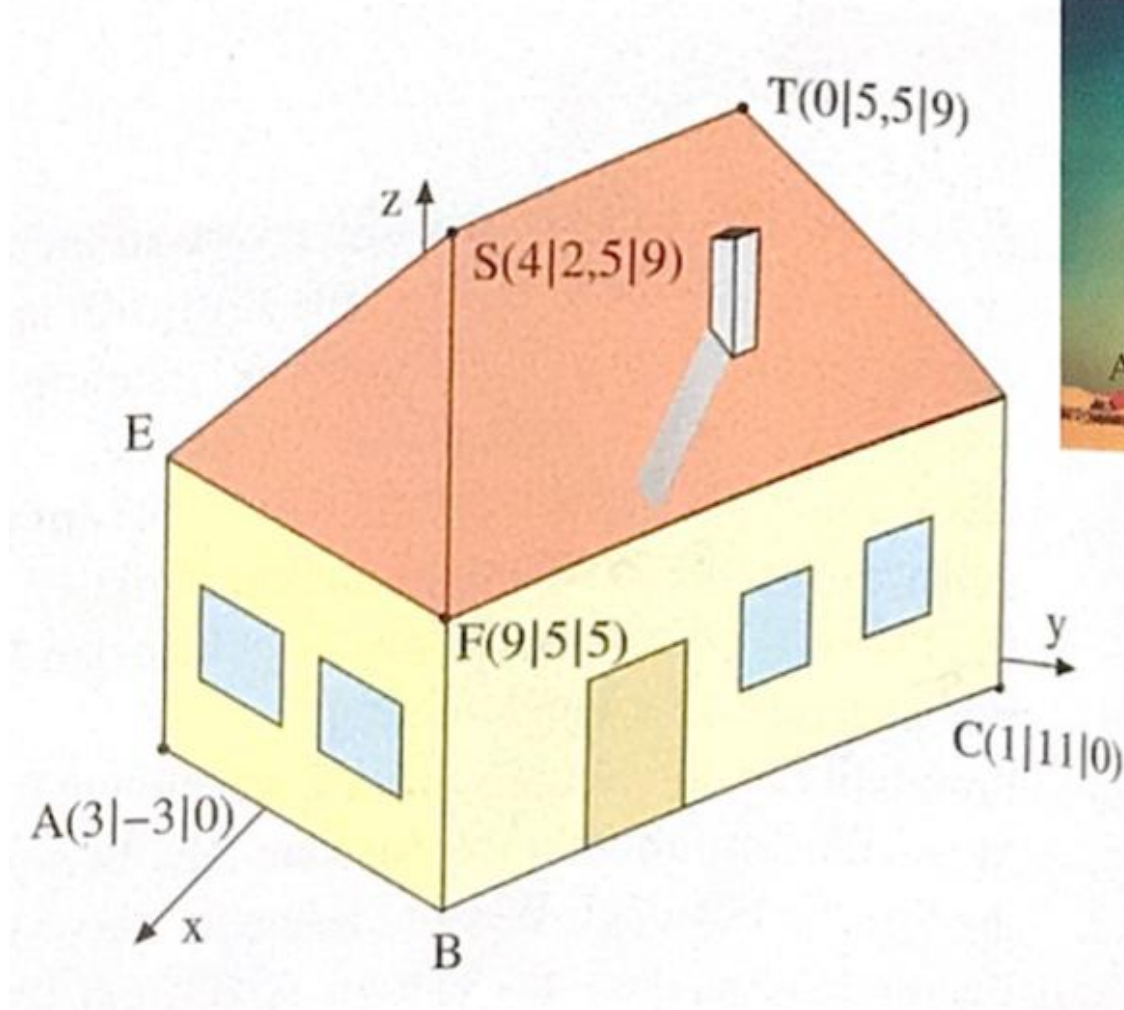
The Support Vector Machine – upper secondary education

- Data points of three classes
- Find a **plane**, that best separates the data points of three classes.
- Define „best“.
- Classify data points according to their position in relation to the **separating plane**.

*lines, planes, distances, dot product,
statistical measures*



A glance at the schoolbook



Images from: Bigalke / Köhler. Mathematik Gymnasiale Oberstufe Nordrhein-Westfalen
Qualifikationsphase Grundkurs Cornelsen, 2014.

Critical discussions

... in the mathematics classroom!



- Bias, discrimination, fairness
- Diversity of data
- Privacy and data protection
- Measurement errors; outliers

Digital Learning material

... based on Jupyter Notebooks

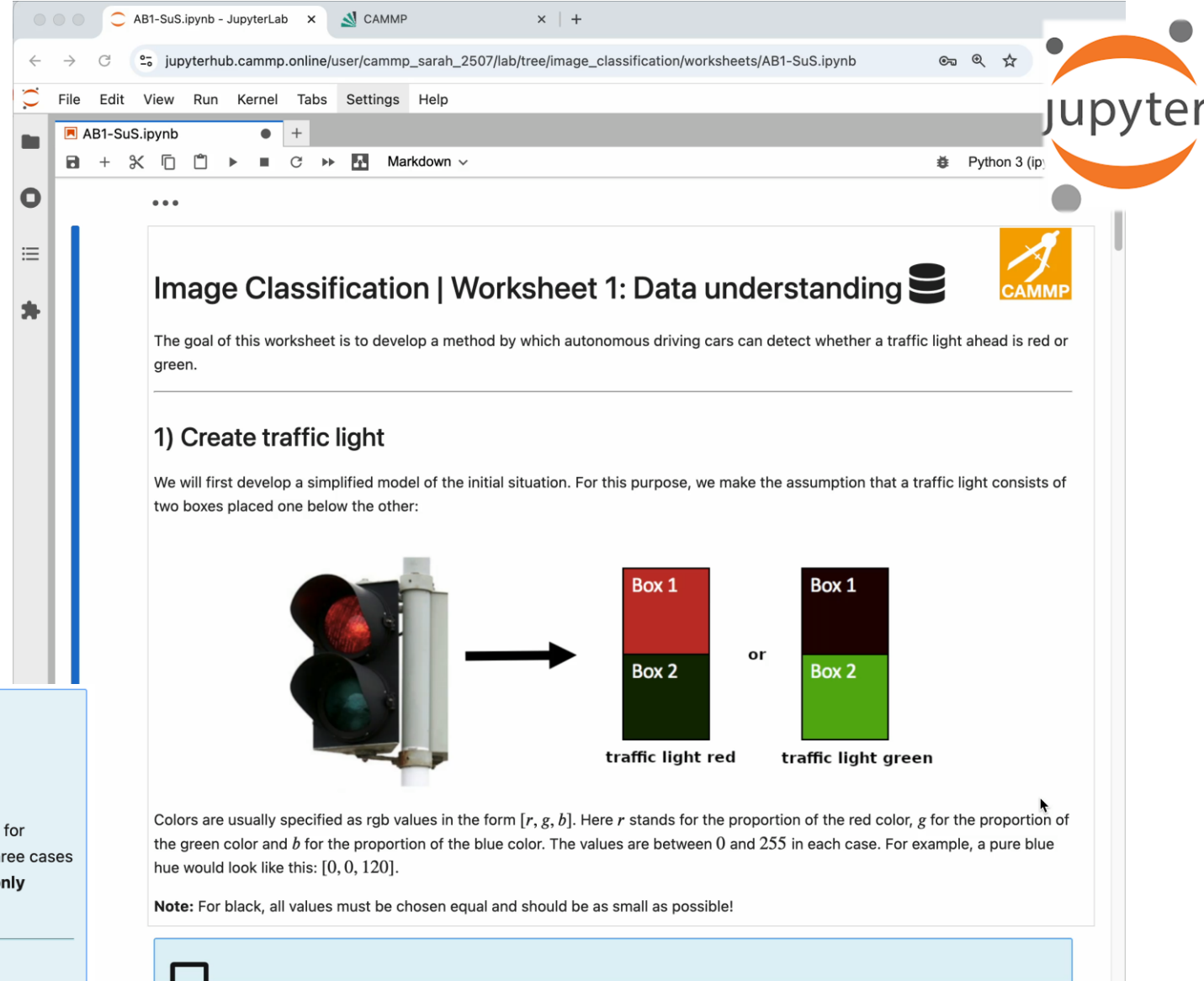
- No programming skills needed (“fill in the gap” approach)
- No software installation
- Direct feedback
- In-depth tasks
- Staggered hints



a) Case distinction in the formula

Write on the [answer sheet](#) a general distinction of cases for formula $\vec{n} * (\vec{x} - \vec{p}) = \cos(\alpha) \cdot |\vec{n}| \cdot |\vec{x} - \vec{p}|$ for the three cases 1-3 shown above. Think why it is sufficient to consider **only angles between 0° and 180°!**

 Hint



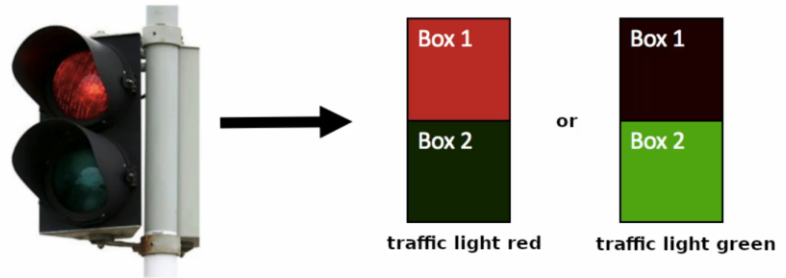
The screenshot shows a Jupyter Notebook interface with a browser window. The URL is `jupyterhub.cammp.online/user/cammp_sarah_2507/lab/tree/image_classification/worksheets/AB1-SuS.ipynb`. The notebook is titled "AB1-SuS.ipynb" and is in "Markdown" view. The content of the notebook is as follows:

Image Classification | Worksheet 1: Data understanding

The goal of this worksheet is to develop a method by which autonomous driving cars can detect whether a traffic light ahead is red or green.

1) Create traffic light

We will first develop a simplified model of the initial situation. For this purpose, we make the assumption that a traffic light consists of two boxes placed one below the other:



Colors are usually specified as rgb values in the form $[r, g, b]$. Here r stands for the proportion of the red color, g for the proportion of the green color and b for the proportion of the blue color. The values are between 0 and 255 in each case. For example, a pure blue hue would look like this: $[0, 0, 120]$.

Note: For black, all values must be chosen equal and should be as small as possible!

Access the teaching and learning material

Project Website: www.cammp.online/english

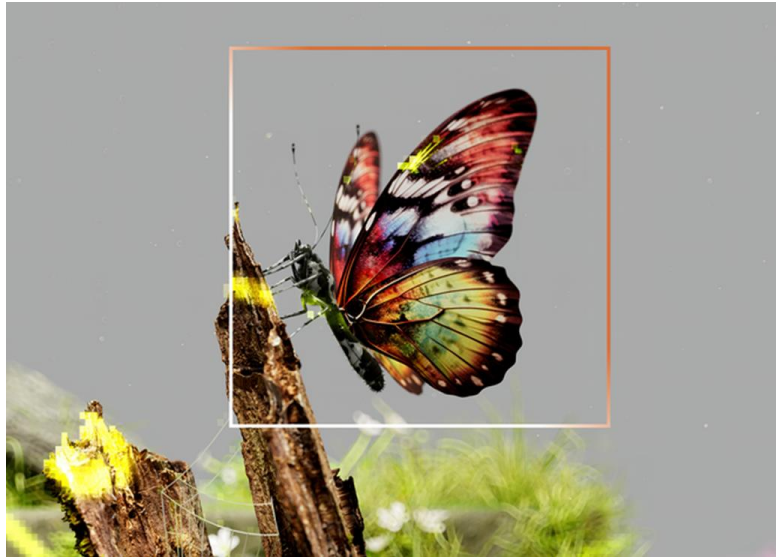
Learning material (based on Jupyter Notebooks): <https://jupyterhub.cammp.online/>

Useraccount: cammp_YOURPSEUDONYM
[prefix cammp_ is important]

Password: [can be determined freely]

Simple Simulation of an ANN (based on GeoGebra):
<https://www.geogebra.org/calculator/kpjk8qb6>

Our focus today



A workshop on image classification and the Support Vector Machine



A learning path to better understand Artificial Neural Networks



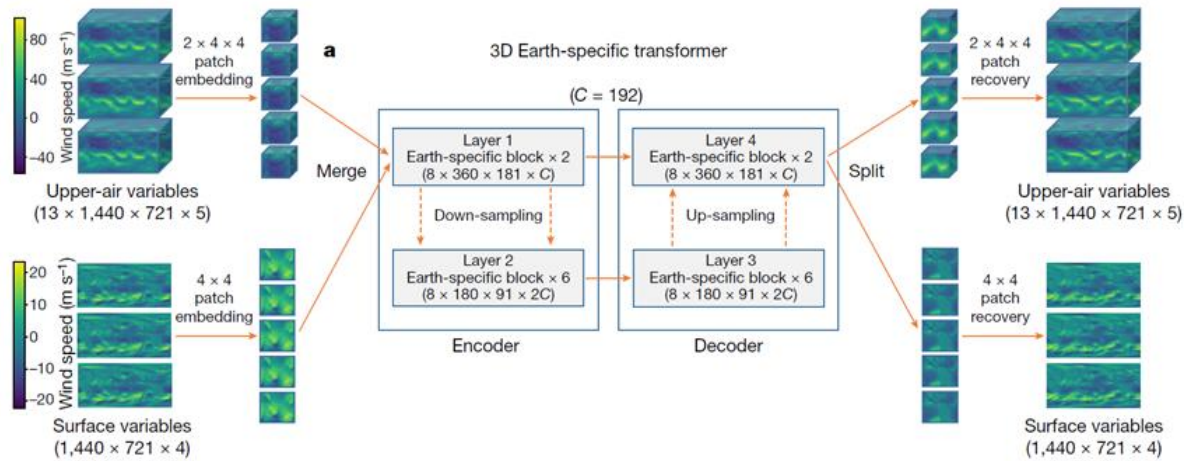
Conclusions about AI education and ways to implement

Artificial neural networks are big and powerful

- ANN basis of many AI applications (GenAI like Chatbots, image generation, weather forecast, ...)
- Current ANN models are complex and very large
- Focus on regression models rather than GenAI models: e.g. weather forecast

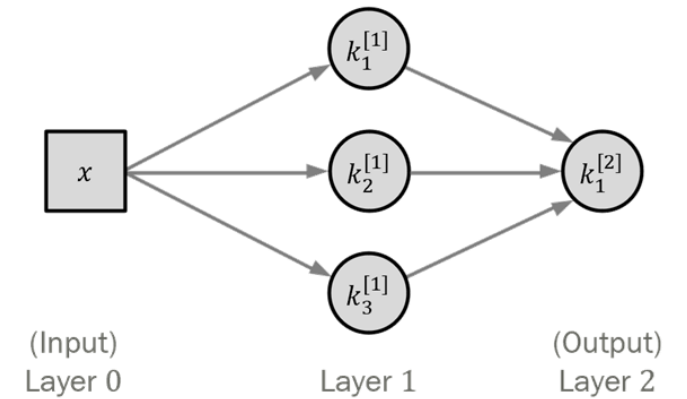


A current ANN structure



Architecture of the Pangu-Weather model (Bi et al., 2023)

Simplified to

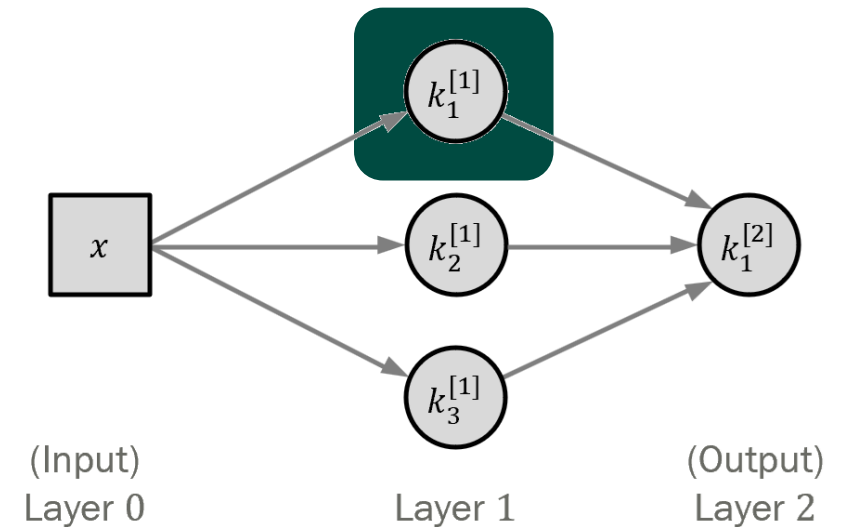


A toy ANN to tackle misconceptions

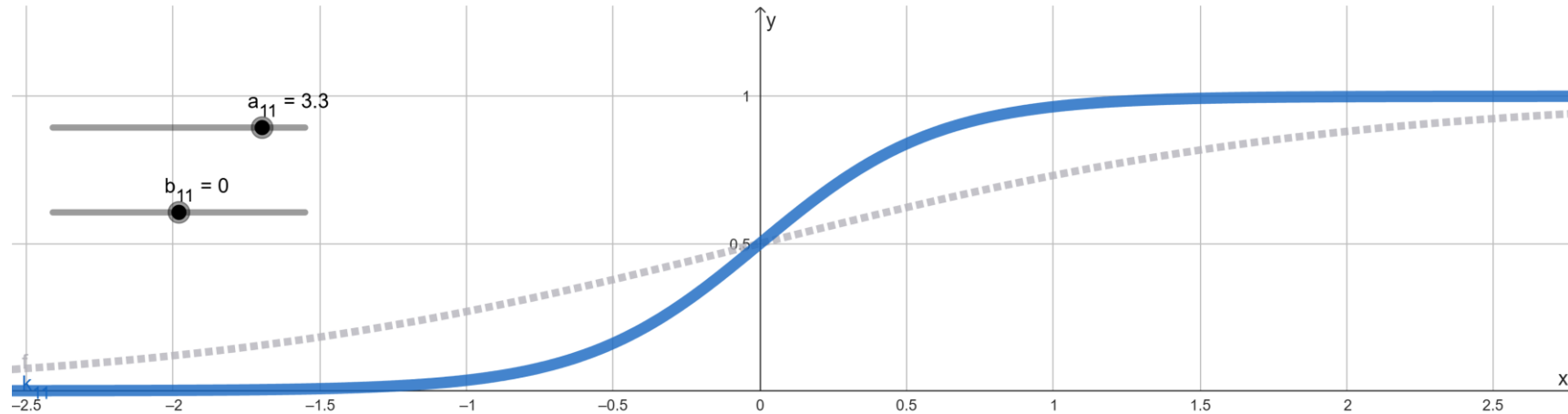
The ANN...

- “... *learns physics*”
- “... *recognizes patterns*”
- “... *also works on data it has not seen yet*”
- “... *is way to complex to understand*”
- ...

⇒ Explore the functionality of this toy network to demystify misconceptions



One node of an ANN



$$k(x) = \sigma(w \cdot x + b)$$

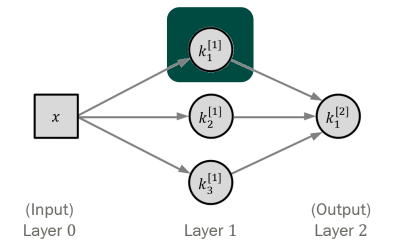
$$= \frac{1}{1 + e^{-(w \cdot x + b)}}$$

Linear function

Linear function with parameters $w, b \in \mathbb{R}$

Activation function

Non-linear function σ with $\sigma(x) = \frac{1}{1+e^{-x}}$

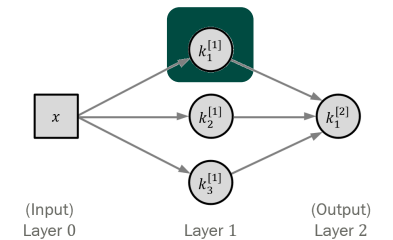
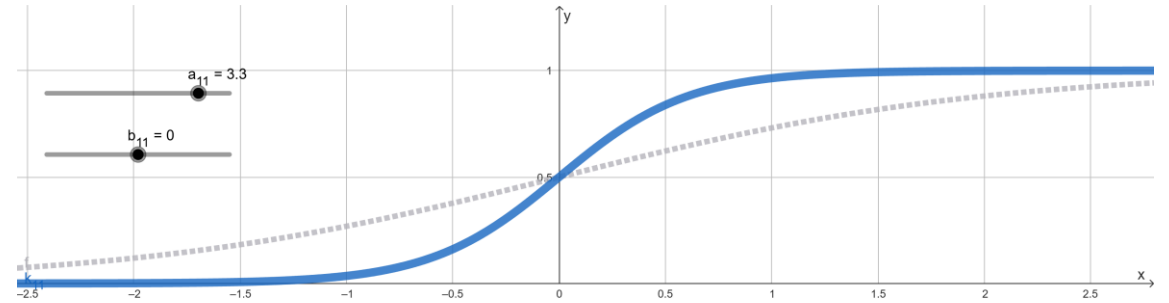


Kurvendiskussion of one node

Students can use mathematical tools to analyse this function

- Domain and codomain
- Limits of $k(x)$ for $x \rightarrow \pm\infty$
- Finding extremes
- Influence of the parameters w and b

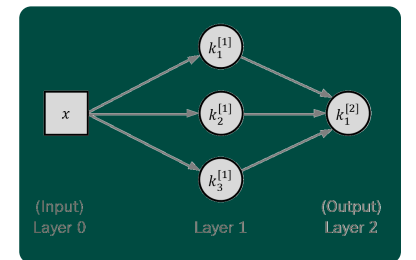
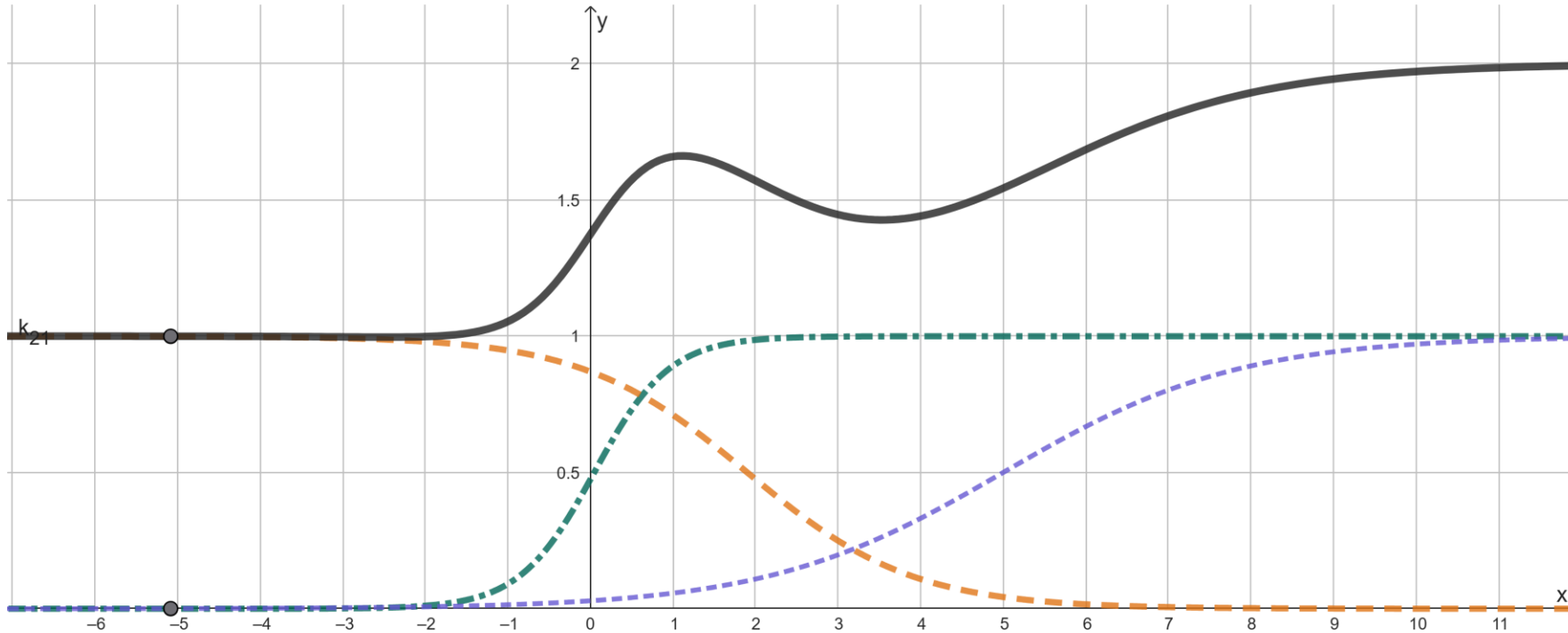
$$k(x) = \frac{1}{1 + e^{-(w \cdot x + b)}}$$



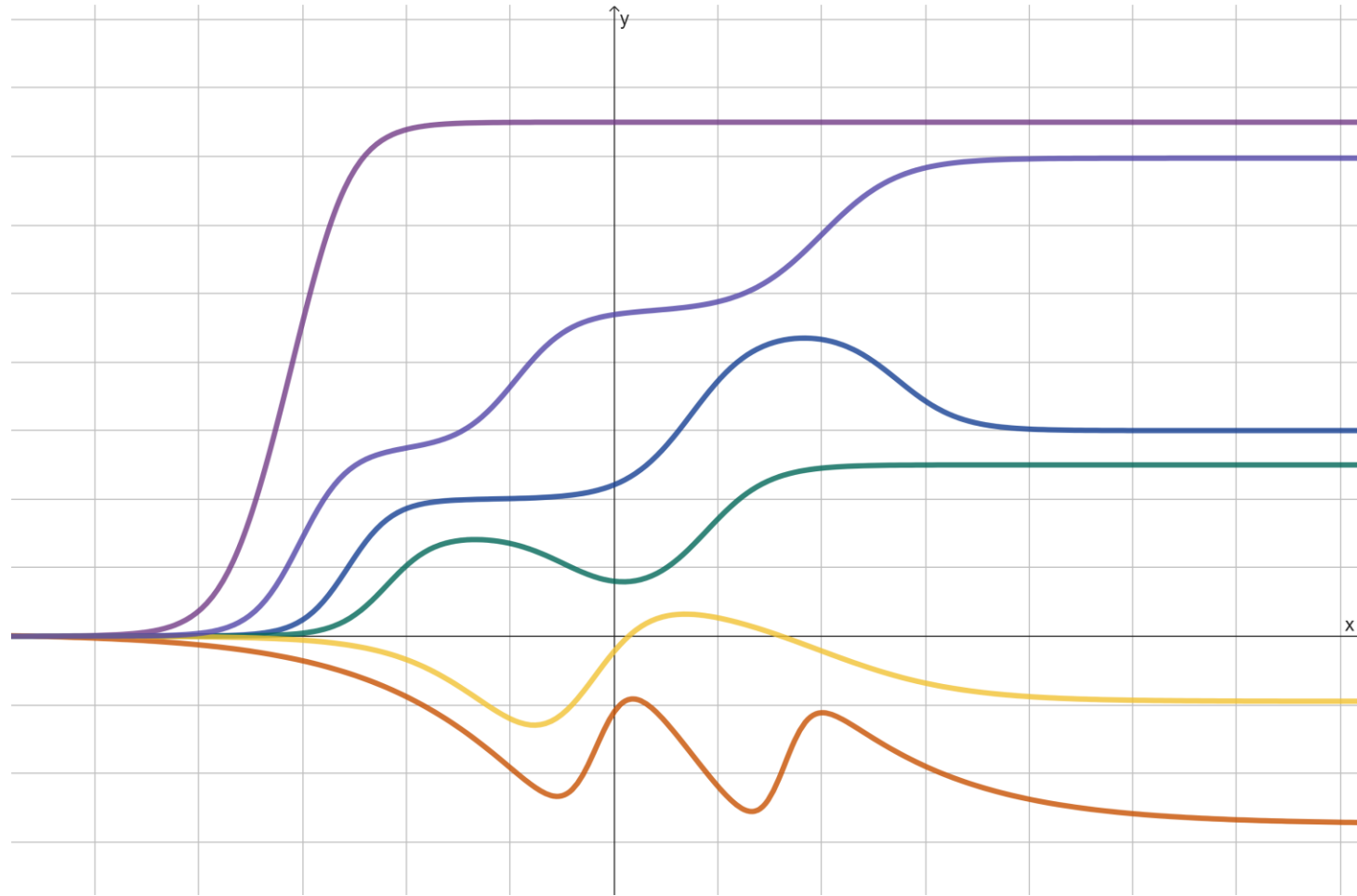
The toy network

The network is the weighted sum of the previously considered functions

$$K(x) = w_1^{[2]} \cdot k_{1,1}(x) + w_2^{[2]} \cdot k_{1,2}(x) + w_3^{[2]} \cdot k_{1,3}(x) + b_{2,1}$$

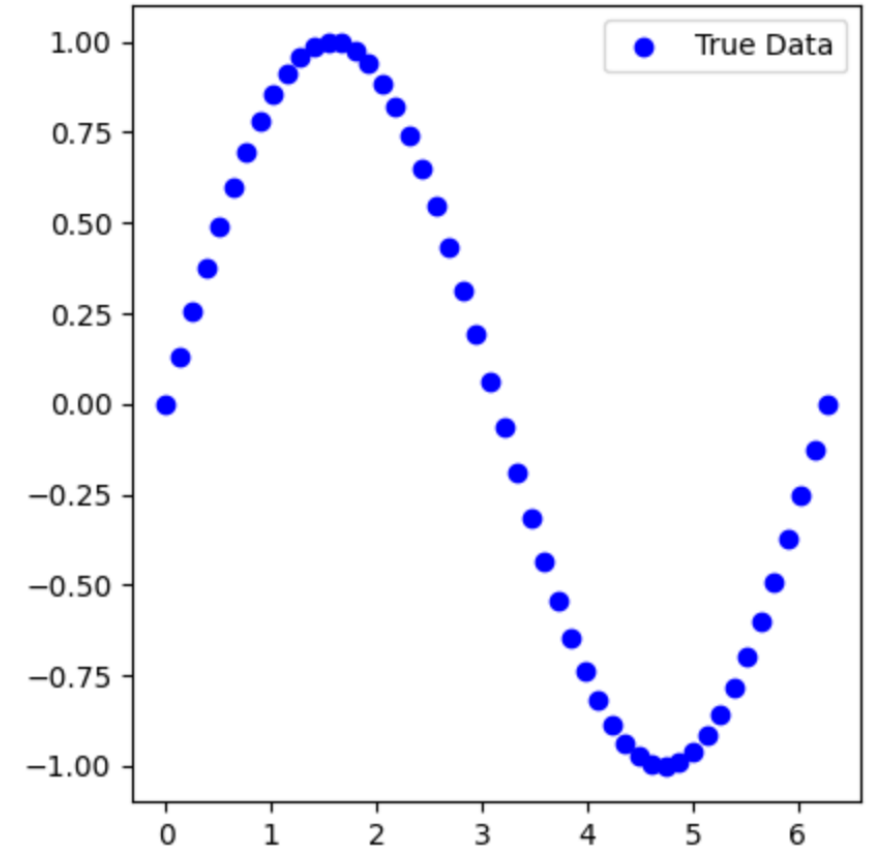
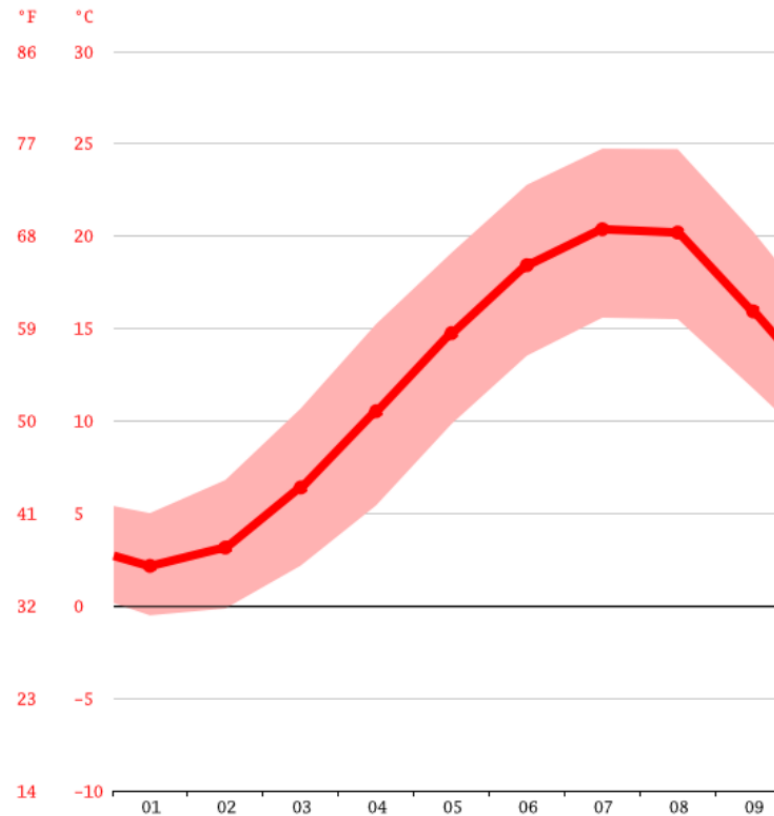


Flexibility of the toy network



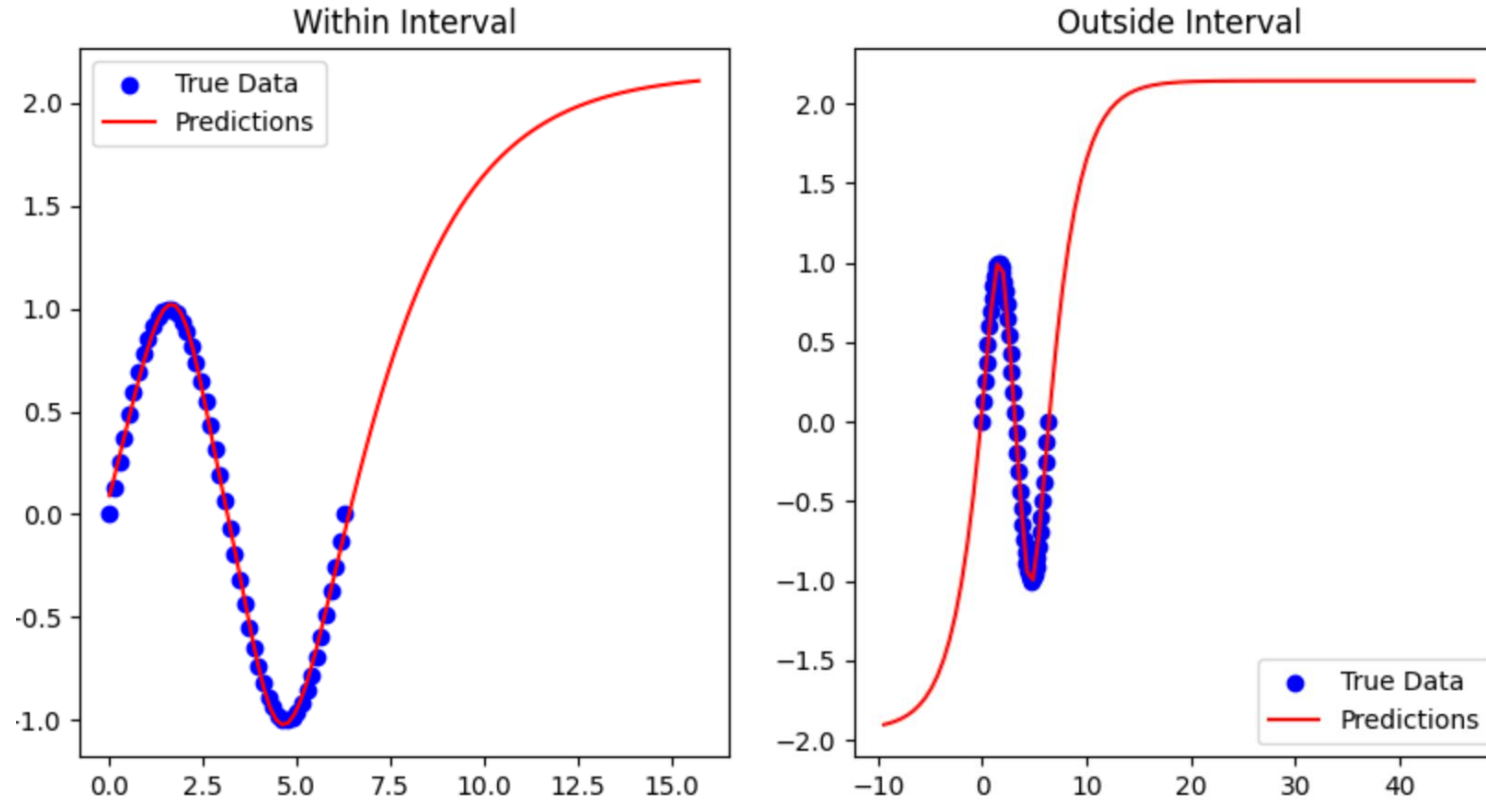
Approximating periodic data

A regression problem



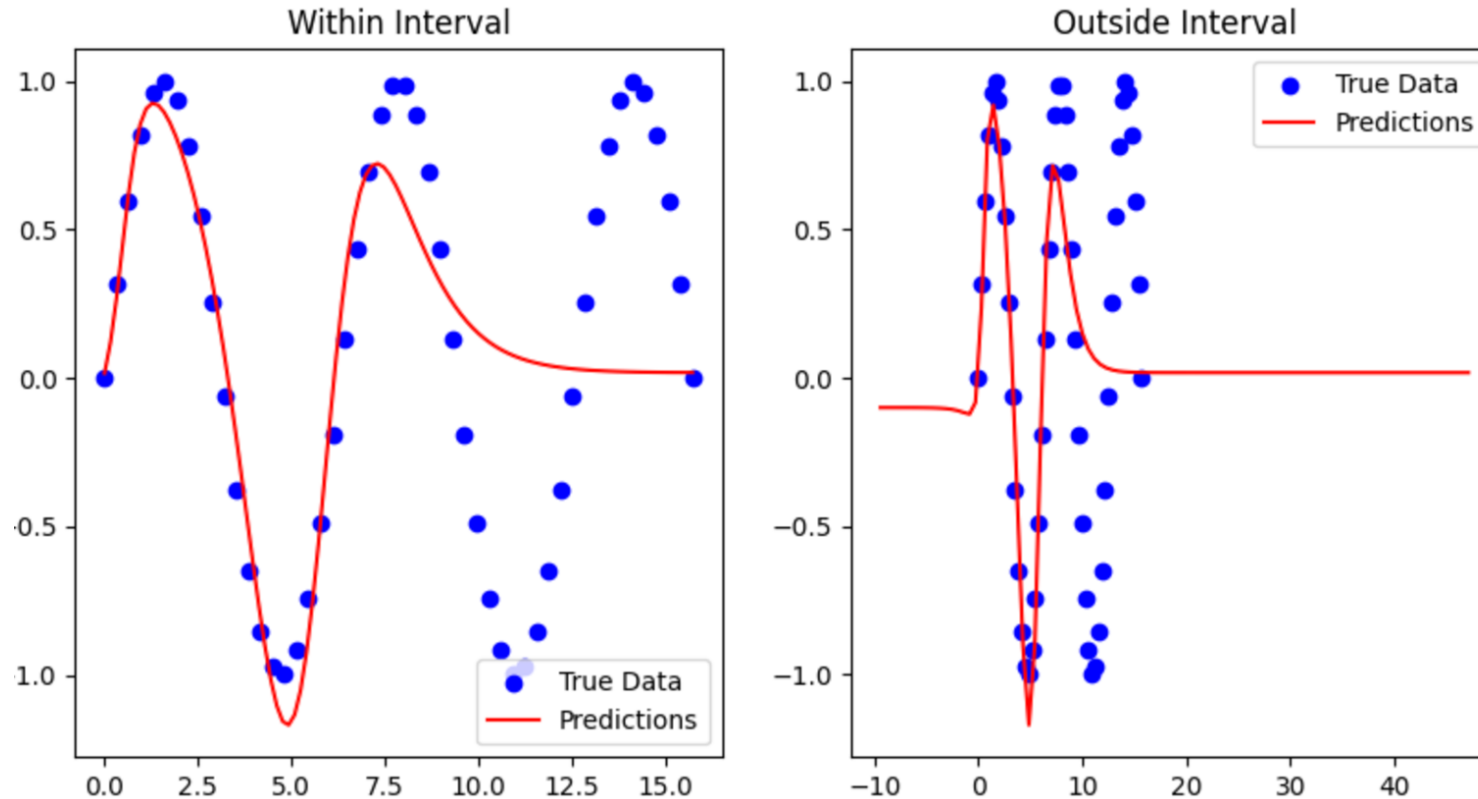
Approximating periodic data

A regression problem



Approximating periodic data

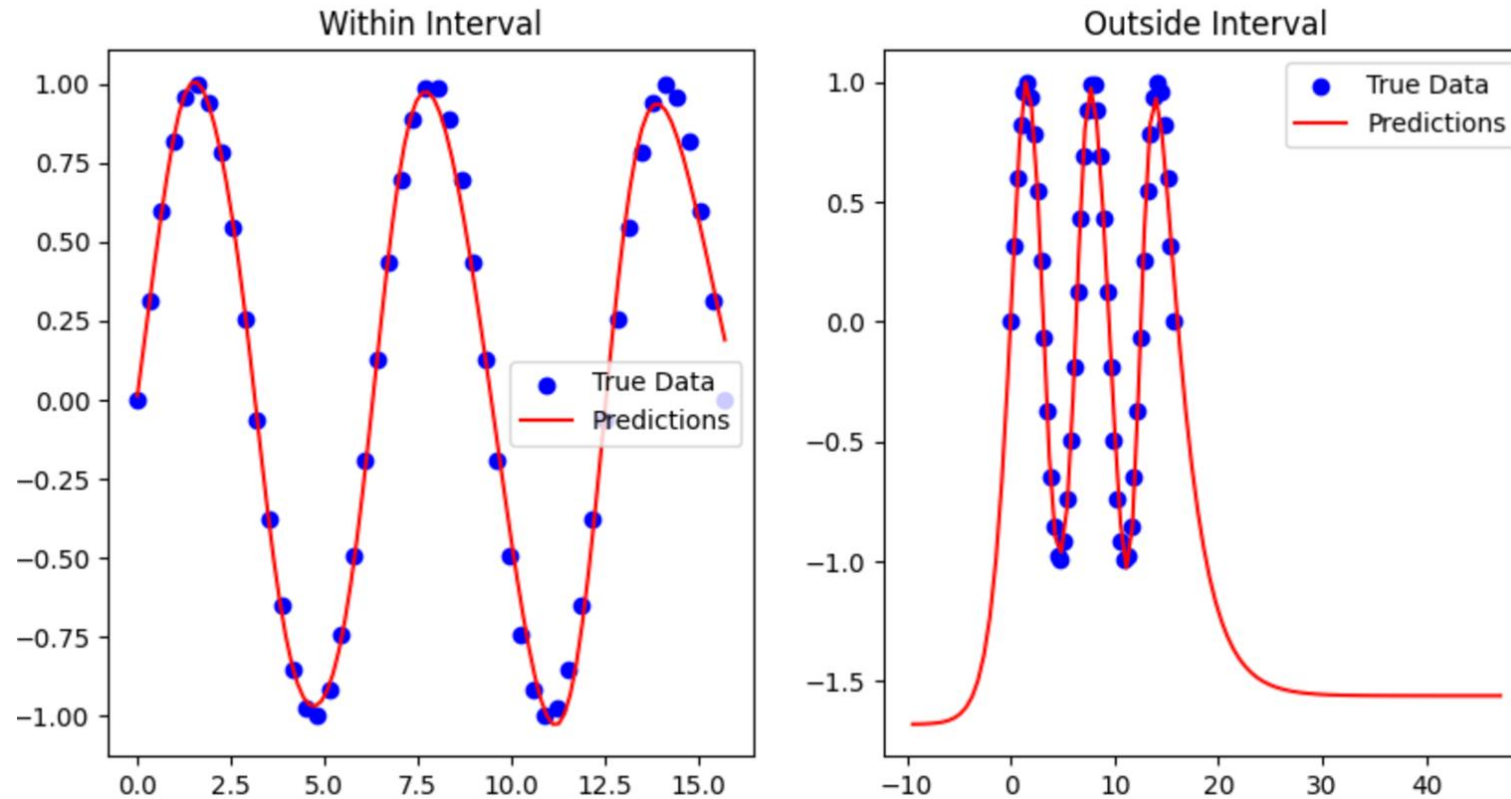
A regression problem



How can we modify the network for this task?

Approximating periodic data

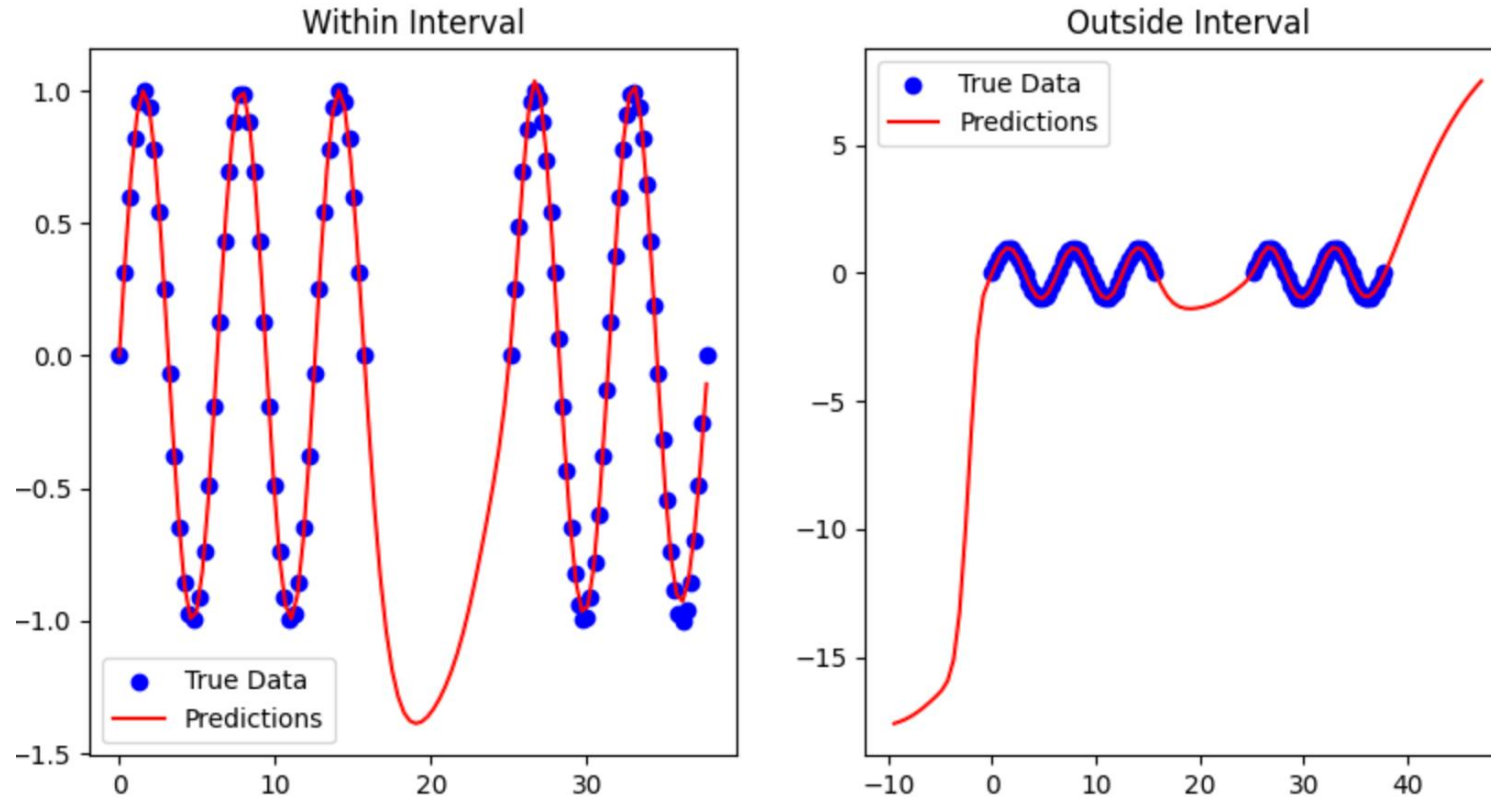
A regression problem



Network structure: 1 - 6 - 1

Approximating periodic data

A regression problem



Network structure: 1 - 500 - 1

What can students learn from this?

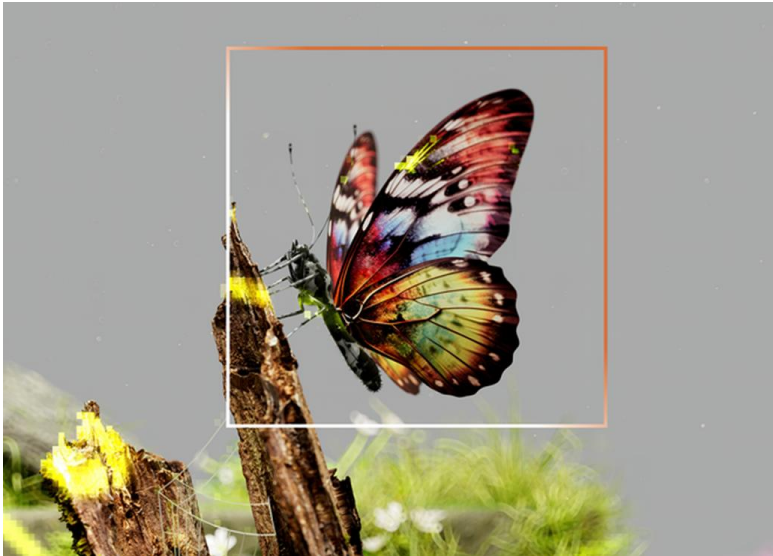
ABOUT ANN

- Recognize ANN as highly adjustable mathematical functions:
 - By adding nodes, any function can be approximated as precisely as desired, but only on a bounded set
 - The functions can approximate relationships in given data BUT the approximation only works in the intervals in which data points are given
- ANN models don't "understand"
- ANN models don't "know" when they don't know

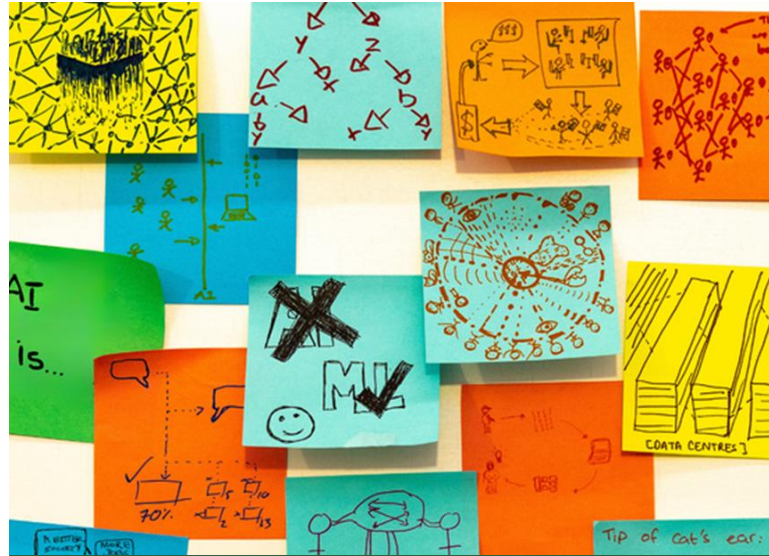
ABOUT MATHEMATICS

- Develop and train mathematical skills
 - e.g. functional thinking, finding extremes, ...

Our focus today



✓
A workshop on image classification and the Support Vector Machine



✓
A learning path to better understand Artificial Neural Networks



Conclusions about AI education and ways to implement

Black box vs. White box

Balancing black and white box approaches of AI models

SVM

- White Box: Mathematical description of separating lines / planes; idea of maximizing the “margin” (data-point free area)
- Black Box: Optimization algorithm

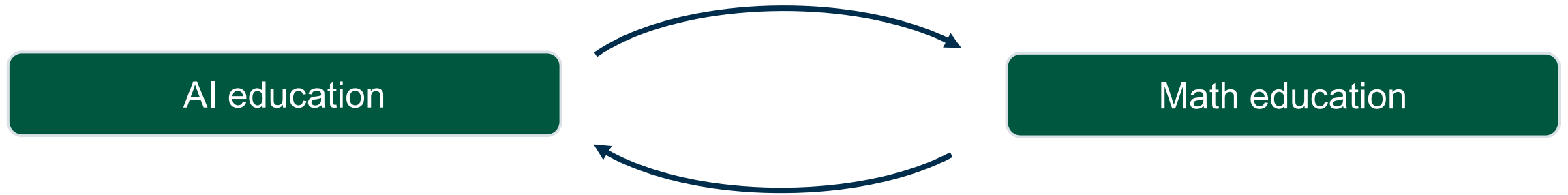
ANN

- White Box: Functionality of the ANN as a mathematical function
- Black Box: Optimization (“Training”) of the ANN



AI education in the math classroom: mutual benefits

In accordance with the AI competencies by Long & Magerko (2020)



Understanding the mathematical concepts of AI models enables...

- a deeper understanding of their results.
- a realistic assessment of opportunities and risks.
- an idea of the human role in their development.
- ...

AI education in mathematics education enables ...

- the classification of mathematics as a key qualification for modern applications.
- the application and development of school mathematical skills.
- interdisciplinary links with other school subjects.
- ...

Literature

- Bi, K., Xie, L., Zhang, H. et al. Accurate medium-range global weather forecasting with 3D neural networks. Nature 619, 533–538 (2023). <https://doi.org/10.1038/s41586-023-06185-3>
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- Touretzky, D., Gardner-McCune, C. & Seehorn, D. (2022). Machine learning and the five big ideas in AI. International Journal of Artificial Intelligence in Education, 1-34. <https://doi.org/10.1007/s40593-022-00314-1>
- UNESCO (2024). AI competency framework for students.

Thank you !

Access to learning material?
Suggestions and exchange?



www.cammp.online/english



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Discussion

REFLECTING THE APPROACH

- Are AI models a good learning content for the math classroom?
 - What advantages and challenges do you see?
-
- What can be transferred into your context?
 - What ethical or societal conversations does this type of hands-on understanding open up for students? (e.g. regarding last month's talk)

INTERDISCIPLINARY APPROACH

- How much math when teaching about AI in the CS classroom?
- How much Programming when teaching about AI in the math classroom?